

LANGUAGE MODELS ARE OPEN KNOWLEDGE GRAPHS

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ABSTRACT

This paper shows how to construct knowledge graphs (KGs) from pre-trained language models (e.g., BERT, GPT-2/3), without human supervision. Popular KGs (e.g., Wikidata, NELL) are built in either a supervised or semi-supervised manner, requiring humans to create knowledge. Recent deep language models automatically acquire knowledge from large-scale corpora via pre-training. The stored knowledge has enabled the language models to improve downstream NLP tasks, e.g., answering questions, and writing code and articles. In this paper, we propose an unsupervised method to cast the knowledge contained within language models into KGs. We show that KGs are constructed with a single forward pass of the pre-trained language models (without fine-tuning) over the corpora. We demonstrate the quality of the constructed KGs by comparing to two KGs (Wikidata, TAC KBP) created by humans. Our KGs also provide open factual knowledge that is new in the existing KGs. Our code and KGs will be made publicly available.

1 INTRODUCTION

Knowledge graphs (KGs) are an important resource for both humans and machines. Factual knowledge in KGs is injected into AI applications to imitate important skills possessed by humans, e.g., reasoning and understanding. KG construction is mainly supervised, requiring humans to handwrite every fact, such as Freebase (Bollacker et al., 2008) and Wikidata. KGs can also be constructed in a semi-supervised way, in which a semi-automatic extractor is used to obtain the facts from web corpora (e.g., NELL (Carlson et al., 2010) and Knowledge Vault (Dong et al., 2014)). Humans however still need to interact with the extractor to improve the quality of the discovered facts. Therefore, human supervision, which is often expensive, is required in constructing KGs.

Recent progress in language models (LMs), such as BERT (Devlin et al., 2018) and GPT-2/3 (Radford et al., 2019; Brown et al., 2020), has led to superior results even outperforming humans in a wide range of tasks, e.g., sentence classification (Wang et al., 2018), question answering (Brown et al., 2020). Pre-trained LMs are also capable to write poetry, music, and code, while such tasks often require we human to spend a significant amount of time in learning the relevant knowledge to work well. In fact, these pre-trained LMs automatically acquire factual knowledge from large-scale corpora (e.g., BookCorpus (Zhu et al., 2015), Common Crawl (Brown et al., 2020)) via pre-training. The learned knowledge in pre-trained LMs is the key to the current success. We therefore consider the following question: instead of using the manually created knowledge, *can we use the knowledge stored in pre-trained LMs to construct KGs?*

In this paper, we design an unsupervised approach called MAMA that successfully recovers the factual knowledge stored in LMs to build KGs from scratch. MAMA constructs a KG with a single forward pass of a pre-trained LM (without fine-tuning) over a textual corpus. As illustrated in Figure 1, MAMA has two stages: Match and Map. **Match** stage generates a set of candidate facts by matching the facts in the textual corpus with the knowledge in the pre-trained LM. General or world knowledge from large-scale corpora is embedded in the LM, thus candidate facts in the target corpus are often covered by the knowledge in the LM. The candidate facts are matched through an efficient beam search in the attention weight matrices of the pre-trained LM without fine-tuning. **Map** stage produces an open KG via mapping the matched candidate facts from Match stage to both fixed KG schema and open schema. If the schema of candidate facts exists in the KG schema, we map the candidate facts directly to the fixed KG schema. Otherwise, we reserve the unmapped candidate

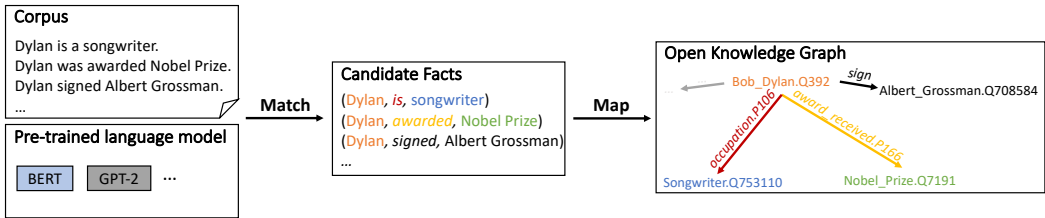


Figure 1: Overview of the proposed approach MAMA. MAMA constructs an open knowledge graph (KG) with a single forward pass of the pre-trained language model (LM) (without fine-tuning) over the corpus. Given the input: a textual corpus containing passages and sentences, e.g., English Wikipedia, and a pre-trained LM, e.g., BERT, GPT-2/3, MAMA (1) generates a set of candidate facts via *matching* the knowledge in the pre-trained LM with facts in the textual corpus, e.g., a candidate fact (*Dylan, is, songwriter*) from the sentence “Dylan is a songwriter.”, and (2) produces an open KG by *mapping* the matched candidate facts to both an existing KG schema, e.g., (*Bob_Dylan.Q392, occupation.P106, Songwriter.Q753110*) in Wikidata schema, and an open schema, e.g., (*Bob_Dylan.Q392, sign, Albert_Grossman.Q708584*).

facts in the open schema. This results in a new type of KG, *open KG*, with a mixture of mapped facts in fixed KG schema and unmapped facts in the open schema.

Our contributions are as follows:

1. We show *how to construct KGs from pre-trained LMs*. The KGs are constructed with a single forward pass of the pre-trained LMs without fine-tuning over the textual corpora. This helps researchers explicitly understand what the language models learn, bridging the deep LM and KG communities through enhanced model transparency.
2. We propose an unsupervised two-stage approach, MAMA, to first *match* the candidate facts in the corpora with the knowledge stored in LMs, then *map* the matched candidate facts to both fixed and open schema to produce a KG.
3. We generate a new type of KG, namely open KG, consists of *mapped facts* in the fixed KG schema of existing KGs (Wikidata and TAC KBP) annotated by humans; and *unmapped facts* in the open schema that are new in the reference KG schema. The reach of this result is broad and has downstream utility for knowledge graph construction, deep neural network interpretation, and information extraction.

2 MAMA

We introduce an unsupervised end-to-end approach Match and Map (MAMA) as illustrated in Figure 1 to construct open knowledge graphs (KGs) from language models (LMs). MAMA constructs the KGs with a single forward pass of the pre-trained LMs (without fine-tuning) over the corpora. The two stages of MAMA are:

Match generates a set of candidate facts from a textual corpus. LMs contain global or world knowledge learned from large-scale corpora, which often does not perfectly match the knowledge in the target corpus. The goal of this stage is to match the knowledge stored in pre-trained LMs with facts in the corpus. Each fact is represented as a triplet (*head, relation, tail*)¹, in short, (*h, r, t*), and passed to Map stage. Match procedure is detailed in Sec. 2.1

Map produces an open KG using the matched candidate facts from Match stage. The constructed open KG has two portions: (a) mapped candidate facts that are in a fixed KG schema, e.g., (*Dylan, is, songwriter*) is mapped to (*Bob_Dylan.Q392, occupation.P106, Songwriter.Q753110*) according to Wikidata schema; and (b) unmapped candidate facts that are in an open schema, e.g., a candidate fact (*Dylan, signed, Albert_Grossman*) is partially mapped to (*Bob_Dylan.Q392, sign, Albert_Grossman.Q708584*) in the open schema. This stage is described in Sec. 2.2

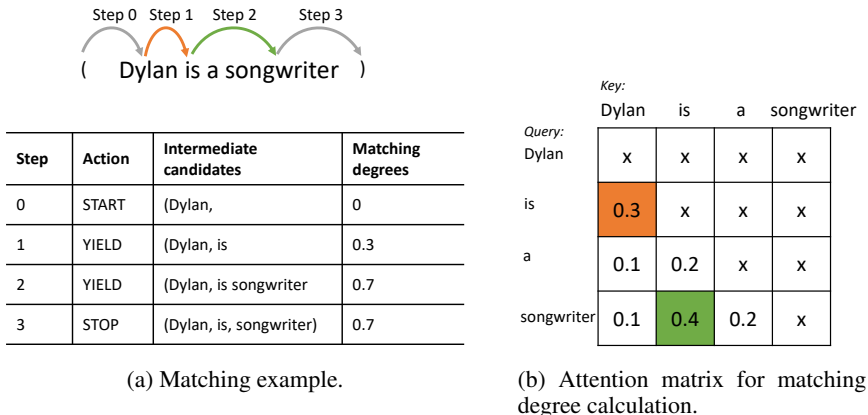


Figure 2: Illustration of Match stage. The upper part of (a) represents the general matching steps of generating the best matched candidate fact (*Dylan, is, songwriter*) from the sentence “Dylan is a songwriter.” The lower portion shows the corresponding step-by-step process. Given a head-tail pair (*Dylan, songwriter*), at each step, the search chooses one of the actions, i.e., START, YIELD, STOP to produce an intermediate candidate fact. The search starts by adding the head “Dylan” as an initial candidate (step 0). The matching degree of the candidate is initialized as 0. Next, a new candidate is yielded if the candidate has not reached the tail “songwriter” (step 1 and step 2), by appending the next largest attended token (with the largest score from the attention matrix (b) of the sentence) to the end of the current candidate, and the corresponding matching degrees are increased by the associated attention scores (0.3 and 0.4) to 0.3 (0+0.3) and 0.7 (0.3+0.4) respectively. Otherwise, the search stops, and the candidate fact with the best matching degree is returned for the head-tail pair (step 3). The attention matrix (b) is from the forward pass of the LM without fine-tuning over the sentence. “x” marks the tokens to prevent searching backward.

2.1 MATCH

We frame the matching procedure as a search problem. To obtain the best matched candidate facts of an input sentence, the candidates with the top matching degrees are returned from a search process. The matching degree is derived from the search in the attention weight matrices of the pre-trained LM, since the attention weight matrices are one of the main containers of the knowledge in the pre-trained LM. The attention weight matrices are simply from the forward pass of the LM without fine-tuning over the sentence.

2.1.1 BEAM SEARCH

We design a simple-yet-effective beam search to find the best matched candidate facts. For every head-tail pair (h, t) in a sentence, the search maintains the k -best matched candidate facts of the pair. Let’s first consider the search from left to right with beam size equals to 1. An example search process is shown in Figure 2. Given a head-tail pair (*Dylan, songwriter*), at each step, the search performs one of the following actions:

START the search from the head. The head h is added as an initial candidate into the beam. For simplicity, we use $START(h)$ to denote the action, which returns a candidate ($h,$. In Figure 2(a), at step 0, the head “Dylan” is added as (*Dylan,* into the beam. The matching degree is initialized to 0.

YIELD a new intermediate candidate in the beam if the current candidate has not reached the tail. The next largest attended token (with the largest score from the attention matrix) is appended to the end of the current candidate to yield the new candidate. The corresponding matching degrees are increased by the associated attention scores. At step 1 (orange arrow in Figure 2(a)), “is” is appended to the current candidate to yield (*Dylan, is, ,* since “is” has the largest attention score with “Dylan” in the attention matrix. The attention score is 0.3 as highlighted in orange in Figure 2(b). The matching degree becomes 0.3 (i.e. 0+0.3). The multi-head attention is reduced to a single head so that every two tokens of the sentence are associated with one attention weight. We experiment with different reduction

¹We use the term “head” and “tail” to denote head and tail’s “entities” or “entity mentions” for simplicity.

Algorithm 1 Beam search for matching candidate facts.

Input: Head-tail pair (h, t) , sentence s , attention matrix \mathbf{A}_s , action manager $\mathcal{O} = \{\text{START}, \text{YIELD}, \text{STOP}\}$, beam size k

Output: Candidate facts $\mathbb{T}_{(h,t)}$

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1:  $\mathbb{T}_{(h,t)} \leftarrow \{\text{START}(h)\}$  ▷ Start by adding the head as a candidate in the beam
2: while  $\exists c \in \mathbb{T}_{(h,t)}[\mathcal{O}(c) = \text{YIELD}]$  do
3:    $\tilde{\mathbb{T}}_{(h,t)} \leftarrow \emptyset$  ▷ Initialize a new beam
4:   for each  $c \in \mathbb{T}_{(h,t)}$  do
5:     if  $\mathcal{O}(c) = \text{YIELD}$  then
6:        $\tilde{\mathbb{T}}_{(h,t)} \leftarrow \tilde{\mathbb{T}}_{(h,t)} \cup \{\text{YIELD}(c, s, \mathbf{A}_s)\}$  ▷ Yield a new candidate if not reached the tail
7:     else
8:        $\tilde{\mathbb{T}}_{(h,t)} \leftarrow \tilde{\mathbb{T}}_{(h,t)} \cup \{\text{STOP}(c, t)\}$  ▷ Stop then produce a valid fact if reached the tail
9:     end if
10:  end for
11:   $\mathbb{T}_{(h,t)} \leftarrow \text{TOP}(k, \tilde{\mathbb{T}}_{(h,t)})$  ▷ Maintain  $k$ -best candidates in the beam
12: end while
13: return  $\mathbb{T}_{(h,t)}$ 

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setups in Sec. A.3 “x” marks the tokens (prior to the current token) that are not considered in the search to prevent searching backward. Step 2 similarly takes YIELD action to produce *(Dylan, is songwriter, .* The matching degree is now 0.7 (i.e. 0.3+0.4). We use YIELD(c, s, \mathbf{A}_s) to denote the action, where c is a current candidate, s represents the sentence, and \mathbf{A}_s is the attention matrix from the forward pass of the pre-trained LM over s , which yields a new candidate.

STOP the search step if the candidate has reached the tail, then add the candidate as a valid candidate fact into the beam. As beam size equals to 1, *(Dylan, is, songwriter)* is the only returned candidate fact for the given pair. The final matching degree of the candidate is 0.7. We denote this step using STOP(c, t), which returns a valid fact.

The details of the proposed beam search are in Algorithm 1. The inputs of the search algorithm are a head-tail pair (h, t) , a sentence s , an attention matrix \mathbf{A}_s of s . Both h and t are identified by the noun chunk in s . \mathbf{A}_s is the attention matrix associated with s from the forward pass of LM without fine-tuning. The search gets started by adding the head h as the initial candidate in the beam (line 1). While there are still new candidates waiting to be yielded (line 2), the search continues, and the top k candidates sorted by the matching degrees are maintained (line 3-11) in the beam. In practice, we implement an action manager \mathcal{O} to decide which action to take at each step. Given a candidate c in the beam, $\mathcal{O}(c) = \text{START}$ always happens at the beginning of the search. If c has not reached the tail t yet, $\mathcal{O}(c) = \text{YIELD}$. Otherwise, $\mathcal{O}(c) = \text{STOP}$. We convert the subwords to the corresponding full words. We also notice some facts are in reverse order in the sentence, e.g., “... said Jason Forcier, a vice president at battery maker A123 Systems Inc.” for facts of relation “org:top_members_employees”, thus enable bidirectionality by running the algorithm in both directions (left to right and right to left). The beam search is implemented by the breadth-first search, which is efficient as the time complexity is $O(k \cdot d)$, where d is the maximum depth of the search tree.

2.1.2 FILTER

Although the basic functionality provided by beam search is sufficient for finding useful candidate facts, we have found a few constraints useful. Given a candidate fact (h, r, t) from beam search result $\mathbb{T}_{(h,t)}$, it remains as a fact if satisfying all the following constraints.

Constraint #1 The matching degree of (h, r, t) is above a threshold. We compare the matching degrees corpus-wide to only reserve the facts that are matched better with the knowledge in LMs. For example, MAMA extracts a fact *(Rolling Stone, wrote, pop song)* from “Rolling Stone wrote: “No other pop song has so thoroughly challenged artistic conventions””, which is not an accurate fact based on the sentence. We observe the associated matching degree is below a proper threshold, while the matching degrees of high-quality facts from the same documents, e.g., *(Dylan, is, songwriter)*, or confident facts from the other documents are beyond the threshold.

Constraint #2 The distinct frequency of r is above a threshold. To avoid facts to be over-specified, e.g., (*Dylan, signed to Sam Peckinpah’s film, Pat Garrett and Billy the Kid*), we require r should take many distinct head-tail pairs in the corpus.

Constraint #3 Relation r is a contiguous sequence in the sentence. We can avoid r that has no meaningful interpretation (Fader et al., 2011), e.g., (*Rolling Stone, wrote challenged, conventions*) from the above sentence.

2.2 MAP

The objective of Map stage is to generate an open KG. The open KG contains (a) *mapped facts* in a KG schema (Sec. 2.2.1), e.g., Wikidata schema, if the schema of the candidate facts is within the existing KG schema; and (b) *unmapped facts* from (a) in an open schema (Sec. 2.2.2).

2.2.1 MAPPED FACTS IN KG SCHEMA

The goal is to map a candidate fact (h, r, t) to a fact (h_k, r_k, t_k) in the KG schema. The reason for mapping to an existing KG schema is to make use of the high-quality schema designed by experts (to avoid duplicated efforts of building from scratch) and enable evaluating the candidate facts with oracle KG facts contributed by human volunteers. We first map both entities h, t to h_k, t_k , then map the relation r to r_k in the reference KG schema. Additional details are presented in Sec. A.1.

Entity linking to KG schema We adapt an unsupervised entity linker based on a mention-to-entity dictionary (Spitkovsky & Chang, 2012) to link the entities for scalability consideration. Besides, contextual information is crucial to link the entities correctly, we use the word embedding of the context to disambiguate the entities, which means we only link the entities with high contextual similarities based on the word embedding. We adopt the entity linker to map h, t to h_k, t_k .

Relation mapping with KG schema We largely follow the relation mapping method proposed by Angeli et al. (2015) to construct an offline relation map between KG relation and relation phrases of the candidate facts. The basic idea is that the more often linked head-tail pairs (i.e., entities are with type information from the entity linking step) co-occur between the candidate facts and KG facts, the more likely the corresponding relations are mapped to each other. In addition, we normalize each relation phrase of the candidate facts by lemmatization, and removing inflection, auxiliary verbs, adjectives, adverbs. After the relation mapping is constructed, one author manually checks whether the top 15 relation phrases are true mappings for each KG relation. We only reserve the true ones in the final relation mapping. This process takes approximately one day. Later, r is mapped to r_k with an efficient look-up operation in the relation mapping.

2.2.2 UNMAPPED FACTS IN OPEN SCHEMA

An unmapped candidate fact (h, r, t) means at least one of h, r, t is not mapped to the KG schema based on the method in Sec. 2.2.1. There are two types of unmapped candidate facts:

Partially unmapped facts represent at least one of h, r, t are mapped to the KG schema. It can be h or t mapped to h_k or t_k based on the entity linker in Sec. 2.2.1. It can also be r that mapped to r_k using the relation mapping in Sec. 2.2.1. This actually results in unmapped facts that are in a mixture of the KG schema and the open schema. As the overall schema of the unmapped facts is not in the KG schema, we use open schema to denote such unmapped facts in the rest of the paper for simplicity. An example is (*Dylan, signed, Albert Grossman*) in Figure 1, where both head and tail are linked to Wikidata schema based on the entity linker in Sec. 2.2.1 but the relation cannot be mapped since there is no relation mapping from “signed” to a KG relation in Wikidata schema.

Completely unmapped facts indicate all h, r, t are not mapped to the KG schema. This means neither the entity linker nor the relation mapping is able to map h, r, t to h_k, r_k, t_k respectively. The resulting unmapped candidate facts stay in the open schema, e.g., a candidate fact (*Jacob, was, A Registered Mennonite*) stays the same in the open schema from a sentence “Jacob was a registered Mennonite in Amsterdam.”.

The resulting open KG is a new type of KG that mixes the fixed KG schema with the flexible open schema, suggesting new directions for the next generation of KGs. The open KG not only contains existing knowledge in the reference KGs, but also extends the fixed KG schema with an additional open schema to improve the coverage of the KGs, that benefits the downstream KG based applications, e.g., QA and commonsense reasoning (Wang et al., 2019; Brown et al., 2020).

KG	# of oracle facts	# of documents
TAC KBP	27,655 ³	3,877,207
Wikidata	27,368,562	6,047,494

Table 1: Dataset statistics of two knowledge graphs: TAC KBP and Wikidata. TAC KBP refers to TAC KBP Slot Filling 2013 challenge. # of oracle facts for TAC KBP is the number of oracle facts in the 2013 task. # of documents for TAC KBP is the number of the documents in the 2013 task. # of oracle facts for Wikidata is the total number of oracle facts in Wikidata. # of documents for Wikidata is the size of English Wikipedia.

3 EXPERIMENTS

How well can language models generate knowledge graphs? We experimentally explore how well can MAMA answer the question in the section. To measure the ability of LMs in generating KGs, we directly measure the quality of resulting open KGs. The open KG contains two types of facts: mapped facts in the fixed KG schema; and unmapped facts in the open schema. We first quantitatively evaluate MAMA by comparing the mapped facts to oracle KGs annotated by humans in Sec. 3.1 then conduct an in-depth analysis of the unmapped facts in Sec. 3.2

3.1 RESULTS ON MAPPED FACTS

We first study the quality of the mapped facts. As the candidate facts have been mapped to the schema of oracle KGs, we are able to quantitatively compare the candidate facts with the oracle facts in the reference KGs.

3.1.1 DATASETS

We compare the mapped facts from MAMA with the facts in two KGs:

TAC KBP TAC Knowledge Base Population (KBP) Slot Filling is a task to search a document collection to fill in the tail/object entity for predefined relations (slots) for a given head/subject entity in a reference KG. We experiment with the reference KG in the 2013 challenge. We use the document collection and oracle facts of the 2013 task. The statistic of the dataset is shown in Table 1

Wikidata We use popular Wikidata as another KG. We use all the oracle facts in Wikidata. We use the English Wikipedia as the text corpus, since a large amount of facts in Wikidata is from English Wikipedia. The statistic is in Table 1

To evaluate the mapped facts, we first use Match stage of MAMA to run over the corresponding documents to generate the candidate facts. Then Map stage is leveraged to map the candidate facts to the schema of TAC KBP and Wikidata respectively. The parameter settings, such as beam size in Algorithm 1, are shared across TAC KBP and Wikidata based on the parameter study in Sec. A.3.

3.1.2 TAC KBP

To verify the ability to produce correct facts, we compare candidate facts from MAMA to the outputs of two open information systems, which also produce triplets in the form of (h, r, t) . After collecting the triplets from the corresponding system, we use the same Map procedure with MAMA (Sec. 2.2.1) to map the triplets to the corresponding KG schema.

Stanford OpenIE leverages POS tag and dependency parser, and generates self-contained clauses from long sentences to extract the triplets, which is the best open information extraction system (Angeli et al., 2015) on TAC KBP (Surdeanu, 2013).

