

Current challenges in understanding the role of enhancers in disease

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1 Current challenges in understanding the role of enhancers in disease

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Preface

Enhancers play a central role in the spatiotemporal control of gene expression and tend to work in a cell type-specific manner. In addition, they are suggested to be major contributors to phenotypic variation, evolution and disease. There is growing evidence that enhancer dysfunction due to genetic, structural, or epigenetic mechanisms contributes to a broad range of human diseases referred to as enhanceropathies. Such mechanisms often underlie the susceptibility to common diseases, but can also play a direct causal role in cancer or Mendelian diseases. Despite the recent gain of insights into enhancer biology and function, we still have a limited ability to predict how enhancer dysfunction impacts gene expression. Here, we discuss major challenges that need to be overcome when studying the role of enhancers in disease etiology and highlight opportunities and directions for future studies, aiming to disentangle the molecular basis of enhanceropathies.

Main text

Regulation of gene expression is accomplished through the integration of events at regulatory elements that are proximal (promoters) and distal (enhancers) to gene transcription start sites (TSSs). Forty years after their discovery¹, enhancers are recognized as playing a central role in the spatiotemporal control of gene expression underlying human development and homeostasis². Enhancers are short stretches of DNA that act as positive regulators of transcription via their ability to bind key proteins – transcription factors (TFs) – and complexes that control gene expression. Enhancer regulation of genes involves the three-dimensional topology of chromatin, affecting the frequency by which enhancers and gene promoters come into close proximity. Through this topology, several configurations can arise beyond single enhancer-gene pairs, including one-to-many and many-to-many enhancer-gene wirings, which may affect the robustness, strength or specificity of gene expression (Fig. 1A).

Enhancer dysfunction has emerged as a central mechanism in the pathogenesis of certain diseases^{3,4} (**Table 1**). In particular, the dysfunction of enhancers by either point mutations or structural variants is a significant mechanism underlying aberrant gene regulation in cancer⁵ and Mendelian diseases⁶. Moreover, genetic variants associated with common diseases are frequently found in cis-regulatory elements including enhancers^{7–12}. Depending on the nature of the genetic alteration, enhancer dysfunction can be classified into two main types⁴. The first type involves small single nucleotide variants (SNVs) and indels in the enhancer sequence that lead to changes in enhancer activity (Fig. 1B). Such variations can for instance alter the affinity of bound TFs or create new binding sites. The second type involves structural variants that lead to deletion, duplication or relocation of the entire enhancer, which impacts chromatin topology and enhancer function (Fig. 1C). Chromosomal rearrangements can lead to rewiring of enhancer-gene connections, which may involve both enhancer adoption/hijacking (gain-of-function; e.g., ref. 13,14), and enhancer disconnection (loss-offunction; e.g., ref. 15). Depending on the genomic alteration, enhancer dysfunction may result in either gain or loss of gene expression in a given tissue as well as more complex alterations of expression patterns (Table 1).

Elucidating the molecular basis of enhancer function in normal and pathological conditions has far-reaching translational implications. Here, we discuss important challenges that need

to be considered in the study of enhancer dysfunction in disease and highlight critical areas of research to address these challenges in the near future (Fig. 2).

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Current challenges in characterizing enhancer dysfunction in disease

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1) Identifying and validating functional enhancers

