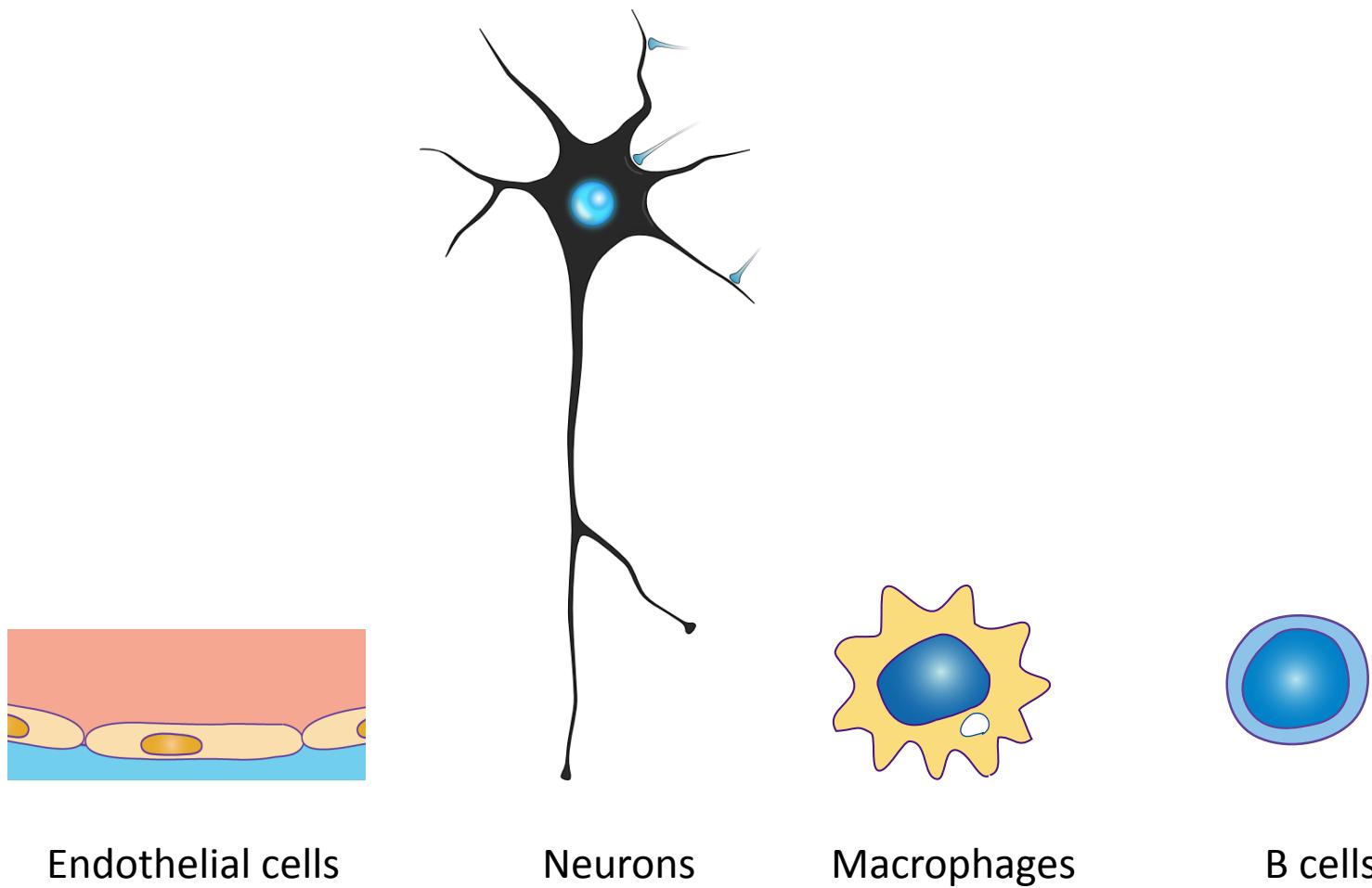


June 29-July 1, 2015
ENCODE 2015:
Research Application and Users Meeting

Selection and function of signal-dependent enhancers: The Macrophage as a Case Study