OpenIE 5.1² is one of the state-of-the-art open information extraction systems, which is the successor to Ollie (Schmitz et al., 2012), and it improves extractions from noun relations, numerical sentences, and conjunctive sentences depending on the linguistic patterns.

We use two families of pre-trained LMs with MAMA. We use BERT_{BASE} and BERT_{LARGE} from Devlin et al. (2018) with MAMA, namely MAMA-BERT_{BASE} and MAMA-BERT_{LARGE}. Besides, GPT-2s from Radford et al. (2019) are used, i.e., MAMA-GPT-2, MAMA-GPT-2_{MEDIUM}, MAMA-GPT-2_{LARGE}, and MAMA-GPT-2_{XL}.

²<https://github.com/dair-iiitd/OpenIE-standalone>

Method	#Params of LM	Precision%	Recall%	F1%
OpenIE 5.1 ²	-	56.98	14.54	23.16
Stanford OpenIE (Angeli et al., 2015)	-	61.55	17.35	27.07
MAMA-BERT _{BASE} (ours)	109M	61.57	18.79	28.79
MAMA-BERT _{LARGE} (ours)	335M	61.69	18.99	29.05
MAMA-GPT-2 (ours)	117M	61.62	18.17	28.07
MAMA-GPT-2 _{MEDIUM} (ours)	345M	62.10	18.65	28.69
MAMA-GPT-2 _{LARGE} (ours)	774M	62.38	19.00	29.12
MAMA-GPT-2 _{XL} (ours)	1558M	62.69	19.47	29.72 (+2.65)

Table 2: Compare the quality of mapped facts on TAC KBP. #Params of LM refers to the number of parameters of the pre-trained LM.

Table 2 shows the results on TAC KBP. We use the official scorer of TAC KBP Slot Filling 2013 to evaluate precision, recall, and F1 on TAC KBP ³

MAMA constructs improved KGs compared to open IE. From the results, we find that all our methods achieve competitive precision, which is greater than 60%, given the unsupervised nature of MAMA. All the proposed methods outperform the two open IE systems. This shows that MAMA is able to produce high-quality knowledge directly from pre-trained LMs by a single forward pass without human supervision. The results show the effectiveness of MAMA in generating candidate facts from Match stage, and producing high-quality KGs through Map stage. We also find that MAMA-GPT-2_{XL} performs the best. MAMA-GPT-2_{XL} outperforms the previous state-of-the-art Stanford OpenIE by over 2.6% in F1. This shows the proposed end-to-end MAMA is able to recover the knowledge stored in pre-trained LMs without relying on any extra linguistic features, such as POS tag and dependency parser used in open IE systems. The main reason leading to the moderate results of OpenIE 5.1 is that the system generates objects of the triplets with extraneous words, which hurt the performance in slot filling tasks. Even though the proposed methods all outperform the two open IE systems in the recall, however improving recall is clearly the future direction to further improve the performance of MAMA. We find that the main cause of the moderate recall is the incorrect entities caused by spaCy noun chunk as summarized in Sec. A.2.

Larger/deeper LMs produce KGs of higher quality. BERT_{LARGE} outperforms BERT_{BASE} since the doubling parameter size. GPT-2s share similar trends, where we observe performance increases when the model size increases. This complies with our intuition on more knowledge is stored in deeper and larger models. Such increases in performance seem subtle on TAC KBP, we find this might due to the relatively small number of oracle facts by noticing a more significant improvement on Wikidata in Sec. 3.1.3. We plan to further improve the results with larger pre-trained LMs, e.g., GPT-3 (Brown et al., 2020), Megatron-LM (Shoeybi et al., 2019).

BERT LMs outperform GPT-2 LMs under similar model sizes. More specifically, BERT_{BASE} performs better than MAMA-GPT-2 in F1, and MAMA-BERT_{LARGE} outperforms MAMA-GPT-2_{MEDIUM} in F1. BERT_{BASE} and MAMA-GPT-2 are similar in size, while MAMA-BERT_{LARGE} and MAMA-GPT-2_{MEDIUM} are similar in model size as well. This is mainly because that the recall of BERT LMs is higher than that of corresponding GPT-2 LMs. The result indicates that the Cloze-style loss function (i.e., masked language model) of BERT is more effective and flexible in recovering more knowledge than the autoregressive LM objective. We also notice that the precision of GPT-2 LMs is higher than that of according BERT LMs. The reason is that the autoregressive LM objective captures more accurate knowledge than Cloze-style loss does by not introducing extra noise (e.g., masked tokens) during pre-training.

3.1.3 WIKIDATA

We select our best BERT based method MAMA-BERT_{LARGE}, and GPT-2 based method MAMA-GPT-2_{XL} on TAC KBP to compare with Stanford OpenIE (the best open IE system on TAC KBP) for scalability experiments on Wikidata. We follow the same definition as the slot filling task to calculate precision, recall, and F1 on Wikidata. Table 3 summarizes the results.

MAMA is scalable to larger corpora. Similar to the trends on TAC KBP, MAMA-GPT-2_{XL} performs the best in precision, recall, and F1. The results show the effectiveness of MAMA in generating candidate facts and high-quality KGs. We also find that MAMA-GPT-2_{XL} outperforms MAMA-BERT_{LARGE} by over 1% in F1. This shows that the larger model (GPT-2_{XL} has 5x more

³There are 2,383 correct oracle facts based on the “manual runs” assessment in TAC KBP.

Method	#Params of LM	Precision%	Recall%	F1%
Stanford OpenIE (Angeli et al., 2015)	-	23.32	13.09	16.77
MAMA-BERT _{LARGE} (ours)	335M	29.52	16.56	21.22
MAMA-GPT-2 _{XL} (ours)	1558M	31.32	17.42	22.39 (+5.62)

Table 3: Compare the quality of mapped facts on Wikidata. #Params of LM refers to the number of parameters of the pre-trained LM.

parameters compared to BERT_{LARGE}) contains more knowledge, and MAMA is able to restore the knowledge. When larger or deeper models (e.g., GPT-3) are used with MAMA, we can expect more gains of the KG quality. Thanks to the efficient nature of MAMA, which relies only on the forward pass of the LMs without fine-tuning, the results suggest that MAMA is scalable to large KGs.

Larger corpora embed more complete KGs. In particular, MAMA-GPT-2_{XL} outperforms Stanford OpenIE by 5.6% in F1. MAMA-BERT_{LARGE} outperforms Stanford OpenIE by approximately 4.4% in F1. Both F1 gains are larger compared to that on TAC KBP. This is because that the LMs contain world knowledge from pre-training corpora, e.g. Wikipedia and Common Crawl. The larger the textual corpora are, the more knowledge our method is able to recover and match to the knowledge stored in LMs. The finding is particularly important, since we are now able to construct larger KGs of high quality from scratch when larger datasets are used, such as WebText2 and Common Crawl (Raffel et al., 2019; Brown et al., 2020). Similar to the observations on TAC KBP, the precision is higher compared to recall. Wikidata is not fully built from Wikipedia, MAMA could improve the recall by running on those larger corpora to collect more facts.

3.2 ANALYSIS OF UNMAPPED FACTS

The open KG constructed by MAMA is a new type of KG combining the fixed KG schema with the flexible open schema. We turn to study the quality of the candidate facts that are not mapped to the above reference KG schema, but are in the open schema generated by MAMA. We manually judge such unmapped facts generated by our best method MAMA-GPT-2_{XL} from 100 sampled documents in Wikidata and TAC KBP respectively.

The quality of unmapped facts is verified by human annotators. We find 35.3% of the unmapped facts are true on Wikidata. We find 83.2% of those true facts are partially unmapped facts as defined in Sec. 2.2.2 e.g., (*Bob_Dylan.Q392, tour_with, the_Grateful_Dead.Q212533*), whose relation is not within the schema of Wikidata, while both head and tail are in the schema. The remaining true facts are completely unmapped facts (Sec. 2.2.2), e.g., a candidate fact (*Jacob, was, A Registered Mennonite*) stays the same in the open schema.

Accurate entity detection is desired. We also notice 45.5% of the untrue unmapped facts on Wikidata are due to the *incorrect entities* detected by the spaCy. *Incorrect or missing entity linking* (to either head or tail) in Sec. 2.2.1 causes additional 9.1% errors in the unmapped facts. 4.5% of the untrue unmapped facts are caused by the *missing relation mapping* in Sec. 2.2.1. The rest errors made by MAMA-GPT-2_{XL} are *incorrect relation phrases*, such as uninformative relation phrases, e.g., (*Dylan, made, his breakthrough*), which is similar to the errors made by open IE systems (Fader et al., 2011). Both entity linking and relation mapping of Map stage rely heavily on the accuracy of entity detection from the spaCy noun chunk. We conclude that the main root cause of the untrue unmapped facts is due to the errors made by the spaCy noun chunk.

We observe similar trends on TAC KBP. We plan to leverage crowdsourcing platforms, e.g., Mechanical Turk, to conduct quantitative evaluations over the unmapped facts to better understand the strengths and shortage of MAMA. We plan to identify more accurate entities by relying on attention weights in LMs (Clark et al., 2019; Hewitt & Manning, 2019) instead of using extra resources. We will also investigate stronger entity linkers (Kolitsas et al., 2018) and learn a more robust relation mapping through weak or distant supervision (Mintz et al., 2009; Ratner et al., 2017). We will investigate more sophisticated approaches, such as graph neural networks (Kipf & Welling, 2016), to generate more accurate relation phrases from the attention weight matrices by considering structural information.

4 RELATED WORK

Knowledge graph construction can be generally categorized into two groups, 1) supervised approaches. Wikidata, Freebase (Bollacker et al., 2008), YAGO (Suchanek et al., 2007), YAGO2 (Hof-

fart et al. [2013], DBpedia [Auer et al. 2007] are built based on human supervision from Wikipedia infoboxes and other structured data sources; 2) semi-supervised approaches. Open information extraction systems, e.g., OLLIE [Schmitz et al. 2012], Reverb [Fader et al. 2011], Stanford OpenIE [Angeli et al. 2015], and OpenIE 5.1 [] aim to leverage carefully-designed patterns based on linguistic features (e.g., dependencies and POS tags), to extract triplets from web corpora for open schema KG. Besides, NELL [Carlson et al. 2010], DeepDive [Niu et al. 2012], Knowledge Vault [Dong et al. 2014] extract information based on a fixed schema or ontology, where humans help improve the accuracy of the extractions. Probase [Wu et al. 2012] produces taxonomies instead of rich typed relations in general KGs. MAMA instead uses learned knowledge stored in pre-trained LMs without human supervision to construct an open KG, which is a mixture of fixed schema and open schema. Different from commonsense knowledge construction using Transformers [Davison et al. 2019] [Bosselut et al. 2019], the proposed method is unsupervised and end-to-end, and constructs general-purpose KGs instead of commonsense knowledge.

Language models, e.g., BERT [Devlin et al. 2018], GPT [Radford et al. 2018], GPT-2/3 [Radford et al. 2019] [Brown et al. 2020], ELMo [Peters et al. 2018], Transformer-XL [Dai et al. 2019], ALBERT [Lan et al. 2019], RoBERTa [Liu et al. 2019], XLNet [Yang et al. 2019] and Megatron-LM [Shoeybi et al. 2019] contain factual knowledge obtained via pre-training on large-scale corpora such as Wikipedia and BookCorpus [Zhu et al. 2015]. Studies have leveraged the pre-trained LMs as virtual KGs, and show reasonable performance in QA tasks [Dhingra et al. 2020] [Guu et al. 2020], and language modeling [Khandelwal et al. 2019]. LMs are further enhanced by KGs [Peters et al. 2019] to improve knowledge-driven tasks. While the existing work utilizes knowledge in an implicit way, the main difference is that our approach explicitly extracts knowledgeable facts from the LMs. Compare to the joint training with knowledge base to improve shallow word embedding [Wang et al. 2014], we show that the knowledge is already stored in the deep LMs. We plan to incorporate domain knowledge into language models to construct domain-specific KGs. The main difference between LAMA [Petroni et al. 2019; 2020] and MAMA is mainly two-fold: (1) LAMA aims to complete Cloze-style statement, e.g., given “Dylan is a _”, LAMA predicts which words/phrases should fill the blank “_”, which has no direct connection to KGs. There are several fundamental limitations when adapting LAMA to construct KGs, e.g., additional queries must be constructed first, and the answers for the queries must be linked to KGs. MAMA aims to solve a reasoning problem, e.g., given a passage, MAMA directly matches the fact in the form of a triplet (*Dylan, is, songwriter*) at the first step, then maps the fact to produce a KG. (2) The benchmark datasets used with MAMA are larger compared to the LAMA benchmark, e.g., Wikidata is 3 orders of magnitude larger compared to the largest dataset in the LAMA benchmark.

Neural network interpretation here specifically refers to pre-trained deep language model analysis. There has been a lot of work to understand what the neural networks learn [Linzen et al. 2016] [Adi et al. 2016] [Tenney et al. 2019]. With regards to analyzing Transformer [Vaswani et al. 2017] based language models (e.g., BERT and GPT-3), substantial recent work focuses on both visualizing and analyzing the attention [Vig 2019; Jain & Wallace 2019; Clark et al. 2019; Michel et al. 2019; Vig et al. 2020; Ramsauer et al. 2020; Hendrycks et al. 2020]. Instead of analyzing or visualizing, we use LMs to generate structured KGs to directly recover what LMs learn from the corpora.

5 CONCLUSION

We show that the knowledge graphs can be constructed by a single forward pass of the language models over textual corpora. We propose a two-stage unsupervised approach MAMA to first match the facts in the corpus with the internal knowledge of the language model, and then map the matched facts to produce a knowledge graph. We demonstrate the quality of the resultant open knowledge graphs by comparing to two knowledge graphs (Wikidata and TAC KBP). The open knowledge graph also features new facts in the open schema, which could have broad implications for knowledge graphs and their downstream applications. The results also suggest that larger language models store richer knowledge than existing knowledge graphs, and generating on even larger high-quality text corpora could continue improving knowledge graphs. Additionally, the knowledge graphs generated by our approach can help researchers to look into what the language models learn, so our interpretable knowledge graphs establish a bridge between the deep learning and knowledge graph communities.

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A ADDITIONAL DETAILS AND ANALYSIS OF MAMA

A.1 METHOD DETAILS

Map stage details To evaluate the mapped facts, we first use Match stage of MAMA to run over the corresponding documents to generate the candidate facts. For Map stage on TAC KBP, we link to the oracle annotation of the entities or spans in the TAC KBP corpus. On Wikidata, the entity linking method described in Sec. 2.2.1 is first leveraged to link entities in the candidate facts to Wikipedia anchors. We build an enhanced mention-to-entity dictionary based on Spitzkovsky & Chang (2012). In particular, we add new Wikipedia anchors to the dictionary which results in 26 million entries comparing to 21 million entries in Spitzkovsky & Chang (2012). Then a Wikipedia anchor to the Wikidata item dictionary is constructed and used to further link the entities to Wikidata. If the head or tail is a pronoun, we further use neuralcoref⁴ for coreference resolution. We use GloVe (Pennington et al., 2014) embedding for disambiguation. The relation mapping is constructed offline for TAC KBP and Wikidata respectively using the method in Sec. 2.2.1. Besides the automatic relation mapping method proposed in Angeli et al. (2015), we manually check whether the top relation phrases are true as described in Sec. 2.2.1. For relation mapping, we randomly sampled a hold-out dataset including 2,000 documents from the TAC KBP corpus and English Wikipedia for the relation mapping construction on TAC KBP and Wikidata respectively. For oracle facts in Wikidata, we only preserve those facts describing relations between entities that could be linked to corresponding Wikipedia anchors. We rule out facts of attributes about entities and facts of auxiliary relations (such as *topic’s main category:P901*) and finally results in 27,368,562 oracle facts.

Implementation details For Wikidata, at Match stage, we randomly split the English Wikipedia data into 20 partitions, and map the data partitions to 20 distributed servers to run. Each server is configured with four Tesla K80 12Gs. We set the max sequence length to 256, and batch size as 32 for MAMA-BERT_{LARGE} and 4 for MAMA-GPT-2_{XL}. We use implementations of pre-trained LMs in Transformers package⁵. We use spaCy sentencizer⁶ to segment the documents into sentences. MAMA-BERT_{LARGE} takes approximately 48 hours, and MAMA-GPT-2_{XL} costs around 96 hours. The resulting candidate facts of Match stage from the 20 servers are then reduced a data server, where a MongoDB database is maintained to store the oracle Wikidata and entity linking results to enable the efficient Map stage. To produce the open KGs, Map stage takes around 18 hours. The setup is similar to TAC KBP. Match stage is done within 48 hours for all the settings. The batch sizes of MAMA-BERT_{BASE}, MAMA-GPT-2, MAMA-GPT-2_{MEDIUM}, MAMA-GPT-2_{LARGE} are 64, 32, 16, 8 respectively.

Parameter settings The parameter settings are shared across TAC KBP and Wikidata. All the choices are based on the parameter study in Sec. A.3. The beam size of Algorithm 1 is set to 6. The matching degree threshold of Constraint #1 (Sec. 2.1.2) is set to 0.005, and the number of distinct head-tail pairs of Constraint #2 (Sec. 2.1.2) is set to 10. To generate the attention weight matrix A_s of a sentence, we reduce the weights of every attention head in the last layer of pre-trained LMs using the mean operator.

A.2 ERROR ANALYSIS

There is still significant room to improve MAMA. To further understand the shortage of MAMA, we conduct an error analysis of the errors in precision (i.e., incorrect facts returned by MAMA) of Table 2 and Table 3. We choose our best method MAMA-GPT-2_{XL} for the study. We sample 100 documents from the Wikidata dataset, and manually check the reasons for the errors. We find 33.1% of the errors are caused by *incorrect entities*, while the relation phrases are correct. The errors are due to the incorrect noun chunk detected by the spaCy⁷. 18.3% of the errors are due to the *missing relation mapping* created in Sec. 2.2.1. Note that we find approximately 23.8% of the errors are actually *correct facts that are new in the reference KGs*. e.g., (*Bob_Dylan.Q392, residence.P551, Nashville.Q23197*) (in Figure 16) is not an existing fact in Wikidata, but it is a correct mapped fact based on our annotation. The rest errors made by MAMA-GPT-2_{XL} are *incorrect relation phrases*,

⁴<https://github.com/huggingface/neuralcoref>

⁵<https://github.com/huggingface/transformers>

⁶<https://spacy.io/api/sentencizer>

⁷<https://spacy.io/usage/linguistic-features/#noun-chunks>

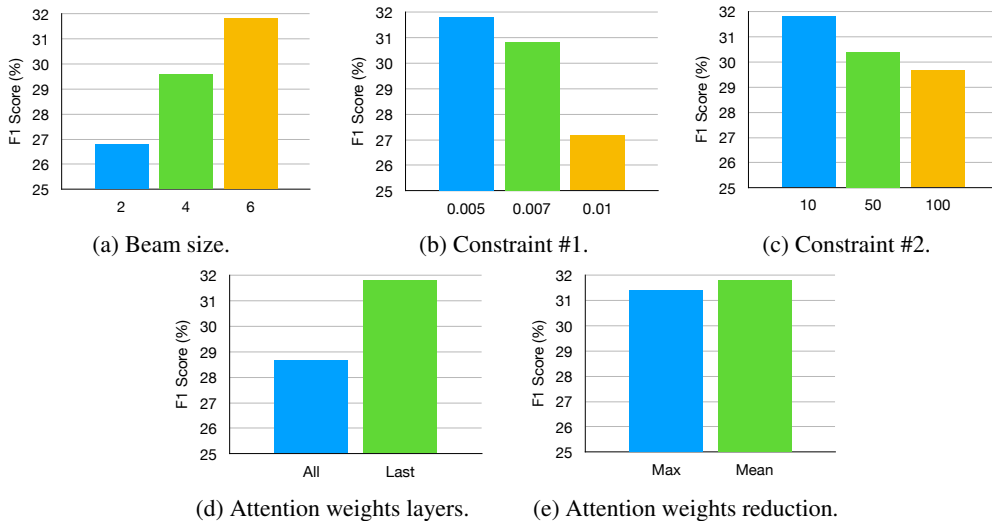


Figure 3: Parameter study with MAMA-BERT_{BASE} on TAC KBP hold-out subset.

such as uninformative relation phrases. We find similar errors are made by MAMA-GPT-2_{XL} on TAC KBP. Similar to Sec. 3.2, enhancing the entity detection, entity linker, relation mapping, and relation generation are helpful. We also plan to leverage lifelong learning (Carlson et al., 2010) to add true facts to the reference KGs to improve the evaluation.

A.3 PARAMETER STUDY

We study the effects of the parameters using MAMA-BERT_{BASE} on TAC KBP. We randomly sample 20% of the oracle query entities as a hold-out dataset to tune the parameters, and use the best parameter setting achieved for both TAC KBP and Wikidata experiments. When studying the effect of a certain parameter, we keep the remaining parameters as default described in Sec. A.1. We use F1 to measure the effects.

Effects of beam size Figure 3(a) illustrates the effects of various beam sizes in Algorithm 1. We find that in general, the larger the beam size is, the better F1 the setting achieves. This is because that MAMA is able to reserve more potentially correct facts when more candidates are allowed in the Match stage. However, F1 improvement gradually becomes subtle, while the computation costs increase more significantly. For sake of the efficiency, we do not explore larger beam sizes. We set the beam size as 6.

Effects of search constraints Figure 3(b) compares the effect of different thresholds of the matching degree of Constraint #1 in Sec. 2.1.2. We set the threshold as 0.005 since it achieves the best result. Note that the summed attention score is normalized by the length of the fact to penalize the cumbersome facts. The matching degree threshold is effective, which is mainly because of the knowledge contained in the self-attention matrix. The score in the attention matrix is representing the chance of the facts to be the true facts based on the stored knowledge. Figure 3(c) shows the impact of the number of distinct head-tail pairs in identifying common relations of Constraint #2 in Sec. 2.1.2. The best result is achieved when it equals 10. This shows that while MAMA mostly identifies frequent relations, it is also able to capture some rare relations for the open schema.

Effects of attention weights Figure 3(d) shows the comparison between attention weights of the last layer and the mean of all layers. The attention weights of the last layer perform better. This is due to the attention weights in lower layers are low-level linguistic knowledge according to (Clark et al., 2019; Ramsauer et al., 2020), which are less relevant to the factual knowledge for the KG construction. Figure 3(e) compares the impact of different attention reduction, i.e., mean, max, over the attention heads of the last layer. We find the “mean” perform better. The reason is that the token often intensively attends to several specific tokens in the sequence (Michel et al., 2019), and the “mean” operator is more sensitive to such information.