A persisting challenge when studying transcriptional regulation in health and disease is to systematically identify functionally meaningful enhancers in a given cell type. Partially, this is because enhancers encompass a diverse group of regulatory sequences, which may utilize various mechanisms to control gene expression^{2,16}. Conceptually, a nucleotide sequence can be assigned as a biologically meaningful enhancer once it is experimentally demonstrated that it modulates the transcription of a gene in cis in its native context. Unfortunately, there is no high-throughput assay that is able to do this globally outside of cell lines. As a result, enhancers are often defined in an indirect, operational, manner². For example, a sequence may be functionally qualified as an enhancer if it increases the activity of a (minimal) promoter in a plasmid reporter assay, or it may be qualified as an enhancer by association, when linked with chromatin accessibility, transcription of enhancer RNAs (eRNAs), or marked by epigenomic features that have been linked to enhancer activity (e.g. p300, H3K4me1, H3K27ac). Functionally cataloging sequences as candidate enhancers has been boosted by the development of Massively Parallel Reporter Assays (MPRAs) that allow systematic largescale testing of enhancer activities of any sequence in episomal contexts², while genome-wide approaches to find putative enhancers by association were employed by several international consortia (ENCODE¹⁷, ROADMAP¹⁸, FANTOM⁷ and BLUEPRINT¹⁹). While these approaches have greatly expanded our ability to map enhancers, they suffer from some limitations. One is that not all sequences that are predicted to act as enhancers based on MPRAs or association strategies necessarily function to increase gene transcription in their endogenous chromosomal contexts. Furthermore, MPRAs do not account for the effects that linear distance and chromatin environment might have on enhancer activity. In principle, CRISPRbased enhancer screens are able to overcome these limitations and can be used to assess the importance of enhancers in their endogenous context^{20–22}. However, these assays may suffer from the intrinsic redundancy or additive effects of enhancers (see Challenge 3), and high false-negative rates². To further complicate things, some proven biologically meaningful enhancers may lack the expected biochemical marks while others may not show enhancer potential in reporter assays^{23,24}, perhaps because these assays are based on plasmids and may not reproduce the function of chromatin-dependent enhancers^{24,25}. Therefore, it is not surprising that there is no strict overlap¹⁶ between the hundreds of thousands of putative enhancers in the human genome predicted based on indirect assays and the number of biologically confirmed enhancers. Future comparisons between predicted and functionally validated enhancers, including CRISPR screens with high-sensitivity and low false-negative rates, should help improve our definition of the fundamental features of functionally operational enhancers.

2) Defining the spatiotemporal window in which a regulatory variant affects enhancer activity

Enhancer activity is highly specific during development, across cell types, cell states, and stimulatory conditions (e.g., inflammatory response, diet, drug treatment). As a consequence, the effect of enhancer dysfunction (**Fig. 1B**), may only be revealed in specific contexts or tissues. This leads to the challenge of defining the spatiotemporal window in which an enhancer dysfunction has a measurable and meaningful impact. This is particularly important for interpreting disease-associated genetic variants in non-coding regions since the cell type in which an enhancer is active can be informative about the disease mechanism. For instance, the finding that genetic variants associated with Alzheimer's disease, a neurodegenerative condition, overlap enhancers in myeloid cells, rather than neurons²⁶, has led to a shift in the research focus of the pathology²⁷. Similarly, functional assessment of obesity-associated variants has identified putatively causal variants with regulatory properties in both adipose and neuronal cell lines²⁸.

To capture inducible and context-dependent enhancers, as well as those restricted to rare cell types or developmental stages, efforts in enhancer mapping need to focus on different stimulatory conditions, environmental contexts, developmental stages and rare cell types^{29–31}. Single-cell technologies are particularly well-suited for studying rare cell types. Specifically,

single-cell chromatin accessibility assays (scATAC-seq, Assay for Transposase-Accessible CRickelshromatin and sequencing) can serve to operationally predict enhancer activity and have facilitated the functional interpretation of disease-associated non-coding variants in adult and fetal tissues^{32–34}. scATAC-seq will be key for expanding putative enhancer maps in diverse (rare) cell types. Furthermore, since chromatin accessibility does not necessarily reflect enhancer activity⁷, further development of single-cell technologies that employ orthogonal measures of enhancer activity, e.g., large-scale perturbation assays^{35,36}, will be crucial to get more confidence in assessing enhancer function for rare cell types.

In line with the importance of cataloging enhancers and their restricted activities, there is an urgent need to assess the functional impact of regulatory variants during development and differentiation. To this end, recent years have seen a promising development of biological models such as transgenic mice and zebrafish^{37–39}, genetically manipulated human-induced Pluripotent Cells (hiPSC) or immortalized precursor cells^{15,27,40,41}, human-mouse chimeras^{42,43} and organoids⁴⁴. These *in vivo* and *ex vivo* models, in combination with assays to assess developmental and differentiation potential, will facilitate the study of genetic variants and determine their impact in contexts closer to human diseases.

