#### Visual Recognition with Humans in the Loop S. Branson et al.

Presented by Thomas Provan

### Problem

How can we augment existing computer vision systems to make them practical to use?

# Motivation

- Existing systems are not current good enough for practical purposes
- Systems are focused on tasks which humans are naturally good at

# Motivation

(A) Easy for Humans





Chair? Airplane? ...

#### (B) Hard for Humans (C) Easy for Humans



Finch? Bunting?...





Yellow Belly? Blue Belly? ...

# Human In The Loop

- Combine an object recognition system with a user interaction system
- Enables the combined system to accomplish practical results in more complicated settings
- Attempts to minimize human labor

# Human In The Loop



The bird is a Black-footed Albatross

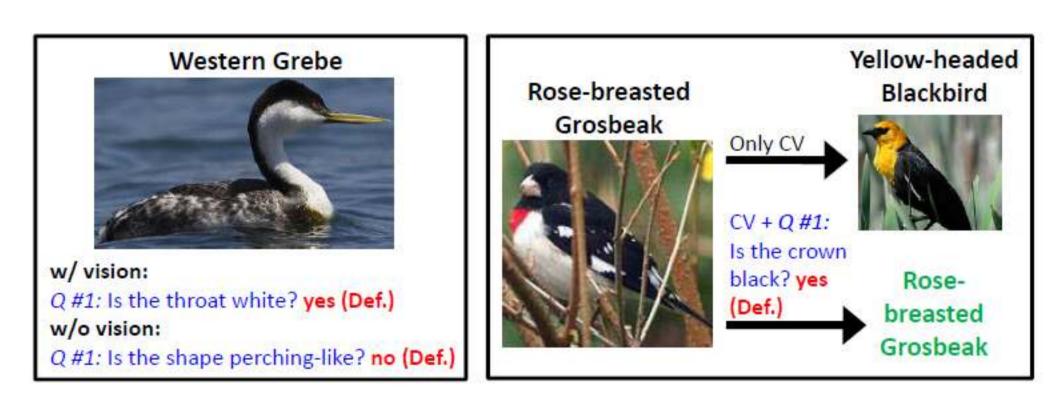


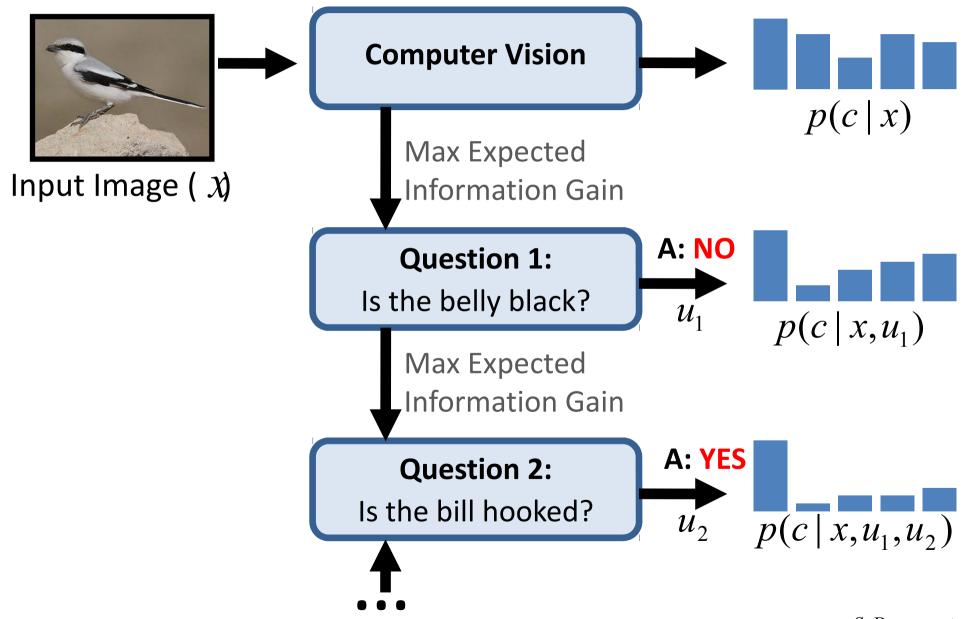
Is the belly white? **yes** Are the eyes white? **yes** The bird is a **Parakeet Auklet** 



Is the beak cone-shaped? yes Is the upper-tail brown? yes Is the breast solid colored? no Is the breast striped? yes Is the throat white? yes The bird is a Henslow's Sparrow