B SAMPLES FROM MAMA ON TAC KBP

B.1 MAPPED FACTS

We randomly sample 100 documents from TAC KBP corpus, then randomly sample sentences from those documents. The uncurated candidate facts and the corresponding mapped facts of the sampled sentences based on our best methods MAMA-BERT_{LARGE} and MAMA-GPT-2_{XL} are shown in Figure 4 and Figure 5 respectively. We also randomly sample several sentences in which MAMA-BERT_{LARGE} differs from MAMA-GPT-2_{XL} in the resulting facts for comparison, which is illustrated in Figure 6. In each table, “**ID**” represents the document ID of a sampled sentence in TAC KBP corpus. “**Sentence**” indicates the sampled sentence. “**Candidate facts to mapped facts**” column contains the candidate facts (on the left side of “→”) and their corresponding mapped facts (on the right side of “→”).

B.2 UNMAPPED FACTS

We randomly sample 100 documents from TAC KBP corpus. From those documents, we show unmapped facts from the sampled sentences from those documents. We manually check the correctness of the unmapped facts according to Sec. 3.2 and show the correct ones. The original candidate facts with the corresponding unmapped facts of the sampled sentences generated by MAMA-BERT_{LARGE} and MAMA-GPT-2_{XL} are shown in Figure 7 and Figure 8. A further comparison of the unmapped candidate facts is illustrated in Figure 9. In each table, “**ID**” represents the document ID of a sampled sentence in TAC KBP corpus. “**Sentence**” indicates the sampled sentence. “**Candidate facts to unmapped facts**” column contains the candidate facts (on the left side of “→”) and their corresponding unmapped facts (on the right side of “→”).

C SAMPLES FROM MAMA ON WIKIDATA

C.1 MAPPED FACTS

Similar to TAC KBP, we randomly sample 100 documents from the Wikidata corpus (i.e., English Wikipedia), then randomly sample sentences from those documents. The uncurated candidate facts and the corresponding mapped facts of the sampled sentences based on our best methods MAMA-BERT_{LARGE} and MAMA-GPT-2_{XL} are shown in Figure 10 and Figure 11 respectively. We also randomly sample several sentences in which MAMA-BERT_{LARGE} differs from MAMA-GPT-2_{XL} in the resulting facts for comparison, which is illustrated in Figure 12. In each table, “**ID**” represents the Wikipedia page’s title of a sampled sentence. “**Sentence**” indicates the sampled sentence. “**Candidate facts to mapped facts**” column contains the candidate facts (on the left side of “→”) and their corresponding mapped facts (on the right side of “→”).

C.2 UNMAPPED FACTS

Similar to TAC KBP, we randomly sample 100 documents from the Wikidata corpus. From those documents, we show unmapped facts from several sampled sentences from those documents. We manually check the correctness of the unmapped facts according to Sec. 3.2 and show the correct ones. The original candidate facts with the corresponding unmapped facts of the sampled sentences generated by MAMA-BERT_{LARGE} and MAMA-GPT-2_{XL} are shown in Figure 13 and Figure 14. A further comparison of the unmapped candidate facts is illustrated in Figure 15. In each table, “**ID**” represents the Wikipedia page’s title of a sampled sentence. “**Sentence**” indicates the sampled sentence. “**Candidate facts to unmapped facts**” column contains the candidate facts (on the left side of “→”) and their corresponding unmapped facts (on the right side of “→”).

D ADDITIONAL OPEN KG SUBGRAPHS FROM MAMA ON WIKIDATA

We sample several documents from the Wikidata corpus. We visualize the mapped facts and unmapped facts from those documents as examples of subgraphs in the resulting open KGs. We show the snapshots of the subgraphs generated by MAMA-BERT_{LARGE} from Figure 16 to Figure 24. We

ID	Sentence	Candidate facts to mapped facts
SF13.ENG.001	But in a sign of just how disenchanted some Afghans have become with their government, Bardardost, a doctoral scholar who lived in France for two decades, is widely believed to be at least fourth in popularity among 42 candidates in the August elections.	(Bardardost, .. A Doctoral Scholar) → (Ramarzan Bardardost, per:title, scholar)
SF13.ENG.002	Bardardost left Pakistan for France in 1982.	(Bardardost, left, Pakistan) → (Ramaszan Bardardost, per:countries_of_residence, Pakistan)
SF13.ENG.007	Mohammed Taniwaj, head of Al-Azhar University, told a schoolgirl to remove her niqab when he spotted her during a tour of an Al-Azhar affiliated school, the independent Al-Masry al-Youm newspaper reported this week.	(Mohammed Taniwaj, head, Al-Azhar University) → (Mohammed Sayed Taniwaj, per:employee_or_member_of, Al-Azhar University)
SF13.ENG.009	He took office in 2006 by defeating longtime incumbent Anwar Chowdhry from Pakistan, who was later barred for alleged financial corruption.	(Longtime incumbent Anwar Chowdhry, from, Pakistan) → (Anwar Chowdhry, per:origin, Pakistan)
SF13.ENG.010	In addition to his wife, Wendy, Do is survived by son Daniel, grandchildren Julie and Joey, and father Pat.	(Do, is survived by son, Daniel) → (Romnie James Do, per:children, Daniel)
SF13.ENG.011	Marshall is charged with grand larceny and fraud and faces up to 25 years in prison if convicted.	(Marshall, with, Fraud) → (Anthony Marshall, per:charges, fraud)
SF13.ENG.011	Marshall, a Tony Award-winning Broadway producer and former U.S. diplomat, sat stone-cold as the jury forewoman read each verdict aloud, the word "guilty" clearly resonating in the otherwise silent courtroom.	(Marshall, .. producer) → (Anthony Marshall, per:title, producer)
SF13.ENG.011	The blue-blooded scene of one of America's most illustrious families appeared to listen impassively as verdicts finding him guilty on 14 counts of grand larceny, conspiracy and fraud were read to a packed courtroom.	(Him, guilty on, Grand Larceny) → (Anthony Marshall, per:charges, grand larceny)
SF13.ENG.011	But Marshall's son, Philip, told a different story.	(Marshall'S Son, .. Philip) → (Anthony Marshall, per:children, Philip)
SF13.ENG.012	Al-Qaida's American spokesman Adam Gadahn, also known as Azzam the American, called on Muslims in the West on Sunday to carry out more attacks like the deadly shooting at the US base in Fort Hood, Texas.	(Adam Gadahn, .. also known, Azzam) → (Adam Gadahn, per:alternate_names, Azzam)
SF13.ENG.012	Gadahn grew up in California and converted to Islam before he moved to Pakistan in 1998 and attended an al-Qaida training camp six years later, according to media reports.	(Gadahn, in, California) → (Adam Gadahn, per:states_provinces_of_residence, California)
SF13.ENG.012	Gadahn moved to Pakistan in 1998, according to the FBI, and is said to have attended an al-Qaida training camp six years later, according to media reports.	(Gadahn, moved to, Pakistan) → (Adam Gadahn, per:countries_of_residence, Pakistan)
SF13.ENG.014	After the Munich attack, he lived in Lebanon, Jordan and several Eastern European countries, where he had close ties to Communist bloc intelligence agencies.	(He, lived in, Lebanon) → (Mohammed Oudeh, per:countries_of_residence, Lebanon)
SF13.ENG.015	Clifton attended Howard University but left before graduating to pursue poetry.	(Clifton, attended, Howard University) → (Lucille Clifton, per:schools_attended, Howard University)
SF13.ENG.017	"Alexander Haig devoted his career to serving our country, both as a soldier and as a diplomat," Albright said.	(Alexander Haig, as, soldier) → (Alexander Haig, per:title, soldier)
SF13.ENG.017	In 1979, he resigned and retired from the Army.	(He, resigned and, The Army) → (Alexander Haig, per:employee_or_member_of, Army)
SF13.ENG.019	McGregor is survived by his wife, Lori, and four children, daughters Jordan, Taylor and Landis, and a son, Logan.	(McGregor, by his wife, Lori) → (Karl McGregor, per:spouse, Lori)
SF13.ENG.020	"Mike was a first-rate journalist, a valued member of our staff for 25 years and we will miss him," Times Editor Kass Stanton said.	(Mike, was, A First-Rate Journalist) → (Mike Penner, per:title, first-rate journalist)
SF13.ENG.020	Penner is survived by his brother, John, a copy editor at the Times, and his former wife, Times sportswriter Lisa Dillman.	(Penner, by his brother, John) → (Mike Penner, per:siblings, John)
SF13.ENG.025	She was charged with theft in Beaumont, Texas, for allegedly wearing a \$10,000 worth of dental work in 2006.	(She, was charged with, Theft) → (Crystal Taylor, per:charges, theft)
SF13.ENG.025	Michigan Native Nancy Kissel was convicted of murder and sentenced in Hong Kong in September 2005.	(Michigan Native Nancy Kissel, convicted of, Murder) → (Nancy Kissel, per:charges, murder)
SF13.ENG.026	Neal returned to New York and concentrated on stage work.	(Neal, returned, New York) → (Patricia Neal, per:states_provinces_of_residence, New York)
SF13.ENG.026	In 1953, married Rosalind Wiseman, the British writer famed for "Charlie and the Chocolate Factory," "James and the Giant Peach" and other tales for children.	(She, married, Rosalind Wiseman) → (Patricia Neal, per:spouse, Rosalind Wiseman)
SF13.ENG.026	Oscar-winning actress Patricia Neal had died of lung cancer at her home on Martha's Vineyard, Massachusetts, on Sunday.	(Neal, had died of, Lung Cancer) → (Patricia Neal, per:cause_of_death, lung cancer)
SF13.ENG.027	Abd al-Hakim, also known as Ammar al-Hakim, has been promoted for months to take his father's place.	(Abd-Hakim'S Son, .. Ammar Al-Hakim) → (Abdul Aziz Al-Hakim, per:children, Ammar al-Hakim)
SF13.ENG.027	Al-Hakim is the head of the Supreme Iraqi Islamic Council (SIIC), the largest Shiite party in Iraq.	(Al-Hakim, of, Supreme Iraqi Islamic Council) → (Abdul Aziz Al-Hakim, per:employee_or_member_of, Supreme Iraqi Islamic Council)
SF13.ENG.027	His former Shiite partners have gathered again to form their own group, the Iraqi National Alliance (INA), which includes the influential Supreme Iraqi Islamic Council (SIIC) of Ammar al-Hakim, who succeeded his father Abdul Aziz al-Hakim, who died in a hospital in Iran last month after a long battle with cancer.	(Al-Hakim, died in, a Iran) → (Abdul Aziz Al-Hakim, per:country_of_death, Iran)
SF13.ENG.028	"It either have Billy doing the fun, being a stranger, or some hotshot trying to be the next Billy Mays," said the guy who actually is the next Billy Mays, his son Billy Mays III.	(The Next Billy Mays, .. his son, Billy Mays III) → (Billy Mays, per:children, Billy Mays III)
SF13.ENG.029	Fignon continued cycling during and after a stint in the Army, and drew attention in the early 1980s when he managed to keep up with fitness camps a pace with his amateurs and professionals rode together.	(Fignon, during and after a stint in, The Army) → (Laurent Fignon, per:employee_or_member_of, The Army)
SF13.ENG.029	Laurent Patrick Fignon was born in Paris on Aug. 12, 1960.	(Laurent Patrick Fignon, was born in, Paris) → (Laurent Fignon, per:city_of_birth, Paris)
SF13.ENG.030	Anderson became the Tigers' manager in June 1979 and built on a foundation that included Alan Trammell at shortstop, Lou Whitaker at second base and Fred Goetz in the outfield and Jack Morris on the pitching staff.	(Anderson, became, The Tigers' Manager) → (Sparky Anderson, per:title, manager)
SF13.ENG.030	In addition to his wife, Carol, Anderson is survived by his sons, Lee and Albert; his daughter, Shirlee Englebrecht; and many grandchildren.	(Anderson, is survived by his sons, Lee) → (Sparky Anderson, per:children, Lee)
SF13.ENG.031	Blake Edwards, a writer and director who was hailed as a Hollywood master of screwball farces and nude comedies like "Victor/Victoria" and the "Pink Panther" movies, died Wednesday night in Santa Monica, Calif. He was 88.	(Blake Edwards, .. writer) → (Blake Edwards, per:title, writer)
SF13.ENG.032	Hwang, who lives in Seoul under tight police security, has written books and delivered speeches condemning Kim's regime as authoritarian and dictatorial.	(Hwang, .. who lives, Seoul) → (Hwang Jang-Yop, per:city_of_residence, Seoul)
SF13.ENG.035	Kaczynska, who was 67, married Kaczynski in 1978 after meeting him in the northern Polish city of Gdansk, where they were both academics.	(Kaczynska, married, Kaczynski) → (Maria Kaczynska, per:spouse, Kaczynski)
SF13.ENG.036	Upon his release he went into India, where he masterminded the 1973 hijacking of a Royal Nepal Airlines plane known to be carrying hundreds of thousands of dollars in cash to fund his banned Nepali Congress party.	(His Release, he went into, India) → (Girija Prasad Koirala, per:countries_of_residence, India)
SF13.ENG.036	Koirala began his political career as a union organizer and was imprisoned for seven years in 1960 after a failed uprising against the monarchy.	(Koirala, career as, A Union Organizer) → (Girija Prasad Koirala, per:title, union organizer)
SF13.ENG.036	Koirala was born in 1925 in Bihar of India where his father Krishna Prasad Koirala and his family were living in exile.	(Koirala, was born in 1925 in, Bihar) → (Girija Prasad Koirala, per:city_of_birth, Bihar)
SF13.ENG.036	Koirala was born in 1925 in Bihar of India at the time when his father Krishna Prasad Koirala along with his family was exiled by Ram rule.	(Koirala, born in 1925, Bihar) → (Girija Prasad Koirala, per:city_of_birth, Bihar)
SF13.ENG.037	Chabrol's survivors also include his third wife, Aureore Pajot, who acted as his script supervisor on nearly all of his movies from 1968 on and whom he married in 1981; and Pajot's daughter, Cecile Maistre, who was an assistant director on his films and wrote the script with him for "The Gun in the Face" (2007).	(Chabrol'S Survivors, wife, Aureore Pajot) → (Claude Chabrol, per:spouse, Aureore Pajot)
SF13.ENG.038	The joint statement said Cunningham was "an inspiring performer and dancer into his 80s, and a visionary choreographer and dedicated teacher throughout his life, he led quietly and by example," the statement said.	(Cunningham, was, An Inspiring Performer) → (Merce Cunningham, per:title, performer)
SF13.ENG.038	Merce Cunningham, the nonagenarian choreographer, is planning for a world without him.	(Merce Cunningham, .. The Nonagenarian Choreographer) → (Merce Cunningham, per:title, choreographer)
SF13.ENG.039	A court on Monday cleared the widower of British reality television star Jade Goody, who died of cancer last year, of rape.	(Goody, .. Cancer) → (Jade Goody, per:cause_of_death, cancer)
SF13.ENG.041	Don Hewitt, the CBS newsmen who invented the highly popular TV newsmagazine "60 Minutes" and produced it for 36 years, died Wednesday.	(Don Hewitt, .. The Cbs Newsmen) → (Don Hewitt, per:title, newsmen)
SF13.ENG.041	"He was the consummate television newsmen," Don Hewitt, a longtime CBS News executive and creator of the long-running "60 Minutes" news program, told Reuters.	(Don Hewitt, .. executive) → (Don Hewitt, per:title, executive)
SF13.ENG.041	Hewitt was already a highly respected TV newsmen.	(Hewitt, was, A Highly Respected Tv Newsmen) → (Don Hewitt, per:title, TV newsmen)
SF13.ENG.041	Donald Shepard Hewitt was born in New York on Dec. 14, 1923, and grew up in the suburb of New Rochelle.	(Donald Shepard Hewitt, born, New York) → (Don Hewitt, per:state_province_of_birth, New York)
SF13.ENG.043	Eleanor Louise Greenewich was born in Brooklyn on Oct. 23, 1940.	(Eleanor Louise Greenewich, born in, Oct.) → (Ellie Greenewich, per:date_of_birth, Oct. 23)
SF13.ENG.044	A little more than a year after Dume died from bladder cancer, the colorful remnants of his estate have been consigned by his family to Star Galleries in Hudson, N.Y., which will auction them Nov. 20.	(Dume, died, Bladder Cancer) → (Dominic Dume, per:cause_of_death, bladder cancer)
SF13.ENG.048	Charles Gwathmey, an architect known for his influential modernist home designs and famous clients like director Steven Spielberg, his died.	(Charles Gwathmey, .. architect) → (Charles Gwathmey, per:title, architect)
SF13.ENG.049	Besides his wife, Mandelbrojt is survived by two sons, Laurent, of Paris, and Didier, of Newton, Mass., and three grandchildren.	(Mandelbrojt, two sons, Laurent) → (Benoit Mandelbrojt, per:children, Laurent)
SF13.ENG.049	For years, he worked for IBM in New York.	(He, worked, Ibm) → (Benoit Mandelbrojt, per:employee_or_member_of, IBM)
SF13.ENG.049	After several years spent largely at the Centre National de la Recherche Scientifique in Paris, Mandelbrojt was hired by IBM in 1958 to work at the Thomas J. Watson Research Center in Yorktown Heights, N.Y. Although he worked frequently with academic researchers and served as a visiting professor at Harvard and the Massachusetts Institute of Technology, it was not until 1987 that he began to teach at Yale, where he earned tenure in 1999.	(Mandelbrojt, hired by, Ibm) → (Benoit Mandelbrojt, per:employee_or_member_of, IBM)
SF13.ENG.056	"It is an issue for everybody in the state because peanuts are a big part of our economy," said Don Koehler, executive director of the Georgia Peanut Commission.	(The Georgia Peanut Commission, director of, Don Koehler) → (Georgia Peanut Commission, org:top_members_employees, Don Koehler)
SF13.ENG.060	"We'll be meeting with scientists, university and science policy officials to explore practical opportunities for exchange and collaboration," Agre, the AAAS president, was quoted as saying.	(The Aaa President, .. Agre) → (American Association for the Advancement of Science, org:top_members_employees, Peter C. Agre)
SF13.ENG.060	However, Alan Leshner, chief executive officer of the American Association for the Advancement of Science, noted that Nobels are generally given for work that's a decade old or more, and that the U.S. doesn't become competitive.	(The American Association, .. Alan Leshner) → (American Association for the Advancement of Science, org:top_members_employees, Alan Leshner)
SF13.ENG.060	Norman Augustine, the former chairman and chief executive of the Lockheed Martin Corporation, Patricia Goldman, former vice chairman of the National Transportation Safety Board, Mary Good, a former president of the American Association for the Advancement of Science, Roger Martin, dean of the Rotman School of Management, Brian O'Neill, a former president of the Insurance Institute for Highway Safety, and Sheila Widnall, a professor at the Massachusetts Institute of Technology.	(The American Association, of, Mary Good) → (American Association for the Advancement of Science, org:top_members_employees, Mary Good)
SF13.ENG.062	"First of all, they never have enough funding," said Andy Kunz, president of the U.S. High Speed Rail Association, a nonprofit that advocates a national high-speed rail network.	(The U.S. High Speed Rail Association, of, Andy Kunz) → (U.S. High Speed Rail Association, org:top_members_employees, Andy Kunz)
SF13.ENG.064	China's shock at the military campaign in the former Yugoslavia helped prod Beijing into playing a bigger role in U.N. peacekeeping, said Bates Gill, director of the Stockholm International Peace Research Institute and co-author of a recent report on China's peacekeeping activities.	(The Stockholm International Peace Research Institute, .. Bates Gill) → (Stockholm International Peace Research Institute, org:top_members_employees, Bates Gill)
SF13.ENG.064	"Non-state actors, for example, a small group of pirates off the coast of Somalia, Al Qaeda and Taliban who operate across borders and have more and more sophisticated means of violence, are becoming bigger and bigger challenges to the international system," said Bates Gill, director of the Stockholm International Peace Research Institute.	(The Stockholm International Peace Research Institute, .. director of, Bates Gill) → (Stockholm International Peace Research Institute, org:top_members_employees, Bates Gill)
SF13.ENG.068	"It is not surprising that one primarily cosmetic business is trying to blow another under the bus by transferring a tax from rich doctors and their wealthy customers to struggling small businesses," John Overstreet, director of the Indoor Tanning Association, said in a statement Saturday.	(The Indoor Tanning Association, director of, John Overstreet) → (Indoor Tanning Association, org:top_members_employees, John Overstreet)
SF13.ENG.076	The chairman of the Swiss Bankers Association, Patrick Odier, told weekly NZZ am Sonntag that Italy and France have shown interest in deals like ones Switzerland signed this week with Germany and Britain.	(The Swiss Bankers Association, .. Patrick Odier) → (Swiss Bankers Association, org:top_members_employees, Patrick Odier)
SF13.ENG.076	The majority of voters in Switzerland, which manages more than 25 percent of the world's foreign-held private wealth, support banking secrecy, according to a survey published last month by the Swiss Bankers Association in Basel.	(The Swiss Bankers Association, in, Basel) → (Swiss Bankers Association, org:city_of_headquarters, Basel)
SF13.ENG.078	"Americans have a right to know the truth" — Islam is a religion of intolerance and violence," said Richard Thompson, legal director of the Thomas More Law Center in Ann Arbor.	(The Thomas More Law Center, in, Ann Arbor) → (Thomas More Law Center, org:city_of_headquarters, Ann Arbor)
SF13.ENG.082	New solutions may be enacted for these problems, though, said Mary Robinson, CEO of the National Council for Adoption.	(The National Council, of, Mary Robinson) → (National Council for Adoption, org:top_members_employees, Mary Robinson)
SF13.ENG.082	"When you close a country, you end up causing more problems than you prevented," said Chuck Johnson, CEO of the National Council for Adoption.	(The National Council, CEO of, Chuck Johnson) → (National Council for Adoption, org:top_members_employees, Chuck Johnson)
SF13.ENG.084	"This is definitely a Goldilocks problem," said Jason Grunert, president of the Bipartisan Policy Center and an energy adviser to the Obama campaign last year.	(The Bipartisan Policy Center, .. Jason Grunert) → (Bipartisan Policy Center, org:top_members_employees, Jason Grunert)
SF13.ENG.085	"Banks are in strong need for the capital markets (to raise funds)," Li Fuun, a director at the China Banking Regulatory Commission, was quoted as saying at a forum over the weekend.	(The China Banking Regulatory Commission, .. a director, Li Fuun) → (China Banking Regulatory Commission, org:top_members_employees, Li Fuun)
SF13.ENG.089	RIA Novosti and Interfax cite Anatoly Isakin, head of Rosoboronexport, as saying Thursday "nothing is blocking the continuation of military-technical cooperation" with Iran.	(Rosoboronexport, .. head of, Anatoly Isakin) → (Rosoboronexport, org:top_members_employees, Anatoly Isakin)
SF13.ENG.089	Rosoboronexport is the only company in Russia that is allowed to export arms, dual-use products and military-related services.	(Rosoboronexport, is the only company in, Russia) → (Rosoboronexport, org:country_of_headquarters, Russia)
SF13.ENG.091	With his wife, Corneille, Middelhoff invested money in 2000 and 2001 with EICH in funds that were formed to buy five properties from KarstadtQuelle, as Arcandor was then known, and leased back to the department store chain before Middelhoff joined the company, according to Middelhoff's spokesman.	(Arcandor, .. KarstadtQuelle) → (Arcandor, org:alternate_names, KarstadtQuelle)
SF13.ENG.091	Arcandor's small-order subsidiary Quelle is in worse shape however, Klaus-Hubert Georg told a press conference in the western city of Essen, where Arcandor is based.	(Arcandor, .. Essen) → (Arcandor, org:city_of_headquarters, Essen)

Figure 5: Mapped facts: MAMA-GPT-2_{XL} on TAC KBP.