3) Understanding the interplay between cis-regulatory elements

Enhancers are not only highly context-dependent, but they also often work together in regulatory domains to achieve the correct gene expression output. Thus, a major challenge is to understand the interplay between enhancers and other regulatory elements, including promoters, and how the joint activity of a domain is influenced by disruptions of individual enhancers. Multiple enhancers for the same gene may allow distinct enhancers to either be activated under different conditions or to cooperate, both of which can lead to robustness in gene activity^{24,37,45–47}. For instance, many developmental genes are associated with "shadow" enhancers with similar transcription factor (TF) binding to ensure robust expression under suboptimal conditions^{45,48,49}, an observation that has been confirmed by 3D topology-based methods that revealed a complex landscape of multiple enhancer interactions per gene^{50–53}. In fact, highly coordinated enhancer activity has been linked to the regulation of cell identity genes⁵⁴, signal integration and compartmentalization of the genome⁵⁵. As a consequence of

such regulatory complexity, many enhancers might not, individually, reveal a strong phenotype when disrupted in their endogenous context^{23,56}, while still possessing endogenous enhancer activity. Thus, the presence of multiple enhancers⁵⁷ *per* gene may either additively or synergistically achieve a higher transcriptional output of a gene or provide redundancy and mutational robustness to its expression. Systematic testing of enhancer-promoter compatibilities will help to better understand the still unclear connectivity rules^{58,59} that control gene transcription in the human genome.

Elucidating the mechanisms and contexts, including the cell type-specific 3D topology, by which regulatory domains and TFs establish robustness or synergism will therefore be crucial to further our understanding of enhanceropathies. Combinatorial interference or perturbation of multiple enhancers within a regulatory domain will be necessary to understand the principles by which enhancers act together and their effects on gene regulation.

4) Identifying the target genes of enhancers

Enhancers ultimately need to be defined by their role in enhancing endogenous gene expression, which leads to the next challenge: the identification of their target genes. This is particularly challenging for enhancers that are located distally to any gene promoter. It is assumed that distal enhancers have to come into physical proximity to their target gene in order to function, as first demonstrated by chromosome conformation capture (3C) methods in the beta-globin locus⁶⁰. Thus, for the operational mapping of target genes, chromatin-topology assays are key to determine the physical proximity between enhancers and their putative genes. These technologies can map direct contacts (chromatin loops) and at the same time identify larger domains, so-called topologically associating domains (TADs), which have a high density of physical chromatin interactions⁶¹. The main caveats of using direct contacts for mapping enhancer-gene pairs are that the 3C-technologies typically require large cell numbers (with some exceptions^{50,52} and may thus miss enhancer-gene pairs that are looped only in a subset of cells or contacts that are highly transient. TAD-based analyses suffer from low resolution since they typically comprise multiple genes and enhancers and can, on their own at best, restrict the search space for putative target genes⁶². There are a

complementary set of approaches to map enhancer-gene pairs such as targeted Hi-C, where chromatin interactions of regions of interest such as promoters and/or enhancers are captured to increase resolution^{55,63}, or expression quantitative trait loci (eQTL) mapping, where enhancer genetic variants are associated with mRNA expression changes across individuals⁶⁴. Other approaches use covariation between molecular phenotypes (e.g., histone marks, chromatin accessibility, expression) of enhancers and genes across individuals or cell types^{9,65-67} or combine chromatin states and long-range interactions⁶⁸, to construct genome-wide maps of enhancer—gene connections in a given cell type. The advantage of these methods is that relying on enhancer-gene co-variation does not assume a specific mechanism of how enhancers regulate gene expression, and can therefore also capture transient enhancer-gene contacts⁶⁹. Here, the caveats are that these methods require molecular data across a large number of individuals or cell types, and they may miss constitutively active enhancers that do not vary much across samples. Given their descriptive (for the 3C technologies) and correlative (for the co-variation methods) nature, all of these approaches provide an operational prediction of putative enhancer-gene pairs. For a functional mapping of target genes, CRISPR-mediated enhancer deletion or inactivation, followed by gene expression analysis^{29,68}, is the most direct way to search for target genes. However, such CRISPR-based approaches may miss links due to low effect sizes and are often limited to cultured cells. In conclusion, current approaches still have difficulties identifying with high confidence the target genes of enhancers, and likely, the combination of different strategies might improve the efficiency of identifying disease-targeted genes⁷⁰. Recent advances in applying machine learning to predict cell-type specific expression based on DNA sequence⁷¹ show great promise to generate defined and experimentally testable hypotheses. These models were enabled by the vast resources of transcriptomics and genomics data that have been assembled by the community, and additional data, particularly from less accessible cellular states and developmental stages, will further improve the power of these methods.