# Human In The Loop

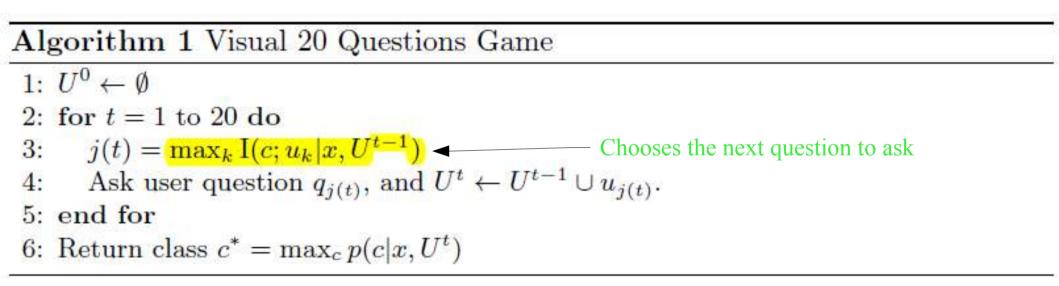




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Algorithm 1 Visual 20 Questions Game

- 1:  $U^0 \leftarrow \emptyset$
- 2: for t = 1 to 20 do
- 3:  $j(t) = \max_k I(c; u_k | x, U^{t-1})$
- 4: Ask user question  $q_{j(t)}$ , and  $U^t \leftarrow U^{t-1} \cup u_{j(t)}$ .
- 5: end for
- 6: Return class  $c^* = \max_c p(c|x, U^t)$



Select the next question that maximizes expected information gain

• Easy to compute if we can to estimate probabilities of the form:

$$p(c \mid x, u_1, u_2...u_t)$$

$$p(c \mid x, u_1, u_2..$$

$$p(c \mid x, u_1, u_2...u_t) \approx p(u_1, u_2...u_t \mid c) p(c \mid x) / Z$$

$$Model of user$$

$$responses$$

$$vision$$

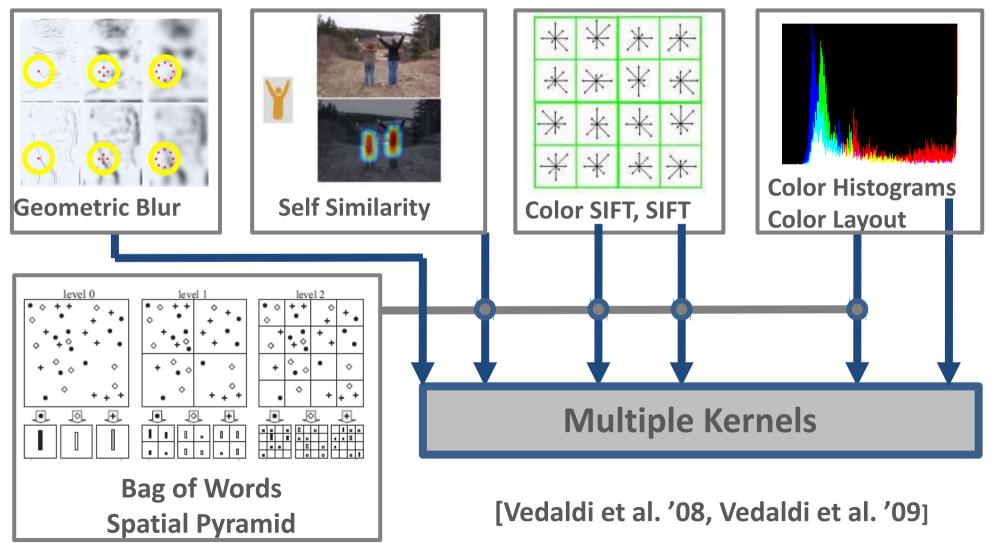
$$factor$$

$$restimate$$

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### **Incorporating Computer Vision**