ID	Sentence	BERT _{LARGE}	Candidate facts to mapped facts
			GPT-2 _{XL}
SF13.ENG.005	The men, who live in the Chicago area, were identified as David Coleman Headley, 49, a US citizen, and Tahawur Hussain Rana, 48, Rana, a native of Pakistan and citizen of Canada.	(Rana, a native of, Canada) → (Tahawur Hussain Rana, per:origin, Canada)	-
SF13.ENG.006	Khan died at his music centre near San Francisco in the United States after a prolonged kidney ailment, the news agency said.	(Khan, died at his, The United States) → (Ali Akbar Khan, per:country_of_death, United States)	-
SF13.ENG.006	By his early 20s he was music director of All-India Radio in Lucknow, broadcasting as a solo artist and composing for the radio's orchestra.	(Khan, moved, Bombay) → (Ali Akbar Khan, per:cities_of_residence, Bombay)	(He, of, All-India Radio) → (Ali Akbar Khan, per:employee_or_member_of, All-India Radio)
SF13.ENG.006	Delying his father, Khan moved to Bombay and began writing scores for films, including Chetan Anand's "Aandhyan" (1952), Satyajit Ray's "Devi" (1960) and "Jagan Simha" "s" "Hungry Stones" (1960).	(Khan, moved, Bombay) → (Ali Akbar Khan, per:cities_of_residence, Bombay)	-
SF13.ENG.007	In 1966 he graduated from the faculty of theology at Al-Azhar, the 10th century university that has trained the majority of Sunni Muslim clerics from Africa and Asia.	-	(He, graduated from, Al-Azhar) → (Mohammed Sayed Tawazi, per:schools_attended, Al-Azhar)
SF13.ENG.009	He became AIBA president in 2006 by defeating longtime incumbent Anwar Chowdhry of Pakistan, who was later barred for alleged financial corruption.	-	(Longtime incumbent Anwar Chowdhry, by, Aiba President) → (Anwar Chowdhry, per:employee_or_member_of, AIBA)
SF13.ENG.011	Marshall could get a lengthy prison term if convicted of grand larceny.	-	(Marshall, of, Grand Larceny) → (Anthony Marshall, per:charges, larceny)
SF13.ENG.012	Gadahn, also known as Azzam al-Amrki, was dressed in white robes and wearing a white turban as he called for attacks on what he described as "high-value targets".	(Gadahn, known as, Azzam Al-Amrki) → (Adam Gadahn, per:alternate_names, Azzam al-Amrki)	-
SF13.ENG.012	Gadahn grew up on a farm in California and converted to Islam before moving to Pakistan in 1998 and reportedly attending an al-Qaida training camp.	(Gadahn, to, Pakistan) → (Adam Gadahn, per:countries_of_residence, Pakistan)	-
SF13.ENG.012	He was born in Oregon on Sept. 1, 1978 to a Jewish family.	-	(He, born in, Sept.) → (Adam Gadahn, per:date_of_birth, 1978-09-01)
SF13.ENG.015	Besides her sister, Philip, survivors include a son, Graham; three daughters, Sidney, Gillian and Alexis; and three grandchildren.	(Her Sister, son, Graham) → (Lucille Clifton, per:children, Graham)	-
SF13.ENG.015	In addition to her daughter, of Columbia, other survivors include two daughters, Sidney Clifton of Los Angeles and Gillian Clifton - Monell of Fort Washington, Md. - a son, Graham Clifton of Wichita; a 2-year-old daughter; and three grandchildren.	(Her Daughter, Sidney Clifton) → (Lucille Clifton, per:children, Sidney Clifton)	-
SF13.ENG.017	Haig, 85, died early Saturday in a Baltimore, Maryland, hospital.	(Haig, died early, A Baltimore, Maryland, Hospital) → (Alexander Haig, per:state_or_province_of_death, Maryland)	-
SF13.ENG.019	McGregor was born in Iowa, but grew up in Colorado where his father, Brian, was a coaching legend at Arvada West.	(McGregor, but grew up, Brian) → (Keli McGregor, per:parents, Brian)	-
SF13.ENG.023	Kissel was born in Adrian, Michigan, but her family had also lived in Minneapolis.	-	(Kissel, in, Minneapolis) → (Nancy Kissel, per:cities_of_residence, Minneapolis)
SF13.ENG.023	Kissel, 46, of Adrian, Michigan, was convicted of drugging then bludge her husband Robert to death in a luxury Hong Kong apartment and sentenced to life in prison in September 2005.	-	(Kissel, of, Michigan) → (Nancy Kissel, per:states_or_provinces_of_residence, Michigan)
SF13.ENG.026	Neal had lung cancer and died at her home in Edgartown, Massachusetts on Martha's Vineyard, said longtime friend Bud Alberts of Knoxville.	(Neal, died at her home in, Edgartown) → (Patricia Neal, per:city_of_death, Edgartown)	-
SF13.ENG.026	She grew up in Knoxville, Tenn., where Neal was turned out acting at 10 after seeing an impassioned speech about "demon rum" at the local temperance union speaking contest.	(She, in, Tenn.) → (Patricia Neal, per:states_or_provinces_of_residence, Tennessee)	-
SF13.ENG.027	Al-Hakim, who died Wednesday of lung cancer in Tehran, was a symbol for many of the re-emergence of Iraq's Shiite political majority after decades of oppression under Saddam Hussein's Sunni-led regime.	-	(Al-Hakim, who died, Tehran) → (Abdul Aziz Al-Hakim, per:city_of_death, Tehran)
SF13.ENG.028	TAMPA Heart disease, exacerbated by cocaine use, killed celebrated TV pitchman Billy Mays, according to the Hillsborough County medical examiner's final autopsy report released Friday.	(Celebrated TV Pitchman Billy Mays, killed, Tampa) → (Billy Mays, per:cause_of_death, TAMPA Heart disease)	-
SF13.ENG.030	Family spokesman Dan Ewald said Anderson died from complications from dementia.	(Anderson, died from, Dementia) → (Sparky Anderson, per:cause_of_death, dementia)	-
SF13.ENG.031	He served in the Coast Guard during World War II and had a rare leading role in the low-budget thriller "Strangler of the Swamp" (1946).	(He, in, The Coast Guard) → (Blake Edwards, per:employee_or_member_of, Coast Guard)	-
SF13.ENG.031	His agent Lou Pitt confirmed he had died, while the "Entertainment Tonight" news show and website reported that he died in Santa Monica on Wednesday night from complications of pneumonia, with Andrews at his side.	(He, died, Complications) → (Blake Edwards, per:cause_of_death, complications)	-
SF13.ENG.033	In a career that spanned seven decades, Ginzburg authored several groundbreaking studies in various fields - such as quantum theory, astrophysics, radio-astronomy and diffusion of cosmic radiation in the Earth's atmosphere - that won of "Nobel Prize caliber," said Gennady Mezyats, the director of the Lebedev Physics Institute in Moscow, where Ginzburg worked.	-	(Ginzburg, in, The Lebedev Physics Institute) → (Vitaly Ginzburg, per:employee_or_member_of, Lebedev Physics Institute)
SF13.ENG.035	Kaczynska, who died aged 59, married Kaczynski in 1978.	(Kaczynska, married, Kaczynski) → (Maria Kaczynska, per:spouse, Kaczynski)	-
SF13.ENG.037	Prime Minister Francois Filion called Chabrol a "great director, producer and screenwriter (who was one of the grand figures of the 'Nouvelle vague,' which revolutionized the style and techniques of cinema by looking at real experience, true life, that which is indiscreet and subtle."	(Alfred Chabrol, was, director) → (Claude Chabrol, per:title, film director); (Chabrol, director, Producer) → (Claude Chabrol, per:title, producer)	-
SF13.ENG.038	Cunningham's survivors include a brother, Jack Cunningham of Centralia.	-	(Cunningham's Survivors, brother, Jack Cunningham) → (Merce Cunningham, per:siblings, Jack Cunningham)
SF13.ENG.038	Mercier Philip Cunningham was born in Centralia, Wash., on April 16, 1919.	-	(Mercier Philip Cunningham, was born in, April) → (Merce Cunningham, per:date_of_birth, 1919-04-16)
SF13.ENG.041	A year after enrolling in New York University on a track scholarship, Hewitt dropped out.	(Hewitt, New York University) → (Don Hewitt, per:schools_attended, New York University)	-
SF13.ENG.041	By the time Hewitt was 6, his family was living in New Rochelle, N.Y.	-	(Hewitt, in, New Rochelle) → (Don Hewitt, per:cities_of_residence, New Rochelle)
SF13.ENG.043	Greenwich also worked as an arranger and singer, a role that saw her working with artists including Frank Sinatra and Ella Fitzgerald.	(Greenwich, worked as an, Singer) → (Ellie Greenwich, per:title, singer)	-
SF13.ENG.044	Dunne was born in 1925 in Hartford, Connecticut, to a wealthy Roman Catholic family and grew up in some of the same social circles as the Kennedys.	-	(Dunne, was born in 1925 in, Connecticut) → (Dominick Dunne, per:state_or_province_of_birth, Connecticut)
SF13.ENG.044	In 1957, Dunne moved to Los Angeles to work on the CBS show "Playhouse 90."	(Dunne, moved to, Los Angeles) → (Dominick Dunne, per:cities_of_residence, Los Angeles)	-
SF13.ENG.044	Dunne found his greatest prominence as a celebrity journalist while covering the 1995 murder trial of football star and actor O.J. Simpson, who had been accused of killing his ex-wife Nicole Brown Simpson and her friend Ronald Goldman.	-	(Dunne, as, A Celebrity Journalist) → (Dominick Dunne, per:title, celebrity journalist)
SF13.ENG.045	Goldstein died in New York on Aug. 28.	(Goldstein, died, Aug.) → (Adam Goldstein, per:date_of_death, XXXX-08-28)	-
SF13.ENG.046	After acting in movies and television, Meredith receded into a quiet life in Santa Fe, writing, painting, golfing and acting in a stage production of "The Old Couple."	(Meredith, in, Santa Fe) → (Don Meredith, per:cities_of_residence, Santa Fe)	-
SF13.ENG.046	In high school and at Southern Methodist University, where, already known as Dandy Don (a nickname bestowed on him by his brother), Meredith became an all-American.	(Dandy Don, where, Southern Methodist University) → (Don Meredith, per:schools_attended, Southern Methodist University)	-
SF13.ENG.046	He spent much of his life backing away from the nickname Dandy Don, particularly during his secluded later decades in New Mexico.	(He, the nickname, Dandy Don) → (Don Meredith, per:alternate_names, Dandy Don)	-
SF13.ENG.048	The group, known variously as the Five or the Whites (for the color of most of their buildings) or the New York School, consisted of Gwathmey, Michael Graves, Eisenman, John Hejduk and Richard Meier.	-	(Gwathmey, of, The New York School) → (Charles Gwathmey, per:employee_or_member_of, New York School)
SF13.ENG.048	Meier, who said he had known Gwathmey for 50 years, has particularly fond memories from the time when Gwathmey was first courting his second wife, Bette - Ann Damson, and they all picked corn for dinner in a field adjacent to a barn Meier was renting on the East End of Long Island.	(Gwathmey, his second wife, Bette-Ann Damson) → (Charles Gwathmey, per:spouse, Bette-Ann Damson)	-
SF13.ENG.057	By 2013, automakers will have dozens of plug-in electric hybrid vehicles and fully electric vehicles, said Jason Forcier, a vice president at battery maker A123 Systems Inc.	-	(A123 Systems Inc., a vice president, Jason Forcier) → (A123 Systems Inc., org:top_members_employees, Jason Forcier)
SF13.ENG.058	Access Industries, a privately held company founded in 1986 by Len Blavatnik, has a diverse portfolio of investments in industry, real estate, media and telecommunications.	(Access Industries, founded in 1986 by, Len Blavatnik) → (Access Industries, org:founded_by, Len Blavatnik)	-
SF13.ENG.069	The indications are positive, said Vincent Cogliano, director of the Monographs program at IARC, which decides on carcinogen classifications.	-	(Iarc, director of, the Vincent Cogliano) → (International Agency for Research on Cancer, org:top_members_employees, Vincent Cogliano)
SF13.ENG.082	"There hasn't been a concerted push to open doors for Muslim orphans because the expectation would be that those efforts would fall flat," said Chuck Johnson, chief executive of the National Council for Adoption, a policy group in Alexandria, Va.	-	(Adoption, in, Alexandria) → (National Council for Adoption, org:city_of_headquarters, Alexandria)
SF13.ENG.084	"The scenario itself is secret," said Eileen McMenamin, vice president of communications for the Bipartisan Policy Center (BPC), which is hosting the event dubbed "Cyber ShockWave."	(The Bipartisan Policy Center, vice president of, Eileen McMenamin) → (Bipartisan Policy Center, org:top_members_employees, Eileen McMenamin)	-
SF13.ENG.088	Bank Julius Baer Co., based in Basel, Switzerland, said because WikiLeaks posted account-holder information from its Cayman outpost amid allegations of money laundering and tax evasion.	-	(Bank Julius Baer Co., based, Basel) → (Bank Julius Baer, org:city_of_headquarters, Basel)
SF13.ENG.094	The other manufacturers are two US companies, L3 Communications and Rapiscan Systems, a unit of OSI Systems, and British rival Smiths Detection.	-	(Rapiscan, Two US Companies) → (Rapiscan Systems, org:country_of_headquarters, US)
SF13.ENG.096	Freedom was founded in the 1930s by R.C. Hoiles and is still majority owned by the Hoiles family.	-	(Freedom, was founded in the, R.C. Hoiles) → (Freedom Communications, org:founded_by, R.C. Hoiles)
SF13.ENG.098	Shares of Millipore have more than doubled since Martin D. Madaus took over as chief executive in January 2005.	(Millipore, have, Martin D.) → (Millipore Corp., org:top_members_employees, Martin D. Madaus)	-

Figure 6: Mapped facts: MAMA-BERT_{LARGE} vs. MAMA-GPT-2_{XL} on TAC KBP.

ID	Sentence	Candidate facts to unmapped facts
SF13.ENG.001	Bashardost was born in 1965 in the southern Ghanzi province and his family migrated to Iran and then to Pakistan after successive coup and factional fighting in Afghanistan .	(Bashardost, in, The Southern Ghanzi Province) → (Ramazan Bashardost, per:stateorprovince_of_birth, The Southern Ghanzi Province)
SF13.ENG.012	Gadahn, also known as Azzam the American, was born in 1978 .	(Gadahn, , Azzam) → (Adam Gadahn, per:alternate_names, Azzam)
SF13.ENG.015	Besides her sister, Philip, survivors include a son, Graham, three daughters, Sadney, Gillian and Alexia; and three grandchildren .	(Her Sister, , Philip) → (Lucille Clifton, per:siblings, Philip)
SF13.ENG.015	The poet laureate of Maryland from 1979 to 1985, Clifton was a writer in residence at Coppin State College, a now Coppin State University, a historically black college in Baltimore .	(Clifton, in, Coppin State College) → (Lucille Clifton, per:school_attended, Coppin State College)
SF13.ENG.015	Her father, Samuel, was a steelworker; her mother, Thelma, worked in a laundry .	(Her Father, , Samuel) → (Lucille Clifton, per:parents, Samuel)
SF13.ENG.015	Lucille Clifton, a distinguished American poet whose work trained lenses wide and narrow on the experience of being black and female in the 20th century, exploring vast subjects like the indignities of history and intimate ones like the indignities of the body, died on Saturday in Baltimore .	(Lucille Clifton, , poet) → (Lucille Clifton, per:title, poet)
SF13.ENG.015	The precise cause of death had not been determined, her sister, Elaine Philip, told The Associated Press on Sunday .	(Her Sister, , Elaine Philip) → (Lucille Clifton, per:siblings, Elaine Philip)
SF13.ENG.037	CLAUDE CHABROL, PIONEER FRENCH FILMMAKER, DIES AT 80 Joseph Berger contributed reporting from New York, and Maïa de la Baume contributed from Paris .	(Claude Chabrol, , Pioneer French Filmmaker) → (Claude Chabrol, per:title, Pioneer French Filmmaker)
SF13.ENG.037	Chabrol's survivors also include his third wife, Aurore Pajot, who acted as his script supervisor on nearly all of his movies from 1968 on and whom he married in 1981; and Pajot's daughter, Cecile Maistre, who was an assistant director on his films and wrote the script with him for "The Girl Cut in Two" (2007) .	(His Third Wife, , Aurore Pajot) → (Claude Chabrol, per:spouse, Aurore Pajot)
SF13.ENG.044	Dominick Dunne, a novelist and journalist who chronicled true-crime tales of the rich and infamous, including O. J. Simpson and Claus von Bulow, and in turn became a celebrity in his own right, died of bladder cancer Aug. 26 at his home in New York City .	(Dominick Dunne, , A Novelist) → (Dominick Dunne, per:title, A Novelist)
SF13.ENG.044	Dunne and his wife, Ellen Griffin Dunne, known as Lenny, were married in 1954 .	(His Wife, , Ellen Griffin Dunne) → (Dominick Dunne, per:spouse, Ellen Griffin Dunne)
SF13.ENG.044	Dunne was born in 1925 in Hartford, Connecticut, to a wealthy Roman Catholic family and grew up in some of the same social circles as the Kennedys .	(Dunne, , Connecticut) → (Dominick Dunne, per:stateorprovince_of_birth, Connecticut)
SF13.ENG.049	A professor emeritus at Yale University, Mandelbrot was born in Poland but as a child moved with his family to France where he was educated .	(Mandelbrot, , Yale University) → (Benoit Mandelbrot, per:employee_or_member_of, Yale University)
SF13.ENG.058	Access Industries, a privately held company founded in 1986 by Len Blavatnik, has a diverse portfolio of investments in industry, real estate, media and telecommunications .	(Access Industries, , A Privately Held Company) → (Access Industries, properties, A Privately Held Company)
SF13.ENG.079	Water and its links to development, peace and conflict were key words in the annual sessions, Anders Bertnell, executive director of Stockholm International Water Institute (SIWI), said in his opening address .	(Stockholm International, , Anders Bertnell) → (Stockholm International Water Institute, per:top_members_or_employees, Anders Bertnell)

Figure 7: Unmapped facts: MAMA-BERT_{LARGE} on TAC KBP.