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5) Understanding the grammar of enhancer activity

Gene regulatory elements, including enhancers^{7–9}, are regulated by TFs, or TF-recruited coactivators, which bind to the enhancer element at any given time and cellular state. It is thus not surprising that genetic variants that disrupt a TF binding motif are enriched among variants associated with molecular phenotypes, such as histone marks⁸ or tissue-specific expression levels⁷², and can be disease-causative⁷³. For example, a mutation in a SOX9 enhancer, associated with Pierre Robin Syndrome, disrupts the binding of the TF MSX174 (see other examples in Table 1). However, the majority of molecular trait-associated Single Nucleotide Polymorphisms (SNPs) do not disrupt known TF binding sites⁸, leading to the next open challenge in understanding enhancer dysfunction: to identify the rules by which enhancer sequence determines its activity. Concepts, such as Variable Chromatin Modules (VCMs), where the effects of a lead SNP affecting a local chromatin domain (e.g. through TF binding site disruption) spread into the local vicinity, can explain the missing mechanism to some extent^{8,10,75}. Recent studies revealed that flanking regions of TF binding sites are highly informative for some TFs to bind⁷⁶ and they impact the enhancer potential of the encompassing regulatory element⁷⁷, suggesting we are still missing part of the grammar for TF binding. In line with this, up to 30% of human TFs have no characterized binding motif⁷⁸. Consequently, interpreting regulatory variant-to-phenotype associations requires fundamental insights into the sequence determinants of TF binding and enhancer activity. Here, sequence-based machine learning to model TF binding^{76,79,80}, enhancer activity^{25,77,81} and topologies⁷¹ show promise. However, major challenges remain, including the difficulty to accurately interpret such models, the lack of sufficient training or validation data, and the need to improve accuracy and generalization across cell types/contexts. In parallel, experimental approaches that measure the functional impact of genetic variants on regulatory activity and TF binding in a large-scale, such as MPRA-based approaches^{82–86} and SNP-SELEX⁸⁷, can provide comprehensive experimental fine-mapping of likely causal variants. Overall, these insights will be crucial for the interpretation of the potential effect and severity of enhancer dysfunction, and thus the potentially implicated genetic variants, within complex regulatory domains.

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6) Understanding how TF cooperation defines enhancer activity and specificity

Enhancers integrate non-mutually exclusive layers of molecular information: their function can be impacted by genetic variants/mutations, by epigenetic chromatin remodeling that is typically set up by lineage-specific TFs, or by signaling cascades regulated by stimulus-responsive TFs⁸⁸ (**Fig. 1**). Here, we focus on the challenge of understanding the role of TFs and

epigenetics on enhancer dysfunction. Lineage- and developmental-stage specific enhancers, typically regulated by lineage-specific TFs, may define the gene expression potential of a cell, and whether or not it will be able to mount a specific response to a given stimulus⁸⁹. In particular, during development or differentiation, enhancers and whole chromatin domains can be primed in progenitor cells towards certain lineages before gene expression changes are obvious, e.g., during adipogenesis⁵⁵. In contrast, enhancers that are under the control of stimulus-responsive TFs essentially act as signaling response elements and connect cellextrinsic signals to gene expression programs. Conceptually, lineage-specific TFs and the chromatin accessibility landscape they set up determine the scope of stimulus-regulated TFs. This way, stimulus-responsive TFs can access enhancers that are pre-marked and kept accessible by lineage-specific TFs, thus integrating the two layers of regulation⁹⁰. As a consequence, some response-TFs, such as NF-kB, bind completely different enhancers depending on the cell type in which they are activated⁹¹. This is consistent with observations that a TF can regulate completely different sets of genes depending on the cell type⁹², which is partially explained by the cooperative interaction of TFs⁹³. Yet, apart from a couple of wellstudied examples, very little is known about the contribution of TF cooperativity, enhancer priming (that can also be TF mediated) and permissive chromatin, which in turn may define the TF regulon (i.e., the set of target genes regulated by a given TF). To fully understand enhancer dysfunction, it is important to study the cell type- or condition-specific TF regulons, and how they are defined by the combinatorial or cooperative binding grammar of enhancer sequences in normal and pathological conditions. Diverse TF-centric studies are even more important given the current literature bias with many studies focusing on a small set of TFs while the majority of TFs are vastly understudied⁹⁴.