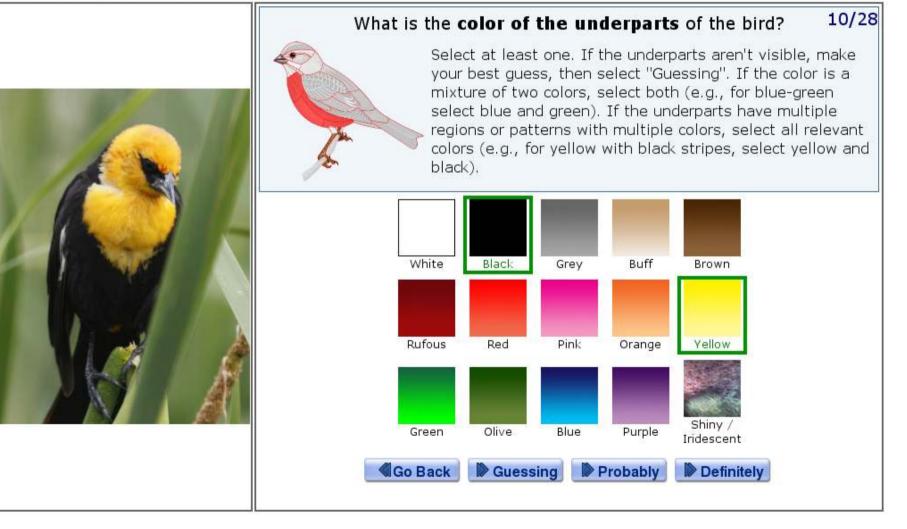
•Used VLFeat and MKL code + color features



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### **Twenty Questions**

You will be asked to answer a series of questions based on identifying visual features from the bird image on the left. Closely follow the specific instructions for each question. Holding the mouse over each selectable option for 1 second will provide additional instructions or examples.



#### Birds 200 Dataset

- •200 classes, 6000+ images, 288 binary attributes
- •Task: Classify the particular bird species