ID	Sentence	Candidate facts to unmapped facts
SF13.ENG.009	He took office in 2006 by defeating longtime incumbent Anwar Chowdhry from Pakistan, who was later barred for alleged financial corruption .	(Longtime Incumbent Anwar Chowdhry, from, Pakistan) → (Anwar Chowdhry, per:countries_of_residence, Pakistan)
SF13.ENG.012	The message also featured several appearances of Adam Gadahn, also known as Azzam al-Amriki, an American who grew up in Riverside County and converted to Islam and joined al-Qaida .	(Adam Gadahn, also, Azzam Al-Amriki) → (Adam Gadahn, per:alternate_names, Azzam Al-Amriki); (Adam Gadahn, , Azzam Al-Amriki) → (Adam Gadahn, , Azzam Al-Amriki)
SF13.ENG.014	Mohammed Oudeh, a former math teacher who became the mastermind of the deadly attack on Israeli athletes at the 1972 Munich Olympics, died Friday in Damascus .	(Mohammed Oudeh, , teacher) → (Mohammed Oudeh, per:title, teacher)
SF13.ENG.014	He settled in the West Bank town of Ramallah, but in 1999, after a trip to Jordan, he was barred by Israel from returning .	(He, in, The West Bank Town) → (Mohammed Oudeh, per:cities_of_residence, The West Bank Town)
SF13.ENG.014	Mohammed Oudeh, a former math teacher who became the mastermind of the deadly attack on Israeli athletes at the 1972 Munich Olympics, died Friday in Damascus .	(Mohammed Oudeh, , A Former Math Teacher) → (Mohammed Oudeh, per:title, A Former Math Teacher)
SF13.ENG.014	In later years, as a growing member of the Palestinian old guard, Oudeh, most commonly known by his guerrilla name, Abu Daoud, showed no remorse for the botched hostage taking and killings of 11 members of the Israeli Olympic team that shook the world .	(Oudeh, , Abu Daoud) → (Mohammed Oudeh, per:alternate_names, Abu Daoud)
SF13.ENG.015	After her husband died in 1984, Clifton taught at the University of California at Santa Cruz before returning to Maryland in 1989 .	(Clifton, of, Santa Cruz) → (Lucille Clifton, per:cities_of_residence, Santa Cruz)
SF13.ENG.015	Lucille Clifton, a National Book Award-winning poet and Pulitzer finalist, has died .	(Lucille Clifton, , poet) → (Lucille Clifton, per:title, poet)
SF13.ENG.015	When she was a girl, Lucille Clifton sat on her mother's lap and listened to her recite poetry .	(Lucille Clifton, , A Girl) → (Lucille Clifton, gender, A Girl)
SF13.ENG.015	In 1967, they moved to Baltimore, and Clifton worked in Washington for the old U.S. Office of Education .	(They, to, Baltimore) → (Lucille Clifton, per:cities_of_residence, Baltimore)
SF13.ENG.015	Clifton's sister, Elaine Philip, said the former poet laureate of Maryland passed away Saturday morning at Johns Hopkins University Hospital in Baltimore .	(Clifton's Sister, , Elaine Philip) → (Lucille Clifton, per:siblings, Elaine Philip)
SF13.ENG.018	Harvey Pekar, the writer of comics whose autobiographical comic book series "American Splendor" chronicled his life as a filing clerk, record collector, freelance jazz critic and one of life's all-around misfits, was found dead July 12 at his home near Cleveland .	(Harvey Pekar, , writer) → (Harvey Pekar, per:title, writer)
SF13.ENG.018	Pekar attended what became Case Western Reserve University, served in the Navy in the late 1950s and worked a series of menial jobs before taking a 30-year job as a filing clerk at a VA hospital in Cleveland .	(Pekar, what became, Case Western Reserve University) → (Harvey Pekar, per:schools_attended, Case Western Reserve University)
SF13.ENG.025	Kissel, 46, of Adrian, Michigan, was convicted of drugging then bashing her husband Robert to death in a luxury Hong Kong apartment and sentenced to life in prison in September 2005 .	(Kissel, , Adrian) → (Nancy Kissel, per:cities_of_residence, Adrian); (Kissel, , Michigan) → (Nancy Kissel, per:stateorprovince_of_residence, Michigan)
SF13.ENG.038	Mercier Philip Cunningham was born in Centralia, Wash., on April 16, 1919 .	(Mercier Philip Cunningham, was born in, Centralia) → (Mercer Cunningham, was born in, Centralia); (Mercier Philip Cunningham, in, Wash.) → (Mercer Cunningham, per:stateorprovince_of_birth, Wash.); (Mercier Philip Cunningham, in, April) → (Mercer Cunningham, per:date_of_birth, 1919-04-16)
SF13.ENG.038	Cunningham's survivors include a brother, Jack Cunningham of Centralia .	(Cunningham's Survivors, , Jack Cunningham) → (Mercer Cunningham, per:siblings, Jack Cunningham)
SF13.ENG.041	By the time Hewitt was 6, his family was living in New Rochelle, N.Y .	(Hewitt, in, New Rochelle) → (Don Hewitt, per:cities_of_residence, New Rochelle)
SF13.ENG.041	He is survived by his third wife, former television news correspondent Marilyn Berger; his sons, Steven and Jeffrey; his daughter, Lisa Casera; his stepdaughter, Jillian Childers Hewitt, whom Hewitt adopted; and three grandchildren .	(Hewitt, , Jillian Childers Hewitt) → (Don Hewitt, per:children, Jillian Childers Hewitt)
SF13.ENG.044	Dunne was born in 1925 in Hartford, Connecticut, to a wealthy Roman Catholic family and grew up in some of the same social circles as the Kennedys .	(Dunne, was born in 1925 in, Connecticut) → (Dominick Dunne, per:stateorprovince_of_birth, Connecticut)
SF13.ENG.044	Dunne was part of a famous family that also included his brother, novelist and screenwriter John Gregory Dunne; his brother's wife, author Jean Dixon; and his son, Griffin .	(His Brother, , Novelist) → (His brother, per:title, Novelist)
SF13.ENG.044	Dunne and his wife, Ellen Griffin Dunne, known as Lenny, were married in 1954 .	(Dunne, and, Ellen Griffin Dunne) → (Dominick Dunne, per:spouse, Ellen Griffin Dunne)
SF13.ENG.047	The women were accompanied by Reyna Luisa Tamayo, the mother of political prisoner Orlando Zapata, who died at age 42 in a hunger strike February 23 to protest prison conditions .	(Political Prisoner Orlando Zapata, , Reyna Luisa Tamayo) → (Orlando Zapata, per:children, Reyna Luisa Tamayo)
SF13.ENG.061	InterContinental Hotels Group, owners of Holiday Inn Worldwide, has been ordered by a court to pay \$25 million in damages to a franchisee for fraud .	(Intercontinental Hotels Group, , Holiday Inn Worldwide) → (InterContinental Hotels Group, org:parents, Holiday Inn Worldwide)
SF13.ENG.079	Water and its links to development, peace and conflict were key words in the annual sessions, Anders Bertnell, executive director of Stockholm International Water Institute (SIWI), said in his opening address .	(Stockholm International, director, Anders Bertnell) → (Stockholm International Water Institute, per:top_members_or_employees, Anders Bertnell); (Stockholm International Water Institute, per:top_members_or_employees, Anders Bertnell) → (Stockholm International Water Institute, per:top_members_or_employees, Anders Bertnell)
SF13.ENG.090	ECO member countries include Afghanistan, Azerbaijan, Iran, Kazakhstan, Kyrgyzstan, Pakistan, Tajikistan, Turkey, Turkmenistan and Uzbekistan .	(Eco Member Countries, , Azerbaijan) → (Economic Cooperation Organization, org:members, Azerbaijan)
SF13.ENG.096	A phone message left Sunday at Freedom's Irvine, California, headquarters was not immediately returned .	(Freedom's Irvine, , California) → (Freedom Communications, org:stateorprovince_of_headquarters, California)
SF13.ENG.097	STX Finland is part of the international STX Europe Group, with shipyards in Brazil, Norway, France, Romania and Vietnam .	(STX Finland, of, The International Six Europe Group) → (STX Finland, org:subsidiaries, The International Six Europe Group)

Figure 8: Unmapped facts: MAMA-GPT-2_{XL} on TAC KBP.

ID	Sentence	BERT _{LARGE}	Candidate facts to mapped facts GPT-2XL
SF13.ENG.001	Bashardost was born in 1965 in the southern Ghazni province and his family migrated to Iran and then to Pakistan after successive coup and factional fighting in Afghanistan.	(Bashardost, in, The Southern Ghazni Province) → (Ramazan Bashardost, per:stateorprovince,of, birth, The Southern Ghazni Province)	-
SF13.ENG.009	He took office in 2006 by defeating longtime incumbent Anwar Chowdhry from Pakistan, who was later barred for alleged financial corruption.	(Longtime Incumbent Anwar Chowdhry, from, Pakistan) → (Anwar Chowdhry, per:origin, Pakistan)	(Longtime Incumbent Anwar Chowdhry, from, Pakistan) → (Anwar Chowdhry, per:countries,of, residence, Pakistan)
SF13.ENG.012	Gadahn, also known as Azzam the American, was born in 1978.	(Gadahn, .. Azzam) → (Adam Gadahn, per:alternate, names, Azzam)	(Gadahn, also known as, Azzam) → (Adam Gadahn, per:alternate, names, Azzam)
SF13.ENG.012	The message also featured several appearances of Adam Gadahn, also known as Azzam al-Amriki, an American who grew up in Riverside County and converted to Islam and joined al-Qaida.	(Adam Gadahn, also known as, Azzam Al-Amriki) → (Adam Gadahn, per:alternate, names, Azzam Al-Amriki)	(Adam Gadahn, also, Azzam Al-Amriki) → (Adam Gadahn, per:alternate, names, Azzam Al-Amriki)
SF13.ENG.014	Mohammed Oudeh, a former math teacher who became the mastermind of the deadly attack on Israeli athletes at the 1972 Munich Olympics, died Friday in Damascus.	(Mohammed Oudeh, .. teacher) → (Mohammed Oudeh, per:title, teacher)	-
SF13.ENG.014	He settled in the West Bank town of Ramallah, but in 1999, after a trip to Jordan, he was barred by Israel from returning.	(He, in, The West Bank Town) → (Mohammed Oudeh, per:cities,of, residence, The West Bank Town)	(He, settled in, The West Bank Town) → (Mohammed Oudeh, per:cities,of, residence, The West Bank Town)
SF13.ENG.014	Mohammed Oudeh, a former math teacher who became the mastermind of the deadly attack on Israeli athletes at the 1972 Munich Olympics, died Friday in Damascus.	(Mohammed Oudeh, .. A Former Math Teacher) → (Mohammed Oudeh, per:title, A Former Math Teacher)	(Mohammed Oudeh, .. teacher) → (Mohammed Oudeh, per:title, teacher)
SF13.ENG.014	In later years, as a graying member of the Palestinian old guard, Oudeh, most commonly known by his guerrilla name, Abu Daoud, showed no remorse for the botched hostage taking and killings of 11 members of the Israeli Olympic team that shook the world.	(Oudeh, .. Abu Daoud) → (Mohammed Oudeh, per:alternate, names, Abu Daoud)	-
SF13.ENG.015	The precise cause of death had not been determined, her sister, Elaine Philip, told The Associated Press on Sunday.	-	(Her Sister, .. Elaine Philip) → (Lucille Clifton, per:siblings, Elaine Philip)
SF13.ENG.015	After her husband died in 1984, Clifton taught at the University of California at Santa Cruz before returning to Maryland in 1989.	(Clifton, of, Santa Cruz) → (Lucille Clifton, per:cities,of, residence, Santa Cruz)	(Clifton, taught at, the University) → (Lucille Clifton, per:employer, or, member, of, the University of California)
SF13.ENG.015	Lucille Clifton, a National Book Award-winning poet and Pulitzer finalist, has died.	-	(Lucille Clifton, .. poet) → (Lucille Clifton, per:title, poet)
SF13.ENG.015	When she was a girl, Lucille Clifton sat on her mother's lap and listened to her recite poetry.	-	(Lucille Clifton, .. A Girl) → (Lucille Clifton, gender, A Girl)
SF13.ENG.015	In 1967, they moved to Baltimore, and Clifton worked in Washington for the old U.S. Office of Education.	(Clifton, worked in, old U.S. Office) → (Lucille Clifton, per:employer, or, member, of, U.S. Office of Education)	(They, to, Baltimore) → (Lucille Clifton, per:cities,of, residence, Baltimore)
SF13.ENG.015	Clifton's sister, Elaine Philip, said the former poet laureate of Maryland passed away Saturday morning at Johns Hopkins University Hospital in Baltimore.	-	(Clifton's Sister, .. Elaine Philip) → (Lucille Clifton, per:siblings, Elaine Philip)
SF13.ENG.018	Harvey Pekar, the writer of comics whose autobiographical comic book series "American Splendor" chronicled his life as a filing clerk, record collector, free-lance jazz critic and one of life's all-around misfits, was found dead July 12 at his home near Cleveland.	-	(Harvey Pekar, .. writer) → (Harvey Pekar, per:title, writer)
SF13.ENG.018	Pekar attended what became Case Western Reserve University, served in the Navy in the late 1950s and worked a series of menial jobs before taking a 30-year job as a filing clerk at a VA hospital in Cleveland.	-	(Pekar, what, became, Case Western Reserve University) → (Harvey Pekar, per:schools, attended, Case Western Reserve University)
SF13.ENG.025	Kissel, 46, of Adrian, Michigan, was convicted of drugging then bashing her husband Robert to death in a luxury Hong Kong apartment and sentenced to life in prison in September 2005.	-	(Kissel, .. Adrian) → (Nancy Kissel, per:cities,of, residence, Adrian); (Kissel, .. Michigan) → (Nancy Kissel, per:stateorprovinces,of, residence, Michigan)
SF13.ENG.037	CLAUDE CHABROL, PIONEER FRENCH FILMMAKER, DIES AT 80; Berger contributed reporting from New York, and Misa de la Baume contributed from Paris.	(Claude Chabrol, .. Pioneer French Filmmaker) → (Claude Chabrol, per:title, Pioneer French Filmmaker)	-
SF13.ENG.037	Chabrol's survivors also include his third wife, Aureore Pajot, who acted as his script editor on nearly all of his movies from 1960 on and whom he married in 1981; and Pajot's daughter, Cecile Maistre, who was an assistant director on his films and wrote the script with him for "The Girl Cut in Two" (2007).	(His Third Wife, .. Aureore Pajot) → (Claude Chabrol, per:spouse, Aureore Pajot)	(Chabrol's survivors, third wife, Aureore Pajot) → (Claude Chabrol, per:spouse, Aureore Pajot)
SF13.ENG.038	Mercier Philip Cunningham was born in Centralia, Wash., on April 16, 1919.	-	(Mercier Philip Cunningham, was born in, Centralia) → (Mercie Cunningham, was born in, Centralia); (Mercier Philip Cunningham, in, Wash.) → (Mercie Cunningham, per:stateorprovince,of, birth, Wash.); (Mercier Philip Cunningham, in, April) → (Mercie Cunningham, per:date,of, birth, 1919-04-16)
SF13.ENG.038	Cunningham's survivors include a brother, Jack Cunningham of Centralia.	-	(Cunningham's Survivors) → (Jack Cunningham) → (Mercie Cunningham, per:siblings, Jack Cunningham)
SF13.ENG.041	By the time Hewitt was 6, his family was living in New Rochelle, N.Y.	-	(Hewitt, in, New Rochelle) → (Dan Hewitt, per:cities,of, residence, New Rochelle)
SF13.ENG.041	He is survived by his third wife, former television news correspondent Marilyn Berger; his sons, Steven and Jeffrey; his daughter, Lisa Cassara; his stepdaughter, Julian Childers Hewitt, whom Hewitt adopted; and three grandchildren.	-	(Hewitt, .. Julian Childers Hewitt) → (Dan Hewitt, per:children, Julian Childers Hewitt)
SF13.ENG.044	Dominick Dunne, a novelist and journalist who chronicled true-crime tales of the rich and infamous, including O. J. Simpson and Claus von Bulow, and in turn became a celebrity in his own right, died of bladder cancer Aug. 26 at his home in New York City.	(Dominick Dunne, .. A Novelist) → (Dominick Dunne, per:title, A Novelist)	(Dunne, .. novelist) → (Dominick Dunne, per:title, novelist)
SF13.ENG.044	Dunne and his wife, Ellen Griffin Dunne, known as Lenny, were married in 1954	(His Wife, .. Ellen Griffin Dunne) → (Dominick Dunne, per:spouse, Ellen Griffin Dunne)	(Dunne, and, Ellen Griffin Dunne) → (Dominick Dunne, per:spouse, Ellen Griffin Dunne)
SF13.ENG.044	Dunne was born in 1925 in Hartford, Connecticut, to a wealthy Roman Catholic family and grew up in some of the same social circles as the Kennedys.	(Dunne, was born in 1925 in, Connecticut) → (Dominick Dunne, per:stateorprovince,of, birth, Connecticut)	(Dunne, was born in 1925 in, Connecticut) → (Dominick Dunne, per:stateorprovince,of, birth, Connecticut)
SF13.ENG.044	Dunne was part of a famous family that also included his brother, novelist and screenwriter John Gregory Dunne; his brother's wife, author Joan Didion; and his son, Griffin.	-	(His Brother, .. Novelist) → (His brother, per:title, Novelist)
SF13.ENG.047	The women were accompanied by Reyna Luisa Tamayo, the mother of political prisoner Orlando Zapata, who died at age 42 in a hunger strike February 23 to protest prison conditions.	(Political Prisoner Orlando Zapata, .. Reyna Luisa Tamayo) → (Orlando Zapata, per:children, Reyna Luisa Tamayo)	(Political Prisoner Orlando Zapata, .. Reyna Luisa Tamayo) → (Orlando Zapata, per:children, Reyna Luisa Tamayo)
SF13.ENG.049	A professor emeritus at Yale University, Mandelbrot was born in Poland but as a child moved with his family to France where he was educated.	(Mandelbrot, .. Yale University) → (Benoit Mandelbrot, per:employer, or, member, of, Yale University)	(Mandelbrot, was born in, Poland) → (Benoit Mandelbrot, per:country,of, birth, Poland)
SF13.ENG.058	Access Industries, a privately held company founded in 1986 by Len Blavatnik, has a diverse portfolio of investments in industry, real estate, media and telecommunications.	(Access Industries, .. A Privately Held Company) → (Access Industries, properties, A Privately Held Company)	(Access Industries, founded in 1986 by, Len Blavatnik) → (Access Industries, org: founded, by, Len Blavatnik)
SF13.ENG.061	InterContinental Hotels Group, owners of Holiday Inn Worldwide, has been ordered by a court to pay \$25 million in damages to a franchisee for fraud.	(Intercontinental Hotels Group, .. Holiday Inn Worldwide) → (InterContinental Hotels Group, org: parents, Holiday Inn Worldwide)	(Intercontinental Hotels Group, .. Holiday Inn Worldwide) → (InterContinental Hotels Group, org: parents, Holiday Inn Worldwide)
SF13.ENG.079	Water and its links to development, peace and conflict were key words in the annual sessions, Anders Bertzell, executive director of Stockholm International Water Institute (SIWI), said in his opening address.	(Stockholm International, director, Anders Bertzell) → (Stockholm International Water Institute, per:top, members, or, employees, Anders Bertzell)	(Stockholm International, executive director, Anders Bertzell) → (Stockholm International Water Institute, per:top, members, or, employees, Anders Bertzell)
SF13.ENG.079	Water and its links to development, peace and conflict were key words in the annual sessions, Anders Bertzell, executive director of Stockholm International Water Institute (SIWI), said in his opening address.	(Stockholm International, .. executive director, Anders Bertzell) → (Stockholm International Water Institute, per:top, members, or, employees, Anders Bertzell)	(Stockholm International, executive director, Anders Bertzell) → (Stockholm International Water Institute, per:top, members, or, employees, Anders Bertzell)
SF13.ENG.060	ECO member countries include Afghanistan, Azerbaijan, Iran, Kazakhstan, Kyrgyzstan, Pakistan, Tajikistan, Turkey, Turkmenistan and Uzbekistan.	(Eco Member Countries, include, Azerbaijan) → (Economic Cooperation Organization, org: members, Azerbaijan)	(Eco Member Countries, .. Azerbaijan) → (Economic Cooperation Organization, org: members, Azerbaijan)
SF13.ENG.066	Freedom was founded in the 1930s by R.C. Hoiles and is still majority owned by the Hoiles family.	(Freedom, was founded in, The 1930s) → (Freedom Communications, org: date, founded, 1930-XX-XX)	(Freedom, was founded in, The 1930s) → (Freedom Communications, org: date, founded, 1930-XX-XX)
SF13.ENG.066	A phone message left Sunday at Freedom's Irvine, California, headquarters was not immediately returned.	-	(Freedom's Irvine, .. California) → (Freedom Communications, org: stateorprovince,of, headquarters, California)
SF13.ENG.097	STX Finland is part of the international STX Europe Group, with shipyards in Brazil, Norway, France, Romania and Vietnam.	(Stx Finland, of, The International Stx Europe Group) → (STX Finland, org: subsidiaries, The International Stx Europe Group)	(Stx Finland, of, The International Stx Europe Group) → (STX Finland, org: subsidiaries, The International Stx Europe Group)

Figure 9: Unmapped facts: MAMA-BERT_{LARGE} vs. MAMA-GPT-2XL on TAC KBP.