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7) Deciphering the impact and interactions of regulatory mutations in disease

The challenges above culminate in the ultimate challenge of identifying and understanding pathogenic enhancer dysfunction and eventually using this knowledge in clinically relevant studies (**Fig. 2**). The specific challenges that need to be solved for understanding a certain disease depend on the type of enhancer dysfunction and the nature of the genetic alteration (rare vs. common). For rare diseases, few examples of causal enhancer mutations have been established as compared to mutations in the coding genome. It is currently unclear whether

this limited number of reported enhancer mutations in rare disorders is because they do not exist or because we have not been able to find them due to the lack of data and statistical power. Either way, the additional challenge for identifying causal mutations in enhancers vs. coding regions is that each genome carries around 2,000 structural and 8,500 private noncoding variants⁹⁵, which are often not even captured since exome-sequencing is still the standard for diagnosing rare diseases. On the other hand, GWA studies have revealed hundreds of non-coding variants of significance for common disease risk, suggesting that the aggregated effect of variants in multiple enhancers modulate common disease risk. Finemapping studies aimed at identifying the causal variant(s) among those linked in a haplotype block typically integrate significantly associated variants with experimentally determined enhancer characteristics, as discussed above. While successful for the identification of some causal variants (e.g. 96,97), this is often difficult because the relevant cell type and trans-acting nuclear environment are not known (challenge 2), the role of the encompassing regulatory domain is not well understood (challenge 3) and the target gene of the affected enhancer is not identified (challenge 4). Fine-mapping of causal signals and effect size predictions can be improved by expanding the battery of GWA studies with cohorts of diverse ancestries^{98,99}, and computational tools ranking genes based on their dosage-dependent pathogenicity. This allows hypothesis-driven studies where candidate target genes and enhancers are tested simultaneously to measure their combined effects on inferred functions¹⁰⁰. Furthermore, for common diseases, both genetic and environmental factors contribute to the disease etiology. Therefore, the effects of certain non-coding genetic variants might only or preferentially be manifested under certain environmental conditions. Together with a significant shift for using whole-genome instead of whole-exome sequences as a diagnostic utility, and consequently, an increasing amount of whole-genome data accumulating thanks to biobanks and cohort studies¹⁰¹, these tools will likely provide much better constraints on assessing disease causality and could pave the way towards systematic prediction of pathogenicity of regulatory variants and mutations for both rare and common variants 102.

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Future directions

As disease-associated regulatory mutations at enhancers are increasingly identified, there is an urgent need to fully characterize enhancer mutations to enable their use in functional and clinical genomic studies (Box 1). Besides the complexity of studying enhancer function in normal contexts, the characterization of noncoding variants affecting enhancer activity in disease adds additional challenges ranging from the identification of a credible set of regulatory variants to the identification of tissues and developmental contexts in which variants have an effect. Despite the wealth of data on enhancer activity across multiple celland tissue-types, it is challenging to fully utilize the vast potential of such datasets, highlighting the importance of good data-sharing practices. In addition, the majority of available data informing on enhancer activities are derived from populations of cells, disregarding the stochasticity and plasticity of regulatory events across individual cells. However, due to the complexity of the regulatory landscape, we propose that the field should move beyond the generation of enhancer catalogs and invest more in experimental and computational efforts to identify their target genes, in particular for the prioritization of disease-relevant genes susceptible to dysfunction upon misregulation. This can only be uncovered using Systems Biology, computational modeling approaches, and targeted experimental systems. Focused efforts and datasets will enable hypothesis-driven investigations of a set of variants or genes for a given disease phenotype and further inform the modeling of enhancer function from catalog data. Ultimately, the acquired knowledge should allow the implementation of novel strategies to genetically or epigenetically modify enhancer function to treat the associated diseases.