Black-footed Albatross



Groove-Billed Ani



Parakeet Auklet



**Field Sparrow** 

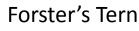


**Vesper Sparrow** 



Arctic Tern







Common Tern



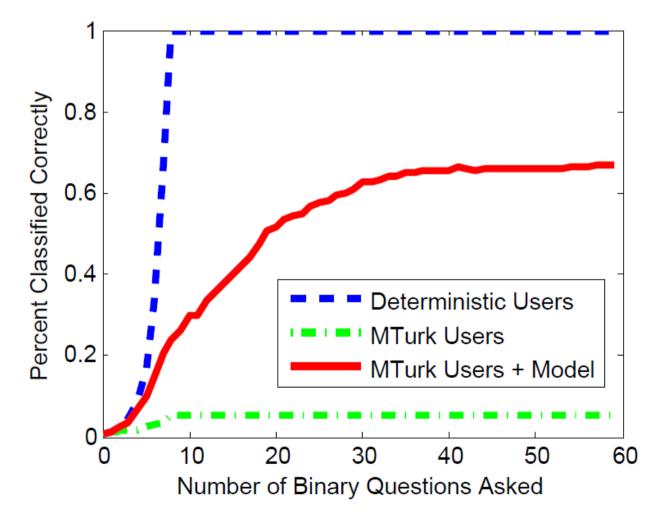
Baird's Sparrow



Henslow's Sparrow S. Branson et al

#### **Results: Without Computer Vision**

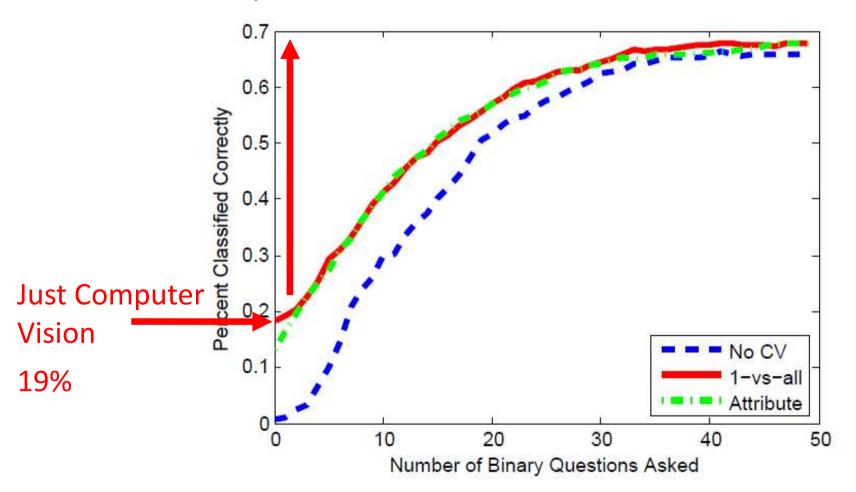
**Comparing Different User Models** 



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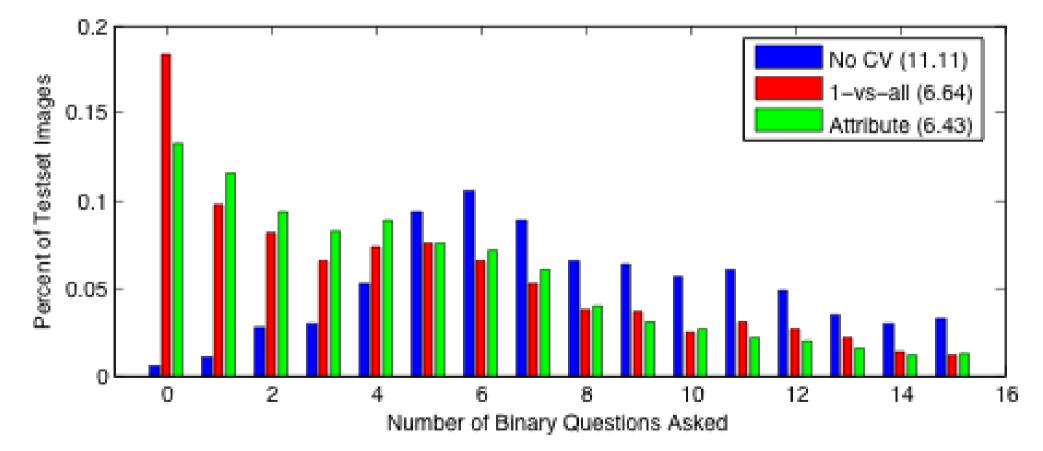
#### **Results: With Computer Vision**

Users drive performance:  $19\% \rightarrow 68\%$ 

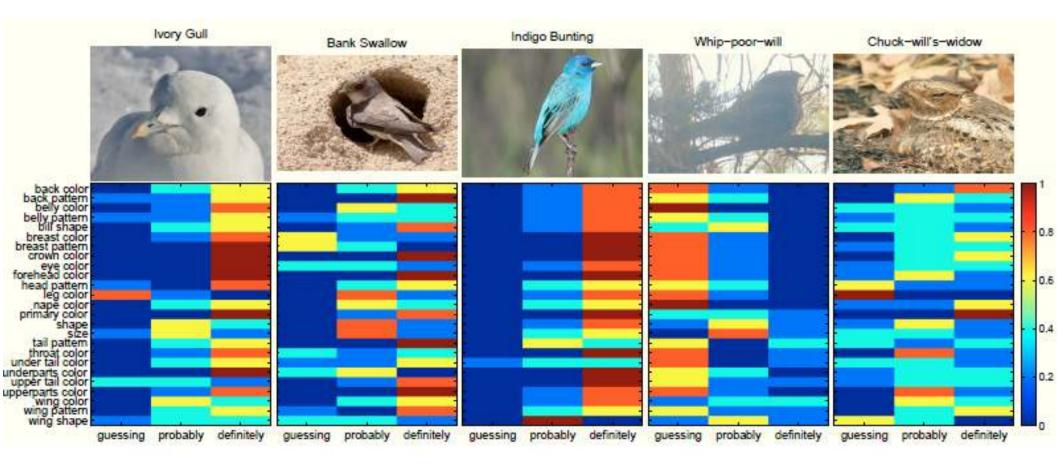


#### **Results: With Computer Vision**

Computer Vision Reduces Manual Labor: 11.1  $\rightarrow$  6.5 questions



## Results



### **Recognition is Not Always Successful**

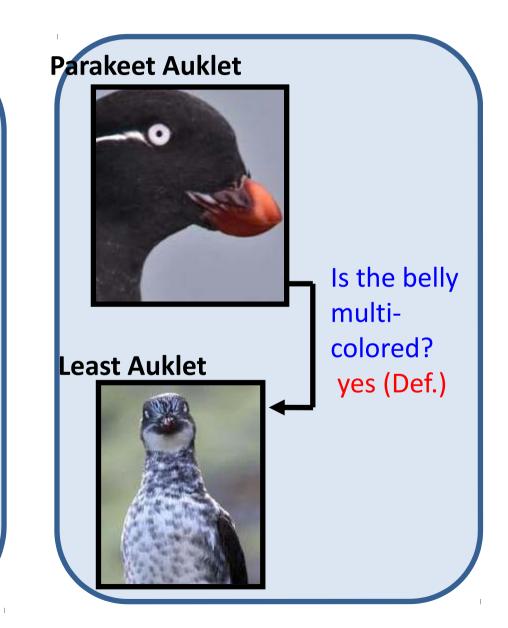
#### **Acadian Flycatcher**



#### **Unlimited questions**

#### Least Flycatcher





# Future Work

- Different recognition systems
- Different interaction systems
- Tailored systems specifically for this framework