ID	Sentence	BERT_LABEL	Candidate facts to mapped facts
Andy_Hilbert	In the same season he was claimed off waivers by the Pittsburgh Penguins on March 9, 2006.	-	(He, by, The Pittsburgh Penguins) → (andy_hilbert_Q526158, member_of_sports_team_P54, pittsburgh_penguins_Q19642)
Bob_Dylan_Q392	Dylan married his backup singer Carolyn Dennis on June 4, 1986.	-	(Dylan, married, Carolyn Dennis) → (Bob_Dylan_Q392, spouse_P26, Carolyn_Dennis_Q5045345)
Bones_Hillman	During the hiatus of Midnight Oil, Hillman returned to New Zealand, working as a studio and live musician with Dave Dobbyn and recorded the album "Available Light".	(Hillman, returned, New Zealand) → (bones_hillman_Q510206, country_of_citizenship_P27, new_zealand_Q664)	
Brent_Hinds	Hinds continues to concentrate on Mastodon, with the majority of his time spent touring or in the studio.	-	(His time, with, Mastodon) → (brent_hinds_Q099746, member_of_P463, mastodon_band_Q548844)
Butch_Vig	Their debut album, "Garbage", was an unexpected smash, selling over 4 million copies and certified double platinum in the UK, United States and Australia.	-	(Over 4 Million Copies, in, United States) → (butch_vig_Q451084, country_of_citizenship_P27, united_states_Q30)
C_Kan	In addition to being known as a rapper, C-Kan is also known for supporting the free use of marijuana in Mexico.	-	(Marijuana, in, Mexico) → (c-kan_Q27734073, country_of_citizenship_P27, mexico_Q90)
Charles_Gratiot_Sr	Gratiot died of a stroke in St. Louis.	(Gratiot, died, St. Louis) → (charles_gratiot_sr_Q2595257, place_of_death_P30, st_louis_Q38022)	
Cheryl_Chan	Cheryl Chan Cheryl Chan Wei Ling, PBM, is a Singaporean politician.	(Cheryl Chan Wei Ling, is, A Singaporean Politician) → (cheryl_chan_Q22003605, occupation_P106, politician_Q82955)	
Cyba_Audi	Audi is an experienced moderator and presenter at local, regional, and international business events.	-	(Audi, is, Presenter) → (cyba_audi_Q18921358, occupation_P106, television_presenter_Q947873)
Douglas_Jenat	He is a Fellow of the AAAA, AAAI, and Cognitive Science Society, and an editor of the J. Automated Reasoning, J. Learning Systems, and J. Applied Ontology.	(He, is, a Fellow, Aaaa) → (douglas_jenat_Q559334, member_of_P463, association_for_the_advancement_of_artificial_intelligence_Q2739680)	
Douglas_Jenat	While attending the University of Pennsylvania, Jenat supported himself through programming, notably designing and developing a natural language interface to a U.S. Navy data base question-answering system serving as an early online ship-board operations manual used on US aircraft carriers.	-	(Jenat, —, The University of Pennsylvania) → (douglas_jenat_Q559334, educated_at_P69, university_of_pennsylvania_Q49117)
Fred_Stewart_(Football_Manager)	Despite managing for 39 years he only ever took charge of two clubs, Stockport County and Cardiff City, and he holds the record for longest serving manager in the history of both clubs.	-	(He, charge, of, Stockport County) → (fred_stewart_(football_manager)_Q5496345, member_of_sports_team_P54, stockport_county_f.c._Q18526)
Felix_Tanco	Born Félix Mamiel de Jesús Tanco y Bosmeniel, in Bogotá, Colombia, he arrived in Cuba at a very young age.	(Bogotá, —, Cuba) → (felix_tanco_Q5511649, country_of_citizenship_P27, cuba_Q241)	
Gary_Ahert_Jr	He also won his second Brownlow Medal, becoming the first Gold Coast player to win the award and the 14th player in VFL/AFL history to win it twice.	(He, also, won, His Second Brownlow Medal) → (gary_ahert_jr_Q3098509, award_received_P166, leigh_mathews_trophy_Q6519632; He, also, won, His Second Brownlow Medal) → (gary_ahert_jr_Q3098509, award_received_P166, brownlow_medal_Q3853498)	
Geeto_Mongol	After meeting his would-be tag team partner, Bepo Mongol, Tatrie returned to the WWF as Geeto Mongol and The Mongols were brought to the United States in 1968.	(The Mongols, were, The United States) → (geeto_mongol_Q11324907, country_of_citizenship_P27, united_states_Q30)	
Georgiy_Pickhanov	Georgiy Pickhanov Georgiy Valentinovich Pickhanov was a Russian revolutionary, philosopher and a Marxist theoretician.	(Georgiy Valentinovich Pickhanov, was, A Marxist Theoretician) → (georgiy_pickhanov_Q182905, occupation_P106, philosopher_Q4964182)	
Györfi_Doróty	Györfi Doróty Györfi Doróty was a Hungarian teacher, journalist, and politician.	(Györfi Doróty, was, Politician) → (gyorfi_doroty_Q910001, occupation_P106, politician_Q82955)	
Horace_Rawlins	Rawlins died on 22 January 1935 in a nursing home in Reading, Berkshire.	(Rawlins, died, Reading) → (horace_rawlins_Q4436711, place_of_death_P30, reading_berkshire_Q161491)	
Horace_Rawlins	He played in a professional event at Stannore Golf Club in June 1894 but finished well out of the prizes.	-	(He, played, in, Stannore Golf Club) → (horace_rawlins_Q4436711, sport_P641, golf_Q5377)
Hugo_Winterhalter	Winterhalter remained with RCA Victor until 1963, at which time he moved to Kapp; that same year, he also penned the main title theme for the film, "Diamond Head".	-	(Winterhalter, is, Kapp) → (hugo_winterhalter_Q1635872, record_label_P254, kapp_records_Q1728690)
Isanni_Taliano	He competed in the men's 4 × 100 metres relay at the 1924 Summer Olympics.	(He, competed, in, The 1924 Summer Olympics) → (isanni_taliano_Q3502947, participant_of_P1344, 1924_summer_olympics_Q8132)	
Jean_D'Arcy_France_Duchess_of_Berry	The nun of the Order of the Annunciation of the Blessed Virgin Mary still maintains their way of life in monasteries in France, Belgium, Costa Rica and Poland.	-	(Their Way, in, France) → (jean_darcy_france_duchess_of_berry_Q2362020, country_of_citizenship_P27, france_Q142)
John_Maynard_Keynes	John Maynard Keynes John Maynard Keynes, 1st Baron Keynes, was a British economist, whose ideas fundamentally changed the theory and practice of macroeconomics and the economic policies of governments.	(John Maynard Keynes, was, A British Economist) → (john_maynard_keynes_Q9317, occupation_P106, economist_Q188094)	
Joshaun_Culler	Currently, he is Chair of 1916 Professor of English and Comparative Literature at Cornell University.	(He, is, English) → (joshaun_culler_Q93332, languages_spoken_written_or_signed_P1412, english_language_Q1860)	
Jordi_Catala_Perez	In the regional elections of Catalonia in November 2010, he was chosen in the primary as number three for Ciudadanos' candidacy in the province of Barcelona.	(The Province, of, Barcelona) → (jordi_catala_perez_Q557693, member_of_political_party_P102, citizens_spanish_political_party_Q1393123)	
Jules_Deligny	He competed in the freestyle lightweight event at the 1920 Summer Olympics.	(He, competed, in, The 1920 Summer Olympics) → (jules_deligny_Q16209961, participant_of_P1344, 1920_summer_olympics_Q8128)	
Liz_McGregor	Liz McGregor Liz McGregor is a South African author and a journalist who worked for leading South African newspapers such as the "Sunday Times" and the "Rand Daily Mail".	(Liz McGregor, is, A Journalist) → (liz_mcgregor_Q23541069, occupation_P106, journalist_Q1930187)	
Louise_Shropshire	The granddaughter of slaves, Louise Shropshire was born Louise Jarrett on February 15, 1913 in Coffee County, Alabama.	-	(Louise Shropshire, was, Coffee County, Alabama) → (louise_shropshire_Q15430857, place_of_birth_P19, coffee_county_alabama_Q455660)
Maci_Gürdal	He competed in the men's tournament at the 1952 Summer Olympics.	(He, competed, in, The 1952 Summer Olympics) → (maci_gurdal_Q5713835, participant_of_P1344, 1952_summer_olympics_Q6407)	
Markus_Proll	At the beginning of his career, Proll became a target of some ridicule because his name "Proll" literally means something like "toss" in German.	(Proll, in, German) → (markus_proll_Q756626, country_of_citizenship_P27, germany_Q183)	
Mende_Haabe	Their Annual Four match against the new Hawks score the first season try of his career in a 27 - 10 victory, and the Sharks XV again finished top of the Southern Section log.	(The Sharks, —, The Sharks XV) → (mende_haabe_Q6898971, member_of_sports_team_P54, sharks_currie_cup_Q744636)	
Neville_Southall	He moved on to Everton for £350,000 in 1981 and established himself as the club's first-choice goalkeeper by the 1983-84 season.	-	(He, moved, on, to, Everton) → (neville_southall_Q136650, member_of_sports_team_P54, everton_f.c._Q5784)
Oliga_Semenova_Tyan_Shanskaya	She died in Leningrad.	(She, died, Leningrad) → (oliga_semenova_tyan_shanskaya_Q2392240, place_of_death_P30, saint_petersburg_Q656)	
Parvaty_Ratheesh	Parvaty Rathesh made her film acting debut in 2015 similar to that of her younger brother, Padmaraj Rathesh who also made his acting debut in the same year through "Fireman".	(Her Film, that, of, her younger brother, Padmaraj Rathesh) → (parvaty_ratheesh_Q19895785, sibling_P3373, padmaraj_ratheesh_Q19895782)	
Peter_Dalla_Riva	In 1993, he was inducted into the Canadian Football Hall of Fame.	-	(He, was, inducted, into, the, Fame) → (peter_dalla_riva_Q1713358, award_received_P166, canadian_football_hall_of_fame_Q3517653)
Pierre_Dubem	Unlike many former historians, who denigrated the Middle Ages, he endeavored to show that the Roman Catholic Church had helped foster Western science in one of its most fruitful periods.	(The Roman Catholic Church, had, Foster Western Science) → (pierre_dubem_Q314172, field_of_work_P101, philosophy_of_science_Q59115)	
Rafael_Arcadio_Bernal_Supeano	He then served as bishop of the Roman Catholic Diocese of Aracua, Colombia, from 1990 to 2003 and as bishop of the Roman Catholic Diocese of Libano-Honda, Colombia, from 2003 to 2004.	-	(The Roman Catholic Diocese, of, Colombia) → (rafael_arcadio_bernal_supeano_Q2126904, country_of_citizenship_P27, colombia_Q239)
Rajeshwar_Diyal	The Government of India awarded him the second highest civilian award of the Padma Vibhushan, in 1969.	-	(Him, award, The Padma Vibhushan) → (rajeshwar_diyal_Q15448539, award_received_P166, padma_vibhushan_Q672392)
Ray_Johnson_(American_Football)	Ray Johnson Raymond Robert Johnson was an American football defensive back who played three seasons in the National Football League with the Cleveland Rams and Chicago Cardinals.	(The National Football League, back, who, played, Raymond Robert Johnson) → (ray_johnson_(american_football)_Q102177889, sport_P641, american_football_Q41323)	
Sean_Hamish	Sean Hamish Sean Hamish is an American film writer, producer and director best known for "Saint Judy", "Returns to Zero" and "Sister Cities".	(Sean Hamish, is, An American Film Writer) → (sean_hamish_Q19867622, country_of_citizenship_P27, united_states_Q30; Sean Hamish, is, An American Film Writer) → (sean_hamish_Q19867622, occupation_P106, screenwriter_Q28389)	
Seith_Plum	Born in Edmonton, Plum played professionally for Charlton Athletic, Chelsea and Southend United.	-	(Plum, played, professionally, for, Charlton Athletic) → (seith_plum_Q7456581, member_of_sports_team_P54, charlton_athletic_f.c._Q19462)
Seith_Plum	Plum received his only cap for England at age 23 while playing for Charlton Athletic, starting and playing the full 90 minutes in a 4 - 1 win over France on 10 May 1923.	-	(His Only Cap, playing, for, Charlton Athletic) → (seith_plum_Q7456581, member_of_sports_team_P54, charlton_athletic_f.c._Q19462)
Stanislav_Nedkov	A two-time national champion in freestyle wrestling, he has also had first-place finishes in international tournaments in Russia, Moldova, Turkey, and Bulgaria.	(Turkey, —, and, Bulgaria) → (stanislav_nedkov_Q3844375, country_of_citizenship_P27, bulgaria_Q219)	
Stilyan_Petrov	In addition he is Bulgaria's all-time most-capped player with 105 appearances for the side.	-	(He, is, Bulgaria's All-Time Most Capped Player) → (stilyan_petrov_Q166263, country_of_citizenship_P27, bulgaria_Q219)
Stilyan_Petrov	On 14 January 2010, it was announced that Petrov had come second in Bulgaria's Player of the Year.	(Petrov, in, Bulgaria's Player) → (stilyan_petrov_Q166263, country_of_citizenship_P27, bulgaria_Q219)	
Thomas_A_Dunn	He graduated from DePaul University College of Law in 1971.	-	(He, graduated, from, DePaul University College) → (thomas_a_dunn_Q13082998, educated_at_P69, depaul_university_college_of_law_Q5244034)
Thomas_Scott_(Commentator)	In 1803, Scott left the Lock Hospital to become Rector of Aston Sandford in Buckinghamshire where he remained until his death in 1821.	(His Death, he, remained, until, Aston Sandford) → (thomas_scott_commentator_Q2759831, place_of_death_P30, aston_sandford_Q3080893)	
Thor_Heyerdahl	Thor Heyerdahl Thor Heyerdahl was a Norwegian adventurer and ethnographer with a background in zoology, botany and geography.	-	(Thor Heyerdahl, was, A Norwegian Adventurer) → (thor_heyerdahl_Q133622, country_of_citizenship_P27, norway_Q20)
Venkatesh_Kulkarni	His first novel "Naked in Decan" won the 1984 American Book Award of the Before Columbus Foundation and was listed among the top ten novels of the decade by the "Chicago Tribune".	(His First Novel, —, won, The 1984 American Book Award) → (venkatesh_kulkarni_Q5604876, award_received_P166, american_book_awards_Q463606)	
Vinor_Castro_Dz_Souza	Vinor Castro de Souza Vinor Castro de Souza, or simply Vinor Castro, is a Brazilian singer.	-	(Vinor Castro De Souza, is, A Brazilian Singer) → (vinor_castro_dz_souza_Q5604876, position_played_in_team_Q147, forward_(association_football)_Q280658)
Vyacheslav_Tsuykov	He made his debut in the Soviet Top League in 1990 for FC Dynamo Moscow.	(He, for, FC Dynamo Moscow) → (vyacheslav_tsuykov_Q4503307, member_of_sports_team_P54, fc_dynamo_moscow_Q11497)	
Zare_Markovski	Zare Markovski played in KK Rabotnički and MZT Skopje.	(Zare Markovski, played, in, KK Rabotnički) → (zare_markovski_Q4023877, member_of_sports_team_P54, kl_rabotnicki_Q2601345)	

Figure 12: Mapped facts: MAMA-BERT_{LARGE} vs. MAMA-GPT-2_{XL} on Wikidata.

ID	Sentence	Candidate facts to unmapped facts
Nick_Aldis	He was also a co- presenter of " Britain ' s Strongest Man " on Challenge TV in the United Kingdom .	(Challenge Tv, in, The United Kingdom) → (Challenge Tv, in, united.kingdom.Q145)
Douglas_Bader	At the age of two , Bader joined his parents in India for a year .	(His Parents, in, India) → (douglas.bader.Q348780, in, india.Q668)
Douglas_Bader	He remained in France after the war , where , having attained the rank of major , he died in 1922 of complications from those wounds in a hospital in Saint - Omer , the same area where Bader would bail out and be captured in 1941 .	(He, in, France) → (douglas.bader.Q348780, in, france.Q142)
Douglas_Bader	On 2 July 1941 he was awarded the bar to his DSO .	(He, was awarded the bar, his DSO) → (douglas.bader.Q348780, was awarded the bar, distinguished.service.order.Q615838)
Helen_Storrow	His father , also named James Jackson Storrow , was a prominent attorney , whose clients included Alexander Graham Bell and the government of Venezuela ; his mother , Ann Maria Perry , was the grandchild of naval hero Commodore Oliver Hazard Perry , and a distant cousin of President Thomas Jefferson .	(His Father, was, A Prominent Attorney) → (His Father, occupation.P106, lawyer.Q40348)
Helen_Storrow	Helen Storrow Helen Osborne Storrow was a prominent American philanthropist , early Girl Scout leader , and chair of the World Committee of the World Association of Girl Guides and Girl Scouts for eight years .	(Helen Osborne Storrow, was, A Prominent American Philanthropist) → (helen.storrow.Q5703224, place.of.birth.P19, united.states.Q30)
Jacob_Van_Ruisdael	His cousin Jacob was a registered Memmote in Amsterdam .	(His Cousin, was, A Registered Memmote) → (His Cousin, was, A Registered Memmote)
Jacob_Van_Ruisdael	He appears to have been strongly influenced by other contemporary local Haarlem landscapists , most notably Cornelis Vroom and Allaert van Everdingen .	(He, have been strongly influenced by, landscape painting) → (jacob.van.ruisdael.Q213612, have been strongly influenced by, landscape.art.Q191163)
Liaquat_Ali_Khan	Liaquat Ali Khan was educated at the Aligarh Muslim University in India , and then at Oxford University in the United Kingdom .	(Liaquat Ali Khan, in, India) → (liaquat.ali.khan.Q295713, in, india.Q668)
Pauline_Baynes	Baynes began her education at a convent school .	(Her Education, at, A Convent School) → (pauline.baynes.Q101951, at, catholic.school.Q1138671)
Pauline_Baynes	Her most vivid recollection of their New Year ' s Eve lunch at Magdalen College was of his gleefully picking nuts out of a bowl of Brussels sprouts .	(Their New Year ' s Eve Lunch, at, Magdalen College) → (pauline.baynes.Q101951, at, magdalen.college.oxford.Q81162)
Pauline_Baynes	By the time that she left , she had already formed the ambition of becoming an illustrator .	(She, formed the ambition of becoming, An Illustrator) → (pauline.baynes.Q101951, formed the ambition of becoming, illustrator.Q644687)
Thor_Heyerdahl	He was an atheist .	(He, was, An Atheist) → (thor.heyerdahl.Q133622, was, An Atheist)
Thor_Heyerdahl	He is buried in the garden of the family home in Colla Micheri .	(He, is buried, Colla Micheri) → (thor.heyerdahl.Q133622, is buried, colla.micheri.Q3647771)
Neville_Southall	He moved on to Cheshire County League club Winsford United at the age of 20 .	(He, to, Cheshire County League Club Winsford United) → (neville.southall.Q436650, to, winsford.united.f.c.Q5213266)
Neville_Southall	He later became a player - coach at York City , Khyol , Shewsbury Town and Dagenham & Redbridge .	(He, became, A Player-Coach) → (neville.southall.Q436650, occupation.P106, A Player-Coach)
John_Maynard_Keynes	After the war , Winsion Churchill attempted to check the rise of Keynesian policy - making in the United Kingdom and used rhetoric critical of the mixed economy in his 1945 election campaign .	(Keynesian Policy-Making, in, The United Kingdom) → (john.maynard.keynes.Q9317, in, united.kingdom.P145)
John_Maynard_Keynes	Apart from Great Britain , Keynes ' s attention was primarily focused on the United States .	(Keynes ' s Attention, was, The United States) → (john.maynard.keynes.Q9317, was, united.states.Q30)
John_Maynard_Keynes	John Maynard Keynes John Maynard Keynes , 1st Baron Keynes , was a British economist , whose ideas fundamentally changed the theory and practice of macroeconomics and the economic policies of governments .	(John Maynard Keynes, was, A British Economist) → (john.maynard.keynes.Q9317, was, united.kingdom.P145)
Ernst_Haeckel	Ernst Haeckel Ernst Heinrich Philipp August Haeckel was a German zoologist , naturalist , philosopher , physician , professor , marine biologist , and artist who discovered , described and named thousands of new species , mapped a genealogical tree relating all life forms , and coined many terms in biology , including " ecology " , " phylum " , " phylogeny " , and " Protista . "	(August Haeckel, was, A German Zoologist) → (ernst.haeckel.Q48246, was, germany.Q183)
Ernst_Haeckel	Haeckel was awarded the title of Excellency by Kaiser Wilhelm II in 1907 and the Linnean Society of London ' s prestigious Darwin - Wallace Medal in 1908 .	(Haeckel, was awarded the, Excellency) → (ernst.haeckel.Q48246, award/received.P166, Excellency)
Ernst_Haeckel	He was also a pacifist until the First World War , when he wrote propaganda in favor of the war .	(He, was, A Pacifist) → (ernst.haeckel.Q48246, was, A Pacifist)

Figure 13: Unmapped facts: MAMA-BERT_{LARGE} on Wikidata.

ID	Sentence	Candidate facts to unmapped facts
Douglas_Bader	He left St Edward' s in early 1928, aged 18 .	(He left, St Edward) → (douglas.bader.Q348780, left, st.edward.s.school.oxford.Q128260)
Douglas_Bader	Group Captain Sir Douglas Robert Stewart Bader, was a Royal Air Force flying ace during the Second World War .	(Douglas Bader Group Captain Sir Douglas Robert Stewart Bader, was, A Royal Air Force Flying Ace) → (douglas.bader.Q348780, was, flying.ace.Q222982)
Douglas_Bader	On 2 July 1941 he was awarded the bar to his DSO .	(He, was awarded the bar to, his DSO) → (douglas.bader.Q348780, was awarded the bar to, distinguished.service.order.Q615838)
Helen_Storrow	His father , also named James Jackson Storrow , was a prominent attorney , whose clients included Alexander Graham Bell and the government of Venezuela ; his mother , Ann Maria Perry , was the grandchild of naval hero Commodore Oliver Hazard Perry , and a distant cousin of President Thomas Jefferson .	(His Father, was, A Prominent Attorney) → (His Father, occupation.P106, lawyer.Q40348)
Helen_Storrow	Helen Osborne Storrow was a prominent American philanthropist , early Girl Scout leader , and chair of the World Committee of the World Association of Girl Guides and Girl Scouts for eight years .	(Helen Osborne Storrow, was, A Prominent American Philanthropist) → (helen.storrow.Q5703224, was, united.states.Q30)
Jacob_Van_Ruisdael	Ruisdael lived and worked in Amsterdam for the rest of his life .	(Ruisdael, lived, Amsterdam) → (jacob.van.ruisdael.Q213612, lived, amsterdam.Q727); (Ruisdael, lived and worked in, Amsterdam) → (jacob.van.ruisdael.Q213612, lived and worked in, amsterdam.Q727)
Liaquat_Ali_Khan	Liaquat Ali Khan was educated at the Aligarh Muslim University in India , and then at Oxford University in the United Kingdom .	(Liaquat Ali Khan, was educated at, The Aligarh Muslim University) → (liaquat.ali.khan.Q295713, was educated at, aligarh.muslim.university.Q196544)
Liaquat_Ali_Khan	After the death of his father in 1919, Ali Khan , with British Government awarding the grants and scholarship, went to England, attending Oxford University' s Exeter College to pursue his higher education .	(His Higher Education, pursue, Oxford University ' s Exeter College) → (liaquat.ali.khan.Q295713, pursue, university.of.oxford.Q34433)
Liaquat_Ali_Khan	In 1940, Khan was made the deputy leader of the Muslim League Parliamentary party .	(Khan, was made the deputy leader of, The Muslim League Parliamentary Party) → (liaquat.ali.khan.Q295713, deputy leader, all-india.muslim.league.Q223898)
Pauline_Baynes	When Baynes' s father retired, he left India and returned to England, settling with Baynes' s mother in a house close to Baynes' s own near Farnham in southwest Surrey .	(He, left, India) → (pauline.baynes.Q101951, left, india.Q668)
Pauline_Baynes	Baynes began her education at a convent school .	(Her Education, at, A Convent School) → (pauline.baynes.Q101951, at, catholic.school.Q1138671)
Pauline_Baynes	Her most vivid recollection of their New Year ' s Eve lunch at Magdalen College was of his gleefully picking nuts out of a bowl of Brussels sprouts .	(Their New Year ' s Eve Lunch, at, Magdalen College) → (pauline.baynes.Q101951, at, magdalen.college.oxford.Q81162)
Pauline_Baynes	She was not a diligent student, frittering away her time on coffee and parties; and she left the Slade without a qualification . But she did achieve the distinction, one shared with her sister, of exhibiting at the Royal Academy of Arts in 1939 .	(She, without a qualification, The Slade) → (pauline.baynes.Q101951, without a qualification, slade.school.of.fine.art.Q1399299)
Thor_Heyerdahl	In 1940, a year into World War II, both Baynes sisters joined the Women' s Voluntary Service .	(His Theories, at, The General Public) → (thor.heyerdahl.Q133622, at, general.public)
Thor_Heyerdahl	His theories rarely won any scientific acceptance, whereas Heyerdahl himself rejected all scientific criticism and concentrated on publishing his theories in popular books aimed at the general public .	(His Theories, at, The General Public) → (thor.heyerdahl.Q133622, at, general.public)
Thor_Heyerdahl	He was an atheist .	(He, was, An Atheist) → (thor.heyerdahl.Q133622, was, An Atheist)
Neville_Southall	As a teenager, Southall had unsuccessful trials at Wrexham, Crewe Alexandra and Bolton Wanderers .	(Southall, at, Wrexham) → (neville.southall.Q436650, at, wrexham.f.c.Q18529); (Southall, at, Crewe Alexandra) → (neville.southall.Q436650, at, crewe.alexandra.f.c.Q19587)
John_Maynard_Keynes	John Maynard Keynes , 1st Baron Keynes , was a British economist , whose ideas fundamentally changed the theory and practice of macroeconomics and the economic policies of governments .	(John Maynard Keynes, was, A British Economist) → (john.maynard.keynes.Q9317, was, united.kingdom.P145)
John_Maynard_Keynes	In January 1889 at the age of five and a half, Keynes started at the kindergarten of the Perse School for Girls for five mornings a week .	(Keynes, at, The Perse School) → (john.maynard.keynes.Q9317, at, the.perse.school.Q7756751)
John_Maynard_Keynes	In January 1892, at eight and a half, he started at a day pupil at St Faith' s preparatory school .	(He, at, St Faith ' s Preparatory School) → (john.maynard.keynes.Q9317, at, st.faith.s.school.Q7593056)
John_Maynard_Keynes	Keynes was a lifelong member of the Liberal Party, which until the 1920s had been one of the two main political parties in the United Kingdom, and as late as 1916 had often been the dominant power in government .	(Keynes, was a lifelong member of, The Liberal Party) → (john.maynard.keynes.Q9317, was a lifelong member of, liberal.party.uk.Q622441)
Ernst_Haeckel	Haeckel was awarded the title of Excellency by Kaiser Wilhelm II in 1907 and the Linnean Society of London' s prestigious Darwin - Wallace Medal in 1908 .	(Haeckel, was awarded the, Excellency) → (ernst.haeckel.Q48246, award/received.P166, Excellency)
Ernst_Haeckel	He was also a pacifist until the First World War, when he wrote propaganda in favor of the war .	(He, was, A Pacifist) → (ernst.haeckel.Q48246, was, A Pacifist)
Ernst_Haeckel	He was also a social Darwinist who believed that "survival of the fittest" was a natural law, and that struggle led to improvement of the race .	(He, was, A Social Darwinist) → (ernst.haeckel.Q48246, was, A Social Darwinist)
Ernst_Haeckel	Ernst Haeckel was born on 16 February 1834, in Potsdam .	(Haeckel, was born on 17 February 1834, in, Potsdam) → (ernst.haeckel.Q48246, was born on 17 February 1834, in, potsdam.Q1711)

Figure 14: Unmapped facts: MAMA-GPT-2_{XL} on Wikidata.