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Competing interests

354 The authors declare no competing interest.

356ble 1. Representative examples of enhancer dysfunction driving disease

Type of disease	Disease	Affected	Enhancer*	Type of disruption	Effect on gene	ref
disease	Thelesses	gene(s)	I CD	Enhancer deletions	exp. LOE	103,104
Monogenic (Mendelian)	-Thalassaemia	<i>-globin</i> genes	LCR		LOE	
	-Thalassaemia	<i>-globin</i> genes	-globin enhancers	Deletion or insertion of promoters alter enhancer-gene connectivity	LOE & GOE	105,106
	PDD2	SHH	ZRS	Rare variant introducing a TFBS	GOE	107
	HPE	SHH	SBE2	Rare variant disrupting a TFBS	LOE	108
	Limb malformations	PAX3, IHH, WNT6	EPH4 enhancers	Deletions, duplications and inversions disrupt the boundaries of a TAD containing the <i>EPH4</i> enhancer and rewire the connectivity with different genes	GOE	13
	5q14.3 microdeletion syndrome	MEF2C	MEF2C enhancers	TAD disruption disconnects MEF2C from associated enhancer	LOE	109
	Pierre Robin syndrome	SOX9	SOX9 enhancer	A point mutation in a conserved enhancer disrupts the binding of MSX1	LOE	74
	Cooks syndrome	SOX9, KCNJ2	SOX9 enhancers	Duplication of a TAD boundary at the SOX9 locus causes neo-TAD formation and KCNJ2 misexpression	GOE	110
	Isolated atrial defect	TBX5	90 kb downstream	Rare variant abrogates heart-specific enhancer activity	LOE	111
	Isolated pancreatic agenesis	PTF1A	25 kb downstream	Rare variants abolish enhancer activity and disrupt the binding of FOXA2 and PDX1	LOE	41
Common (multifactorial)	Obesity	IRX3, IRX5	FTO intronic	Multiple variants on a common haplotype increase the activity of several enhancers	GOE	97
	Type 2 diabetes	ZFAND3	Upstream	SNP disrupts the binding of NeuroD1 and decreases enhancer activity	LOE	112
	Vascular diseases	EDN1	PHACTR1 intronic	SNP located in a distal region interacting with <i>EDN1</i> enhancer	LOE	96
	HBF level	BCL11A	Downstream	SNP disrupts TF binding and diminishes expression in erythroid cells	LOE	113
	Cardiac disorders	SNC5A	SNC10A intronic	SNP in SN10A modulates SNC5A expression in the heart	LOE	114
	Hirschsprung disease	RET	Several enhancers	Several SNPs located in <i>RET</i> enhancers act synergistically to reduce gene expression	LOE	115
	Parkinson	SNCA	Intronic	SNP alters bthe inding of EMX2 and NKX6-1	LOE	40
Can	Burkitt lymphoma	MYC	IgH enhancer	Somatic translocation (enhancer hijacking)	GOE	116,117
	Lung Adenocarcinoma	MYC	450 kb downstream	Somatic duplication of the enhancer	GOE	118
	T-ALL	TAL1	7 kb upstream	Somatic insertions introduce a MYB binding site and induce the formation of a Neo-enhancer	GOE	119,120
	Ph-like ALL	GATA3	Intronic	A rare variant increases enhancer activity	GOE	121
	CLL	AXIN2	Upstream	Common variation in the AXIN2 enhancer modulates CLL susceptibility via differential MEF2 binding	GOE	122
	AML	GATA2, EVI1	GATA2 enhancer	Large somatic inversion relocated <i>GATA2</i> enhancer in the vicinity of <i>EVII</i>	LOE & GOE	14
	Prostate cancer	PCAT19, CEACAM21	PCAT19 Epromoter	Common variant changes the affinity of TFs and switch promoter and enhancer activities	GOE	123,124
856a to	Tabla 1. *Enhance			lated gene unless otherwise stated. Abbreviations: LOE: Lo	as of avmassion	COE, goin