ID	Sentence	BERT _{LARGE}	Candidate facts to mapped facts
Nick_Aldas	He was also a co-presenter of "Britain's Strongest Man" on Challenge TV in the United Kingdom.	BERT _{LARGE} (Challenge TV, in, The United Kingdom) → (Challenge TV, in, united_kingdom.Q145)	GPT-2 _{XL}
Douglas_Bader	Group Captain Sir Douglas Robert Stewart Bader, was a Royal Air Force flying ace during the Second World War.	-	(Douglas Bader Group Captain Sir Douglas Robert Stewart Bader, was, A Royal Air Force Flying Ace) → (douglas_bader.Q348780, was, flying_ace.Q222982)
Douglas_Bader	At the age of two, Bader joined his parents in India for a year.	(His Parents, in, India) → (douglas_bader.Q348780, in, india.Q668)	-
Douglas_Bader	On 2 July 1941 he was awarded the bar to his DSO.	(He, was awarded the bar, his DSO) → (douglas_bader.Q348780, was, awarded the bar, distinguished_service_order.Q61838)	(He, was awarded the bar, his DSO) → (douglas_bader.Q348780, was, awarded the bar, distinguished_service_order.Q61838)
Helen_Storrow	His father, also named James Jackson Storrow, was a prominent attorney, whose clients included Alexander Graham Bell and the government of Venezuela; his mother, Ann Maria Perry, was the granddaughter of naval hero Commodore Oliver Hazard Perry, and a distant cousin of President Thomas Jefferson.	(His Father, was, A Prominent Attorney) → (His Father, occupation.P106, lawyer.Q40348)	(His Father, was, A Prominent Attorney) → (His Father, occupation.P106, lawyer.Q40348)
Helen_Storrow	Helen Osborne Storrow was a prominent American philanthropist, early Girl Scout leader, and chair of the World Committee of the World Association of Girl Guides and Girl Scouts for eight years.	(Helen Osborne Storrow, was, A Prominent American Philanthropist) → (helen_storrow.Q5703234, was, united_states.Q30)	(Helen Osborne Storrow, was, A Prominent American Philanthropist) → (helen_storrow.Q5703234, was, united_states.Q30)
Jacob_Van_Ruisdael	Ruisdael lived and worked in Amsterdam for the rest of his life.	-	(Ruisdael, lived, Amsterdam) → (jacob_van_ruisdael.Q213612, lived, amsterdam.Q727) (Ruisdael, lived and worked in, Amsterdam) → (jacob_van_ruisdael.Q213612, lived and worked in, amsterdam.Q727)
Jacob_Van_Ruisdael	His cousin Jacob was a registered Mennonite in Amsterdam.	(His Cousin, was, A Registered Mennonite) → (His Cousin, was, A Registered Mennonite)	-
Jacob_Van_Ruisdael	He appears to have been strongly influenced by other contemporary local Haarlem landscapers, most notably Cornelis Vroom and Allaert van Everdingen.	(He, have been strongly influenced by, landscape painting) → (jacob_van_ruisdael.Q213612, have been strongly influenced by, landscape_art.Q191163)	-
Liaquat_Ali_Khan	Liaquat Ali Khan was educated at the Aligarh Muslim University in India, and then at Oxford University in the United Kingdom.	(Liaquat Ali Khan, in, India) → (liaquat_ali_khan.Q295713, in, india.Q668)	(Liaquat Ali Khan, was, educated at, The Aligarh Muslim University) → (liaquat_ali_khan.Q295713, was, educated at, aligarh_muslim_university.Q196544)
Liaquat_Ali_Khan	After the death of his father in 1913, Ali Khan, with British Government awarding the grants and scholarship, went to England, attending Oxford University's Exeter College to pursue his higher education.	-	(His Higher Education, pursue, Oxford University's Exeter College) → (liaquat_ali_khan.Q295713, pursue, university_of_oxford.Q34433)
Liaquat_Ali_Khan	Liaquat Ali Khan was made the deputy leader of the Muslim League Parliamentary party.	-	(khan, was, made the deputy leader of, The Muslim League Parliamentary party) → (liaquat_ali_khan.Q295713, deputy leader, all_india_muslim_league.Q223898)
Pauline_Baynes	When Baynes's father retired, he left India and returned to England, settling with Baynes's mother in a house close to Baynes's own near Farnham in southwest Surrey.	-	(He, left, India) → (pauline_baynes.Q101951, left, india.Q668)
Pauline_Baynes	Baynes began her education at a convent school.	(Her Education, at, A Convent School) → (pauline_baynes.Q101951, at, catholic_school.Q1138671)	(Her Education, at, A Convent School) → (pauline_baynes.Q101951, at, catholic_school.Q1138671)
Pauline_Baynes	Her most vivid recollection of their New Year's Eve lunch at Magdalen College was of his gleefully picking nuts out of a bowl of Brussels sprouts.	(Their New Year's Eve Lunch, at, Magdalen College) → (pauline_baynes.Q101951, at, magdalen_college_oxford.Q81162)	(Their New Year's Eve Lunch, at, Magdalen College) → (pauline_baynes.Q101951, at, magdalen_college_oxford.Q81162)
Pauline_Baynes	She was not a diligent student, frittering away her time on coffee and parties; and she left the Slade without a qualification, but she did achieve the distinction, one shared with her sister, of exhibiting at the Royal Academy of Arts in 1939. In 1940, a year into World War II, both Baynes sisters joined the Women's Voluntary Service.	-	(She, without a qualification, The Slade) → (pauline_baynes.Q101951, without a qualification, slade_school_of_fine_art.Q1399299)
Pauline_Baynes	By the time that she left, she had already formed the ambition of becoming an illustrator.	(She, formed the ambition of becoming, An Illustrator) → (pauline_baynes.Q101951, formed the ambition of becoming, illustrator.Q644687)	-
Thor_Jeyerudal	His theories rarely won any scientific acceptance, whereas Jeyerudal himself rejected all scientific criticism and concentrated on publishing his theories in popular books aimed at the general public.	-	(His Theories, at, The General Public) → (thor_jeyerudal.Q133622, at, general_public)
Thor_Jeyerudal	He was an atheist.	(He, was, An Atheist) → (thor_jeyerudal.Q133622, was, An Atheist)	(He, was, An Atheist) → (thor_jeyerudal.Q133622, was, An Atheist)
Thor_Jeyerudal	He is buried in the garden of the family home in Colla Micheri.	(He, is buried, Colla Micheri) → (thor_jeyerudal.Q133622, is buried, colla_micheri.Q3647771)	-
Neville_Southall	As a teenager, Southall had unsuccessful trials at Wrexham, Crewe Alexandra and Bolton Wanderers.	-	(Southall, at, Wrexham) → (neville_southall.Q436650, at, wrexham.f.c..Q18529) (Southall, at, Crewe Alexandra) → (neville_southall.Q436650, at, crewe_alexandra.f.c..Q1987)
Neville_Southall	He moved on to Cheshire County League club Winsford United at the age of 20.	(He, in, Cheshire County League Club Winsford United) → (neville_southall.Q436650, in, winsford_amltd.f.c..Q5213266)	-
Neville_Southall	He later became a player-coach at York City, Rhyl, Shrewsbury Town and Dagenham & Redbridge.	(He, became, A Player-Coach) → (neville_southall.Q436650, occupation.P106, A Player-Coach)	-
John_Maynard_Keynes	In January 1889 at the age of five and a half, Keynes started at the kindergarten of the Perse School for Girls for five mornings a week.	-	(Keynes, at, The Perse School) → (john_maynard_keynes.Q9317, at, the_perse_school.Q7756751)
John_Maynard_Keynes	In January 1892, at eight and a half, he started as a day pupil at St Faith's preparatory school.	-	(He, at, St Faith's Preparatory School) → (john_maynard_keynes.Q9317, at, st_faith_s_school.Q793956)
John_Maynard_Keynes	After the war, Winston Churchill attempted to check the rise of Keynesian policy-making in the United Kingdom and used rhetoric critical of the mixed economy in his 1945 election campaign.	(Keynesian Policy-Making, in, The United Kingdom) → (john_maynard_keynes.Q9317, in, united_kingdom.P145)	-
John_Maynard_Keynes	Keynes was a lifelong member of the Liberal Party, which until the 1920s had been one of the two main political parties in the United Kingdom, and as late as 1916 had often been the dominant power in government.	-	(Keynes, was, a lifelong member of, The liberal party) → (john_maynard_keynes.Q9317, was, a lifelong member of, liberal_party_uk.Q623441)
Ernst_Haeckel	Ernst Haeckel Ernst Heinrich Philipp August Haeckel was a German zoologist, naturalist, philosopher, physician, professor, marine biologist, and artist who discovered, described and named thousands of new species, mapped a genealogical tree relating all life forms, and coined many terms in biology, including "ecology", "phylum", "phylogeny", and "Protista".	(August Haeckel, was, A German Zoologist) → (ernst_haeckel.Q48246, was, germany.Q183)	-
Ernst_Haeckel	Ernst Haeckel was born on 17 February 1834, in Potsdam.	-	(Haeckel, was born on 17 February 1834, in, Potsdam) → (ernst_haeckel.Q48246, was born on 17 February 1834, in, potsdam.Q1711)
Ernst_Haeckel	Haeckel was awarded the title of Excellency by Kaiser Wilhelm II in 1907 and the Linnaeus Society of London's prestigious Darwin-Wallace Medal in 1908.	(Haeckel, was awarded the, Excellency) → (ernst_haeckel.Q48246, award_received.P166, Excellency)	(Haeckel, was awarded the, Excellency) → (ernst_haeckel.Q48246, award_received.P166, Excellency)
Ernst_Haeckel	He was also a pacifist until the First World War, when he wrote propaganda in favor of the war.	(He, was, A Pacifist) → (ernst_haeckel.Q48246, was, A Pacifist)	(He, was, A Pacifist) → (ernst_haeckel.Q48246, was, A Pacifist)

Figure 15: Unmapped facts: MAMA-BERT_{LARGE} vs. MAMA-GPT-2_{XL} on Wikidata.

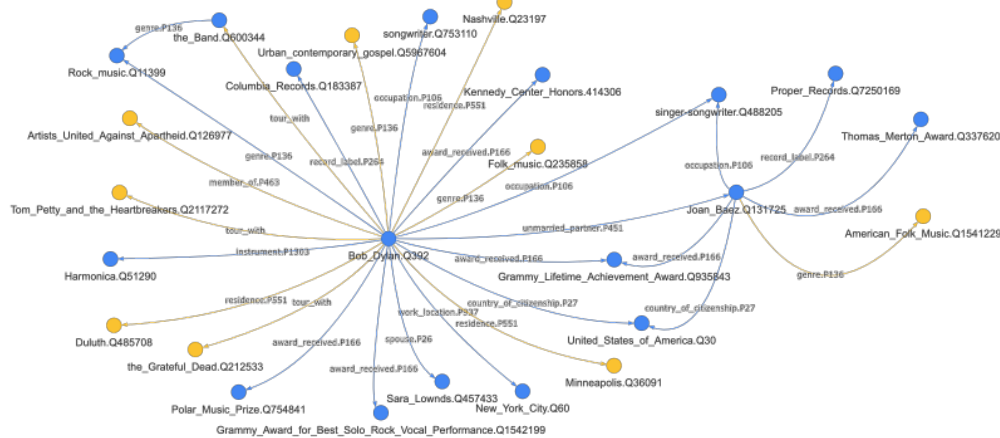


Figure 16: A snapshot subgraph of the open KG generated by MAMA using BERT_{LARGE} from Wikipedia pages neighboring "Bob Dylan". The blue node and arrow represent the mapped facts in the Wikidata schema, while the yellow node and arrow denote the unmapped facts in the open schema. We also visualize the correct facts that are new in Wikidata in yellow.

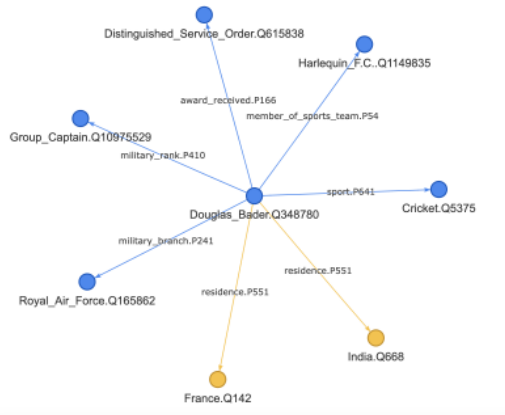


Figure 17: A snapshot subgraph of the open KG generated by MAMA-BERT_{LARGE} from the Wikipedia page “Douglas.Bader”.

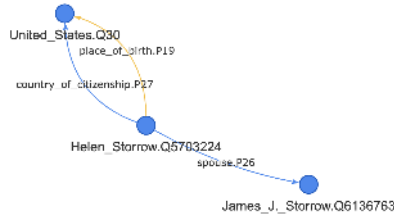


Figure 18: A snapshot subgraph of the open KG generated by MAMA-BERT_{LARGE} from the Wikipedia page “Helen.Storrow”.

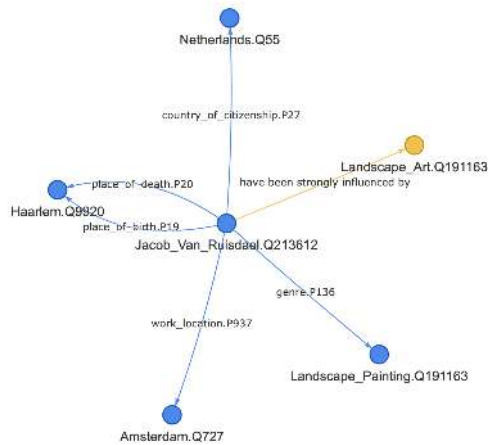


Figure 19: A snapshot subgraph of the open KG generated by MAMA-BERT_{LARGE} from the Wikipedia page “Jacob_van.Ruisdael”.

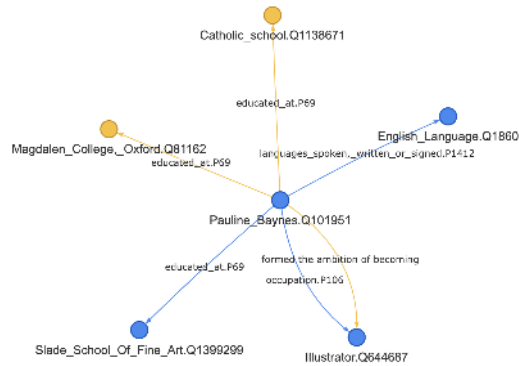


Figure 23: A snapshot subgraph of the open KG generated by MAMA-BERT_{LARGE} from the Wikipedia page “Pauline_Baynes”.

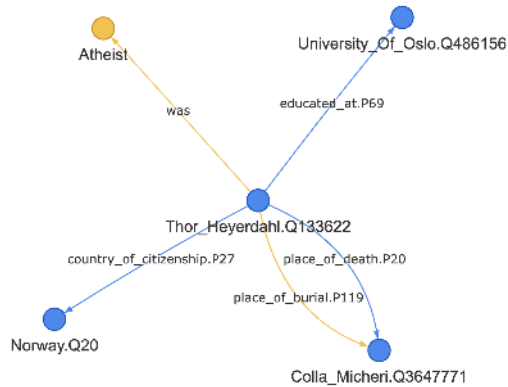


Figure 24: A snapshot subgraph of the open KG generated by MAMA-BERT_{LARGE} from the Wikipedia page “Thor_Heyerdahl”.

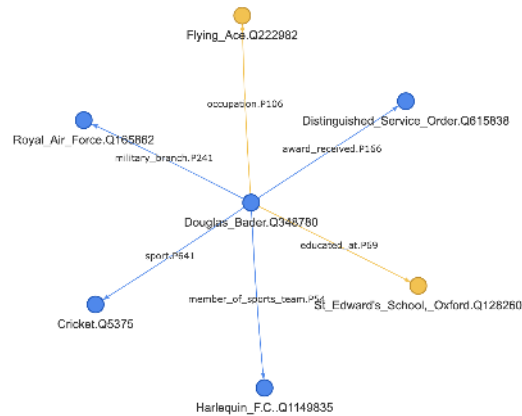


Figure 25: A snapshot subgraph of the open KG generated by MAMA-GPT-2_{XL} from the Wikipedia page “Douglas_Bader”.

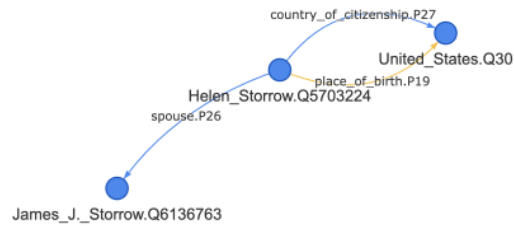


Figure 26: A snapshot subgraph of the open KG generated by MAMA-GPT-2_{XL} from the Wikipedia page “Helen_Storrow”.

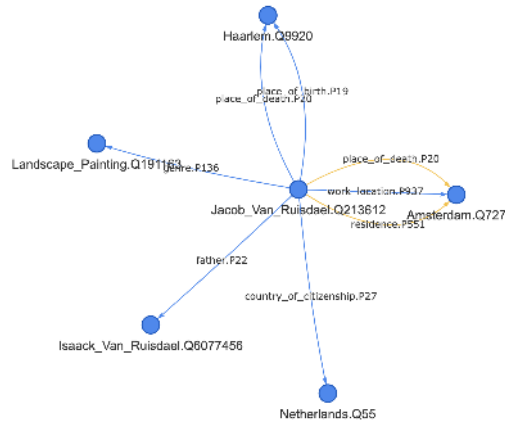


Figure 27: A snapshot subgraph of the open KG generated by MAMA-GPT-2_{XL} from the Wikipedia page “Jacob_van_Ruisdael”.

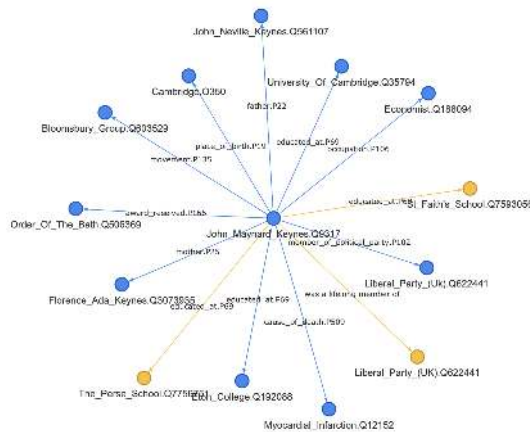


Figure 28: A snapshot subgraph of the open KG generated by MAMA-GPT-2_{XL} from the Wikipedia page “John_Maynard_Keynes”.

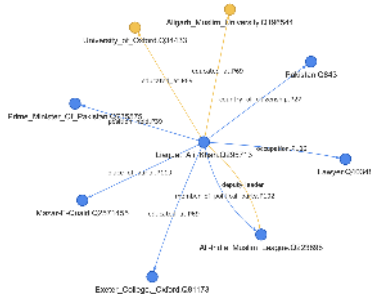


Figure 29: A snapshot subgraph of the open KG generated by MAMA-GPT-2_XL from the Wikipedia page "Liaquat_Ali_Khan".

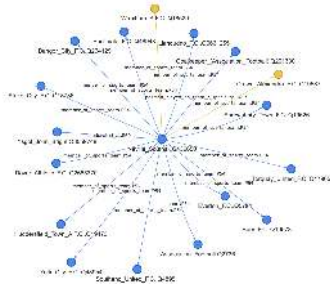


Figure 30: A snapshot subgraph of the open KG generated by MAMA-GPT-2_XL from the Wikipedia page "Neville_Southall".

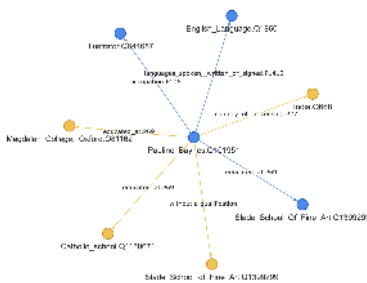


Figure 31: A snapshot subgraph of the open KG generated by MAMA-GPT-2_XL from the Wikipedia page "Pauline_Baynes".

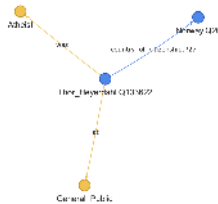


Figure 32: A snapshot subgraph of the open KG generated by MAMA-GPT-2_XL from the Wikipedia page "Thor_Heyerdahl".

Language Models are
Open Knowledge Graphs
Supplementary Materials

Problem

- Knowledge graph construction requires human supervision

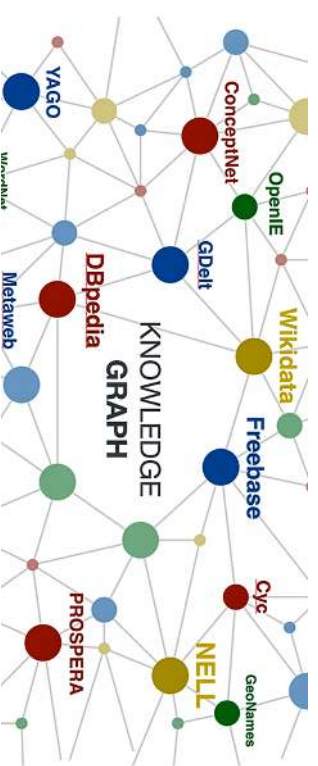
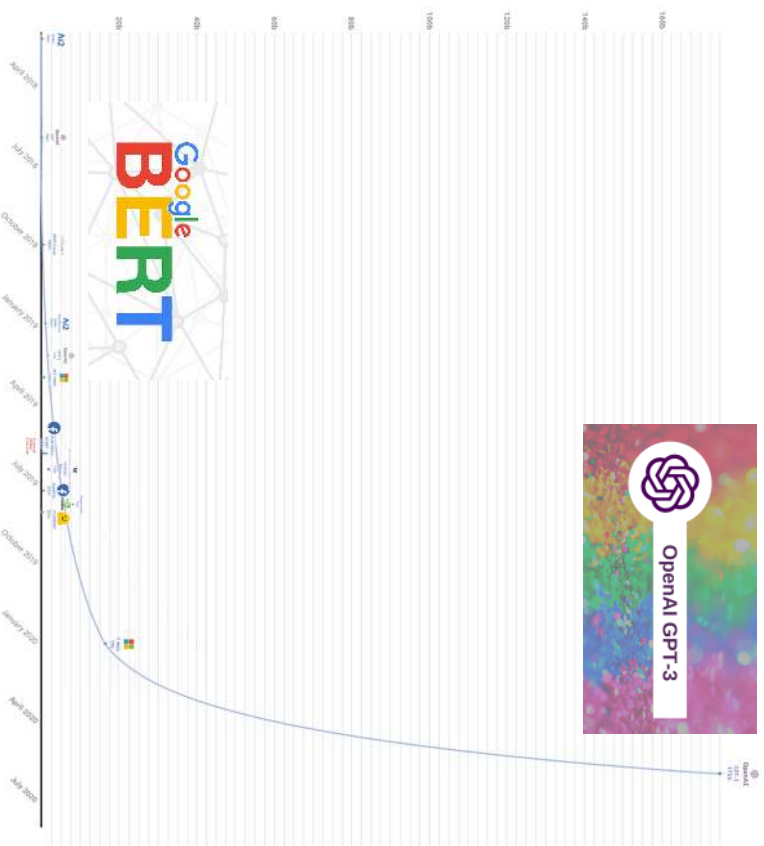


- Language models store knowledge

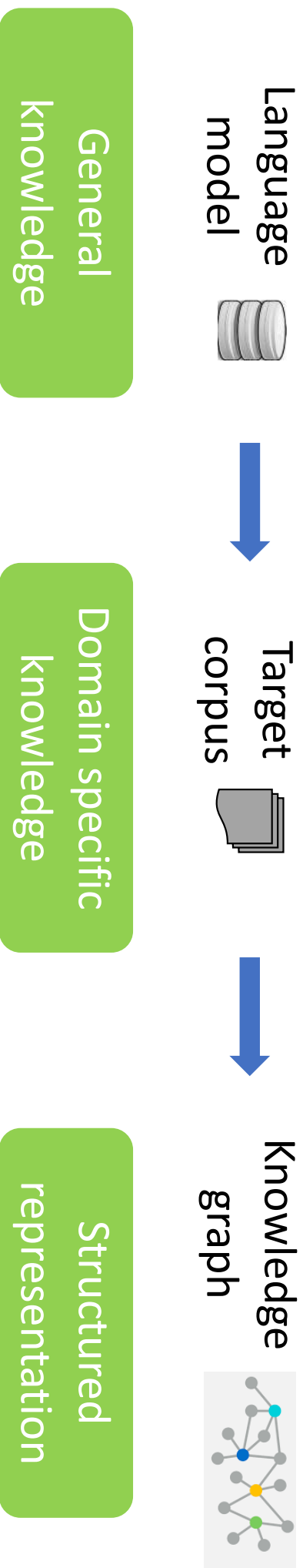


Problem

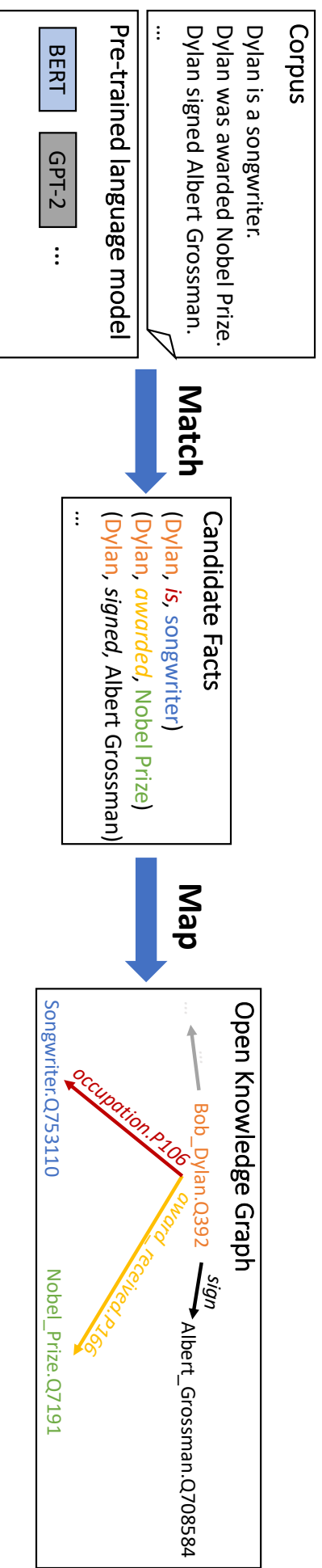
- How to use **language models** to construct **knowledge graphs**?



Challenges

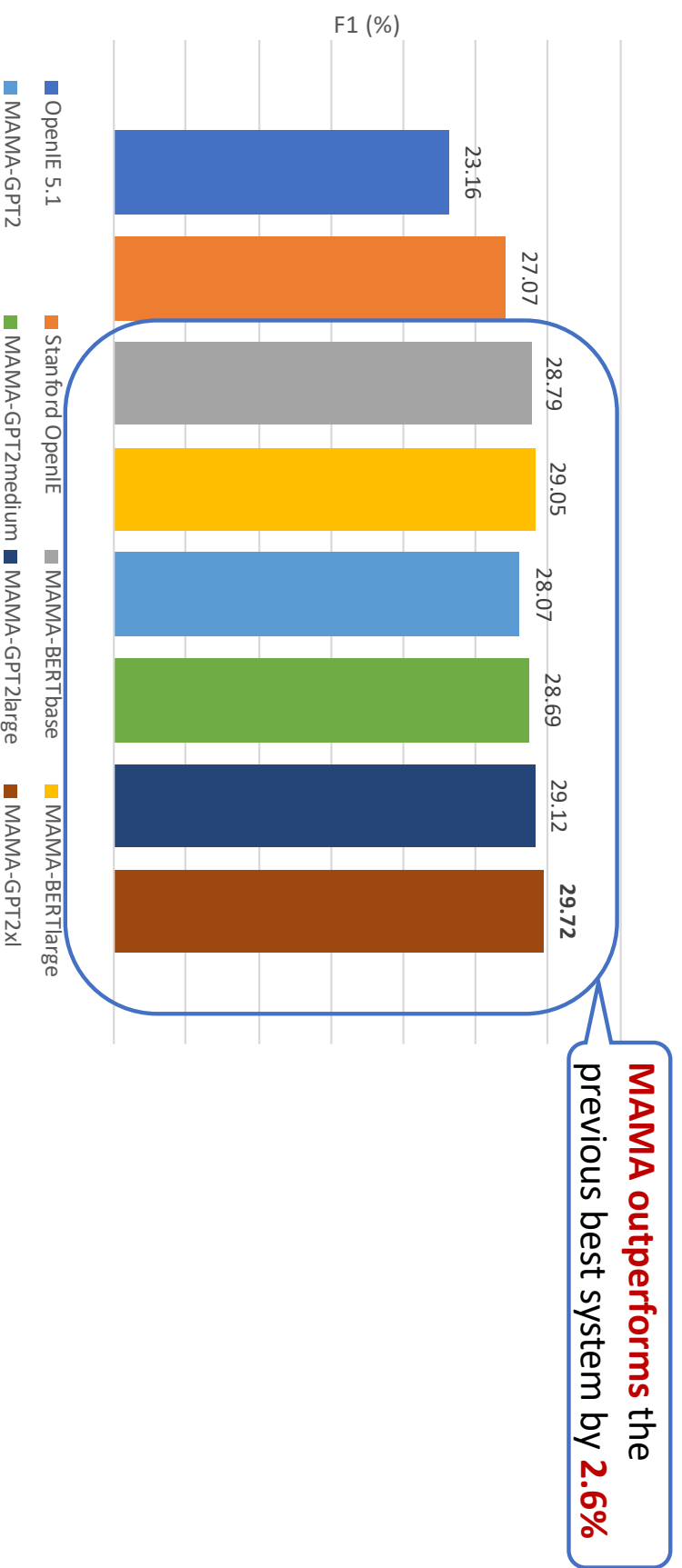


Proposed Approach

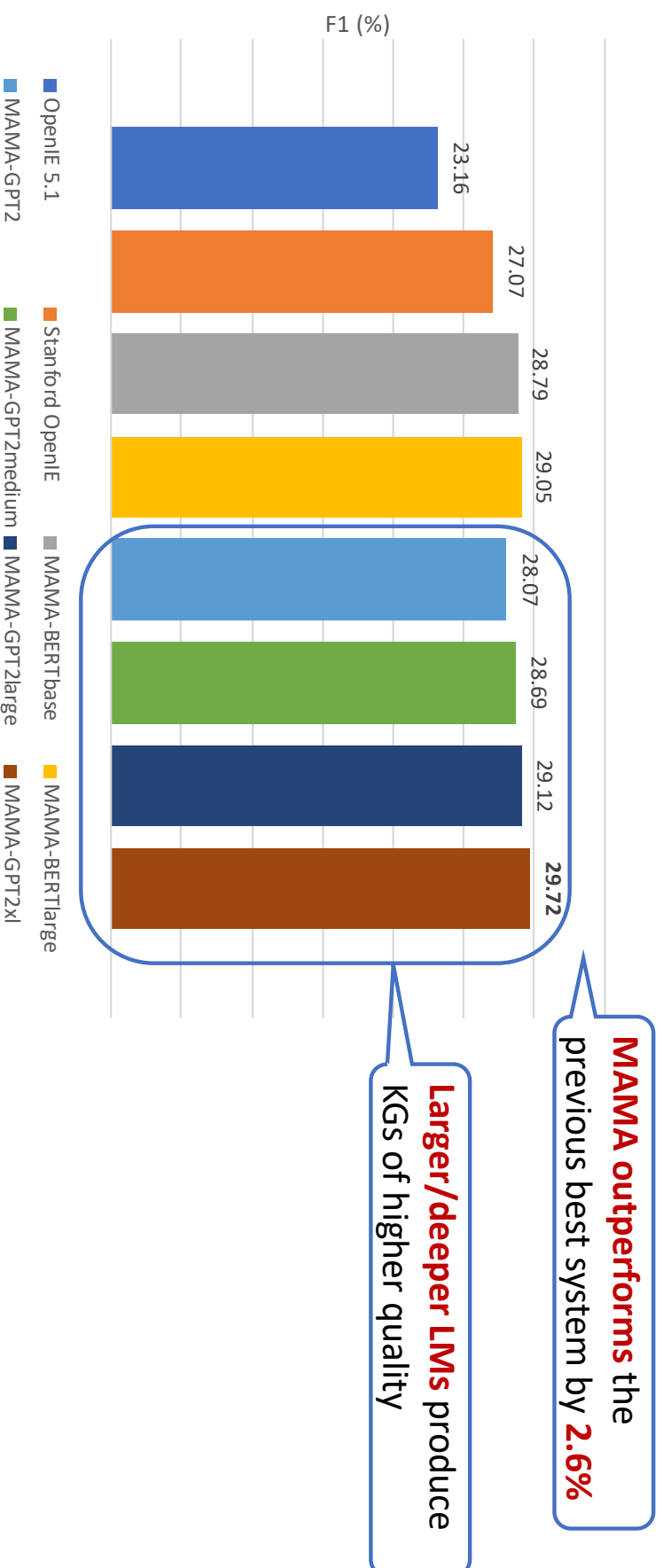


- MAMA constructs an open knowledge graph with **a single forward pass** of the language model (without fine-tuning) over the corpus
 - **Match:** generates a set of candidate facts from a textual corpus
 - **Map:** produces an open knowledge graph from the matched candidates

Results on TAC KBP

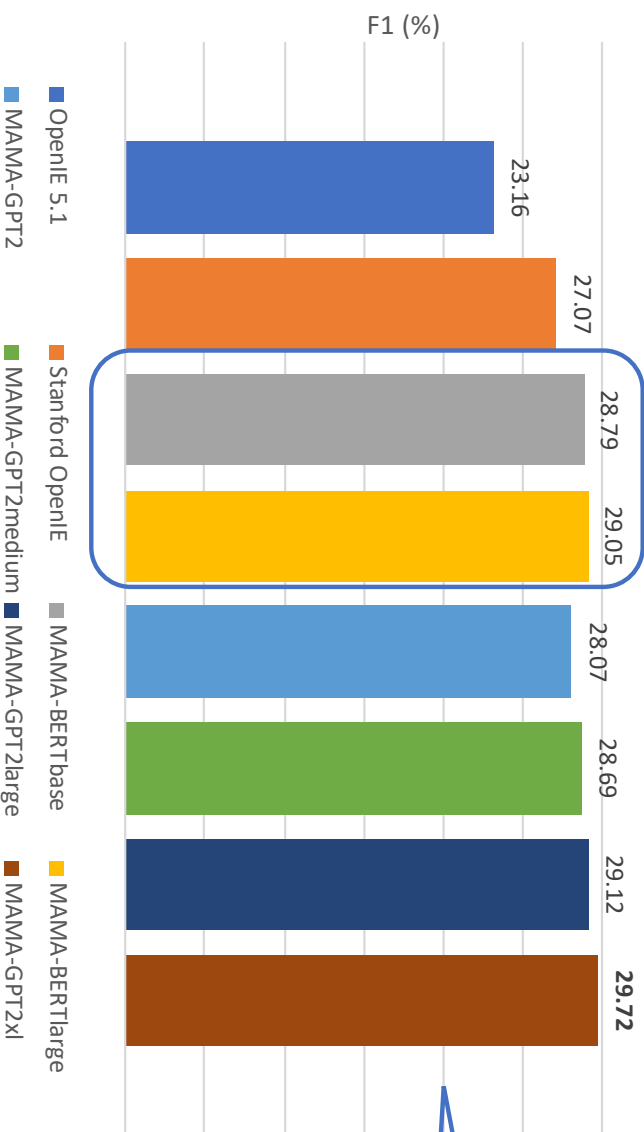


Results on TAC KBP



Results on TAC KBP

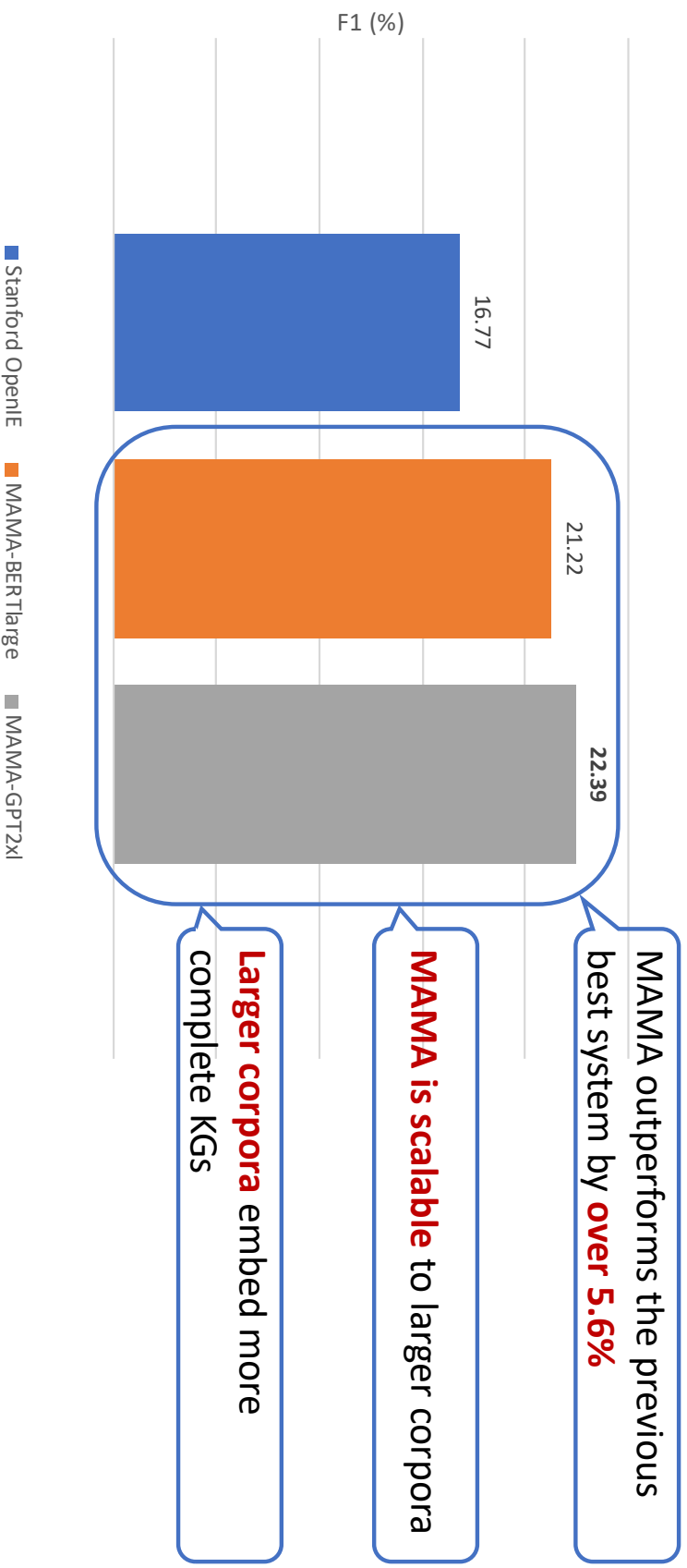
BERT LMs outperform GPT-2 LMs under similar model sizes



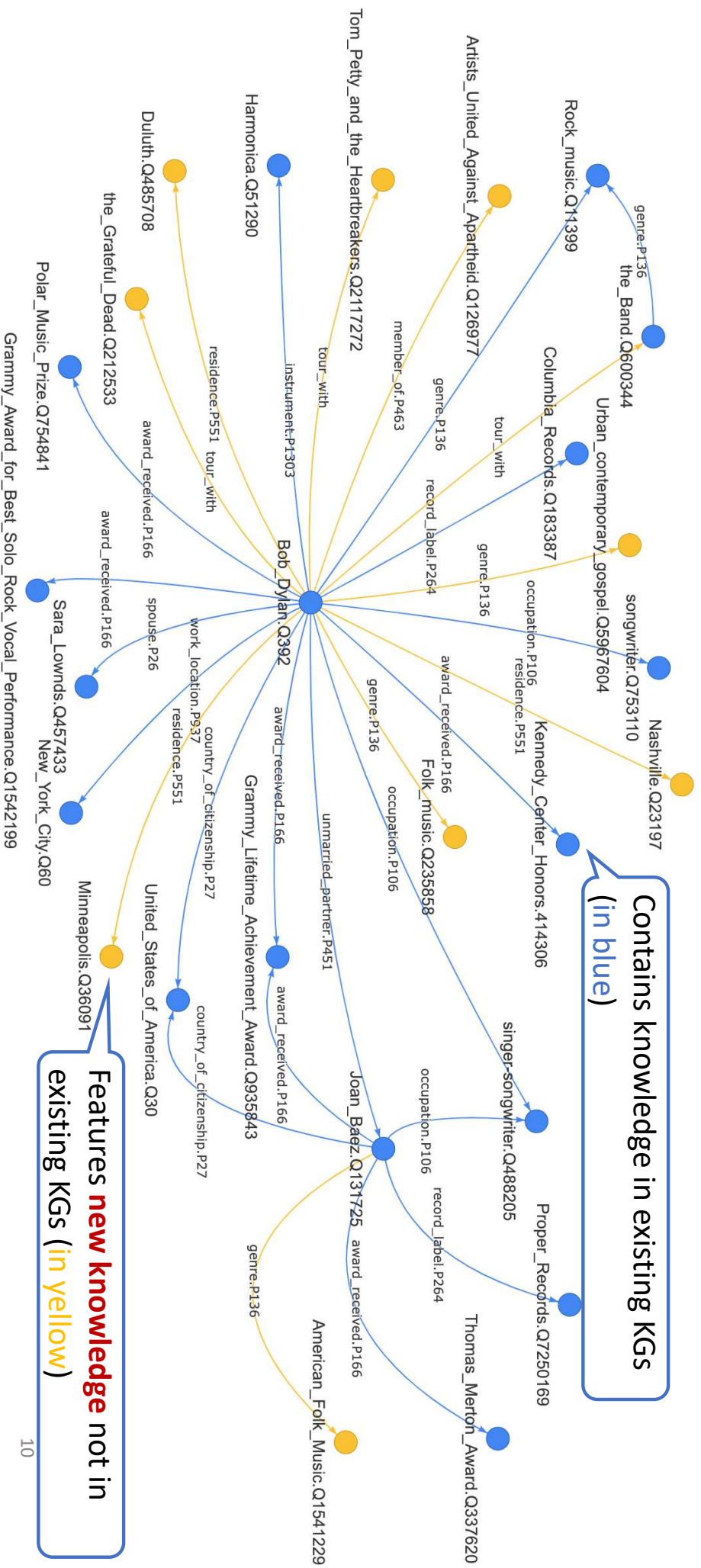
MAMA outperforms the previous best system by **2.6%**

Larger/deeper LMs produce KGs of higher quality

Results on Wikidata



An Open KG Example



An Open Question

- **Assumption:** Language models and knowledge graphs just encode the same world knowledge in two different formats, thus the two formats should be equal. But,

Is the assumption TRUE?

Language Models  Knowledge Graphs

Contributions

- **Problem:** How to construct knowledge graphs from pre-trained language models.
- **Approach:** An unsupervised two-stage approach that constructs knowledge graphs with a single forward pass of the pre-trained language models without fine-tuning over the textual corpora (outperforming compared methods by over 5.6% in F1 on Wikidata).
- **Result:** Open knowledge graphs not only cover the knowledge already in existing knowledge graphs (e.g., Wikidata), but also features open factual knowledge that is new.