856s to Table 1: *Enhancer location relates to the regulated gene unless otherwise stated. Abbreviations: LOE: Loss-of-expression; GOE: gain-365x pression; LCR: Locus control Region; PDD2, preaxial polydactyly type II; HPE: holoprosencephaly; HBF: Fetal hemoglobin; T-ALL, T 358Acute Lymphoblastic Leukemia; CLL: Chronic Lymphoblastic Leukemia; MLL: Myeloid Lymphoblastic Leukemia; Ph-like ALL:

³⁵² delphia chromosome-like Acute Lymphoblastic Leukemia; TF, transcription factor; IgH, immunoglobulin heavy chain 360

Figure legends

Figure 1: Different mechanisms of enhancer function and dysfunction. A) Variations in the interplay between enhancers and target genes. Multiple enhancers can cooperate in a tissue to increase the transcription of a target gene or be active in different tissues to control a complex developmental gene expression pattern. Enhancers can further control multiple genes in a mutually exclusive or shared way. Color code indicates the enhancer activity and gene expression in different tissues or developmental contexts. (B-C) Erroneous regulatory wiring between enhancers and genes, by either enhancer disruption (B) or altered enhancergene connectivity (C), can result in dysregulation of gene expression and ultimately cause disease. Enhancer dysfunction can originate from deletions, duplications and mutations, which can result in either loss or gain of gene expression. Altered enhancer-gene connectivity can be caused by chromosomal translocations or large structural variations that can distort or merge Topologically Associating Domains (TAD). As a consequence, enhancer-gene connectivity can be lost or gained resulting in dysregulated gene expression. Changes in gene expression are indicated by the number of arrows.

Figure 2: Challenges to unravel enhancer-associated diseases. Elucidating the molecular basis of enhancer dysfunction in disease requires critical areas of research to be addressed, each corresponding to one of the challenges described in the main text. Resolving challenges I to VI should lead to the ultimate challenge (VII) of identifying the causal variants, the impacted molecular mechanisms as well as the affected genes of a disease. TF: transcription factor. MPRA: Massively Parallel Reporter Assay.

Box 1. Critical areas of research to further our understanding of enhancer dysregulation in disease

We emphasize the following critical areas of research to advance our understanding of enhancer dysregulation in disease and for better translation of enhancer research into clinical practice:

Area 1

The ongoing community-driven comparison and assessment of experimental approaches for the discovery of enhancer activity should be strengthened. This will help improve our definition of the fundamental features of functionally operational enhancers as well as determine the most appropriate assay given a biological or disease context. We particularly

see the benefits of further developments of CRISPR screens that improve sensitivity and allow measurements in non-cultured cells.

Area 2

We foresee major benefits in further efforts towards developing assays that will allow accurate assessment of enhancer activity in single cells. scATAC-seq is key for expanding the enhancer map repertoire, particularly in rare cell types. In addition, further development of single-cell technologies that employ large-scale perturbations of enhancers or TFs will be key to assessing enhancer function for such cell types. The output of such studies will also help building models of enhancer regulation informed by the dynamics and stochasticity of regulatory events as well as discover mechanisms by which their perturbation contributes to pathology.

Area 3

To better understand the rules by which enhancers work together in regulatory domains to achieve robustness, specificity, or synergism, further efforts are needed to derive assays and strategies that allow combinatorial interference or perturbation of multiple enhancers. It is further imperative to develop in vivo (i.e. in situ) assays that allow the study of the activity of an enhancer in isolation or synergy with other enhancers. The outcomes of such studies would enable us to identify the biological mechanisms by which regulatory domains are formed and the rules by which TFs and the interplay between multiple regulatory elements yield robustness or additive effects. These insights will aid the interpretation of the potential effect and severity of regulatory genetic variants and enhancer dysfunction within complex regulatory domains.

Area 4

Experimental disease systems, such as humanized animal models, organoids, and engineered tissues, are becoming increasingly available for genetic engineering and *in situ* or *ex vivo* functional experiments. It will be key to fully employ these advanced disease models for assessing the functional and pathological consequences of non-coding regulatory variants (genetic and structural). Such experimental systems will allow

interrogation of enhancer activity under a relevant internal or external stimulus for their dynamic and contextual assessment.

Area 5

The research community should increase the already promising work towards developing interpretable and generalizable computational models that can accurately predict TF binding, the activity of enhancers and their target genes, using molecular measurements in any given cell type and condition. From these, the main efforts should ideally be focused on deriving the underlying regulatory DNA code, allowing for direct interpretation of the effects of genetic variants across cell types. Relatedly, we foresee great benefits in putting effort into developing approaches to computationally predict dosage-sensitive and responsive genes, as they are more likely to be adversely affected by cis-acting mutations.

Area 6

To fully understand the molecular basis of enhancer dysfunction, we foresee the need to further develop and apply large-scale TF perturbation assays coupled with GRN analysis to study cell type- or condition-specific TF regulons, and how they are defined by the combinatorial or cooperative binding grammar of enhancer sequences in normal and pathological conditions.

Area 7

Last but not least, we foresee great potential for implementing tools (e.g. CRISPR-based) to genetically or epigenetically modify the functions or chromatin contexts of enhancers to treat enhanceropathies. By targeting enhancers, one can avoid the potential pleiotropic effects associated with drugs/tools directed toward proteins or gene promoters.

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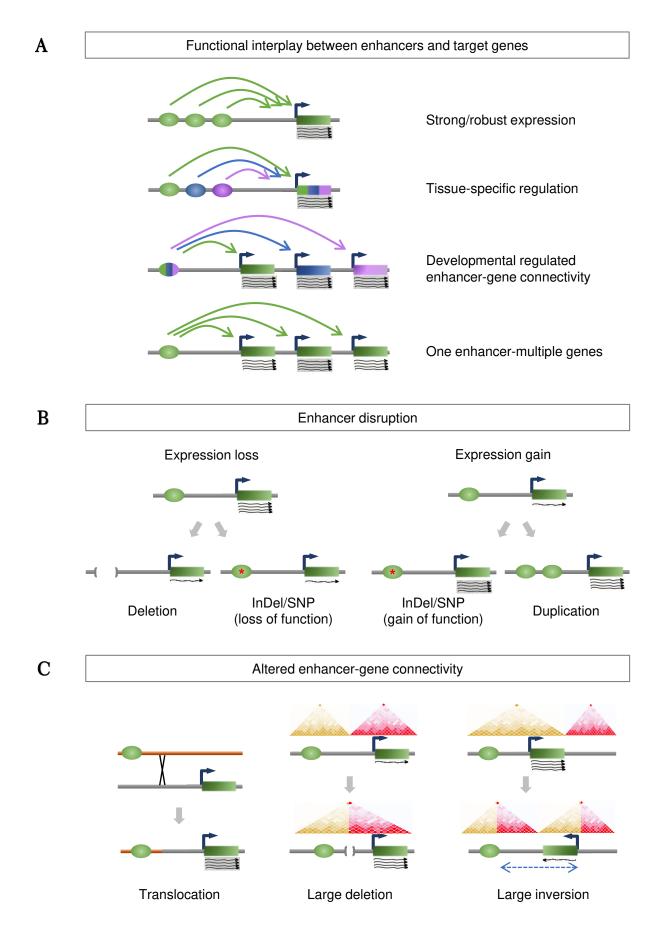
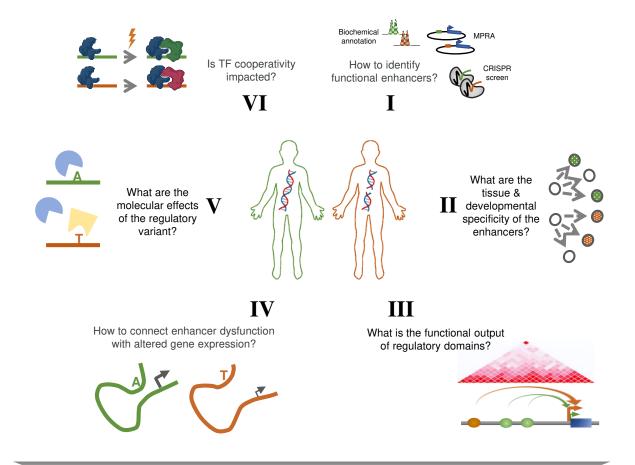


Figure 1 (REVISED)



VII Linking regulatory variants with phenotype