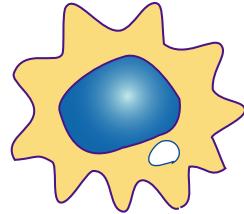
Christopher K Glass MD, PhD
Department of Cellular and Molecular Medicine
Department of Medicine
University of California, San Diego

Different cell types arise from differential transcription of the same genome

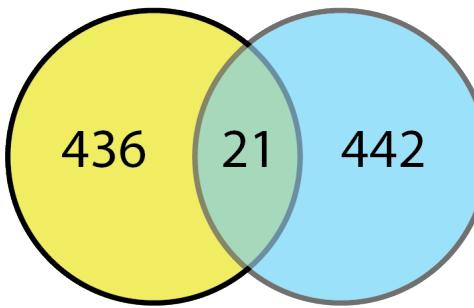


Transcriptional responses to the same signal can be cell type-specific

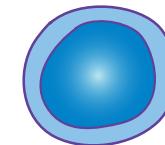
mRNAs increased > 4-fold 1h after TLR4 ligation



Macrophage

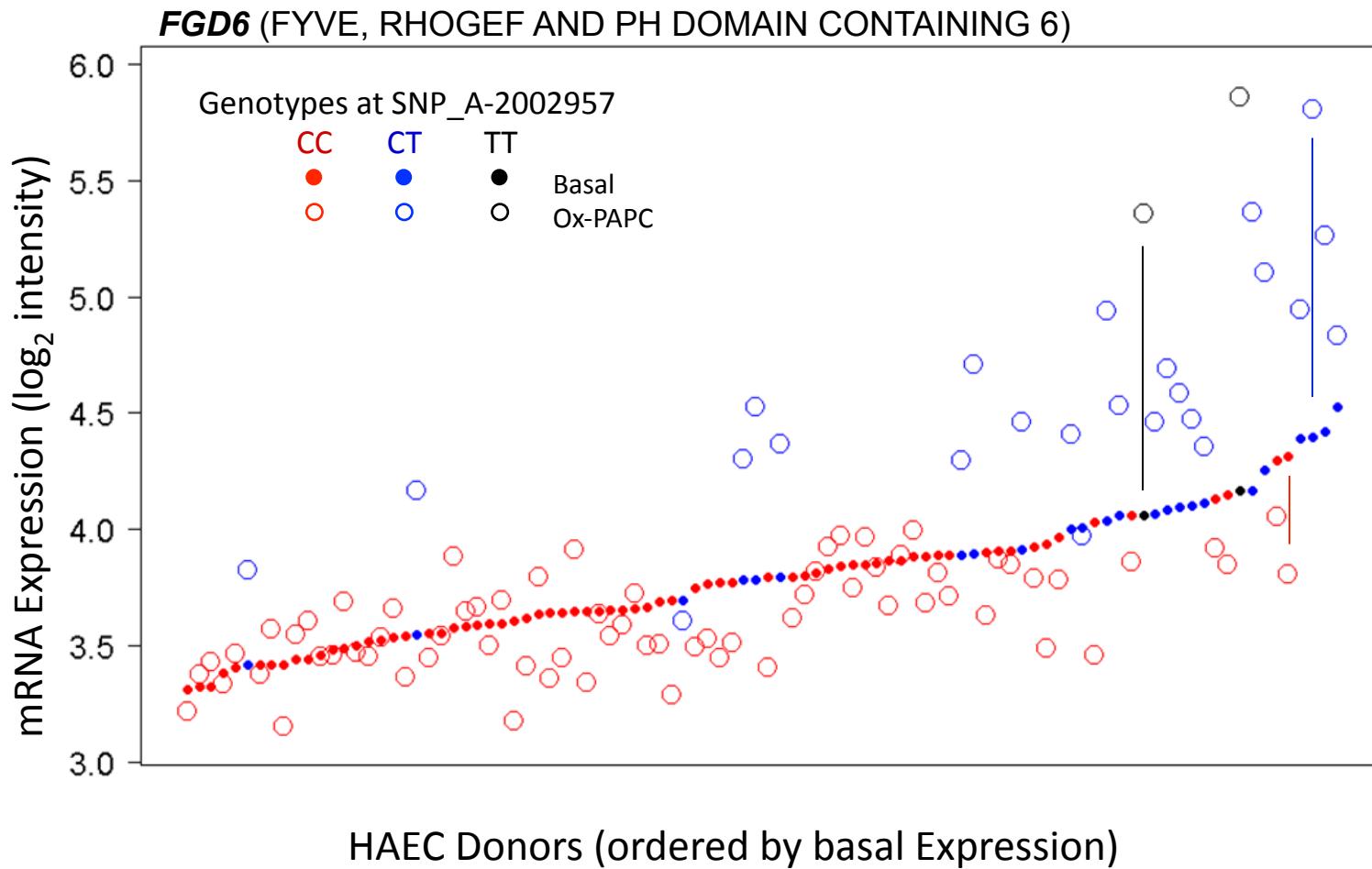


B cell

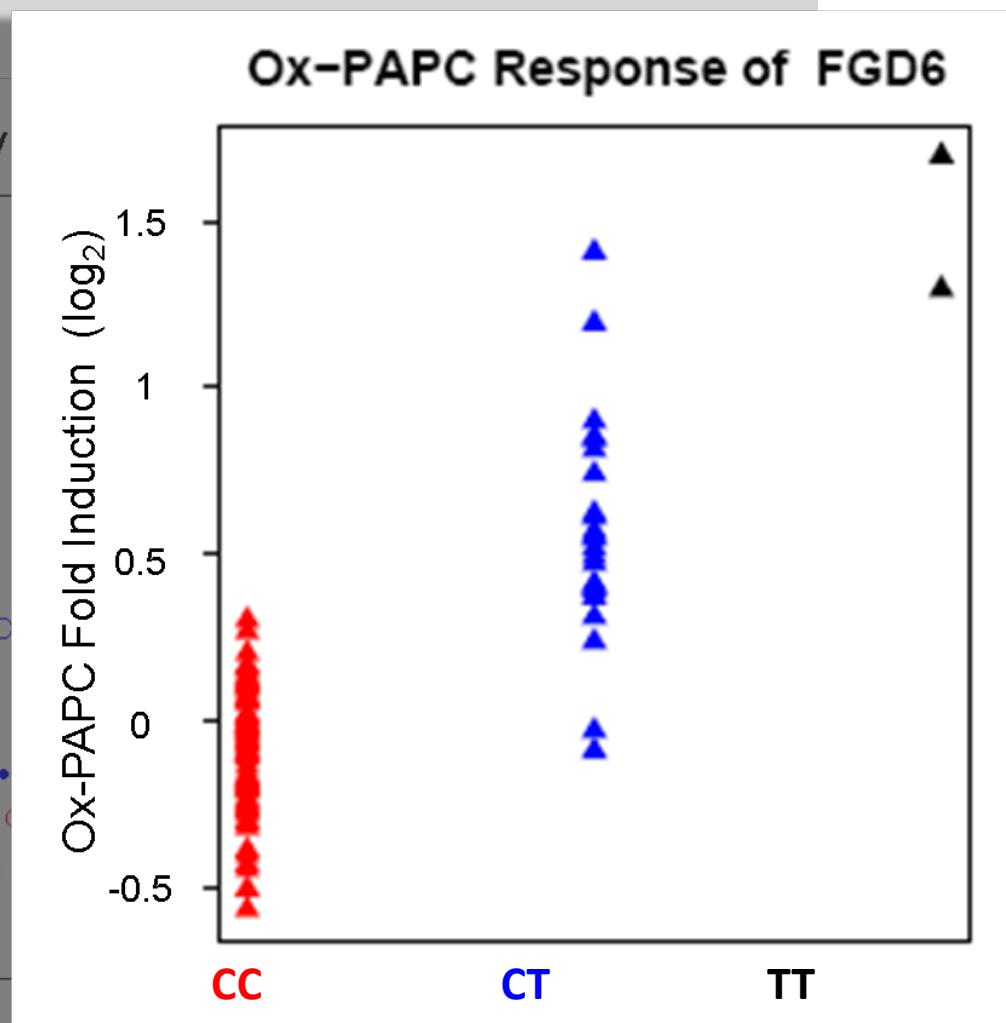
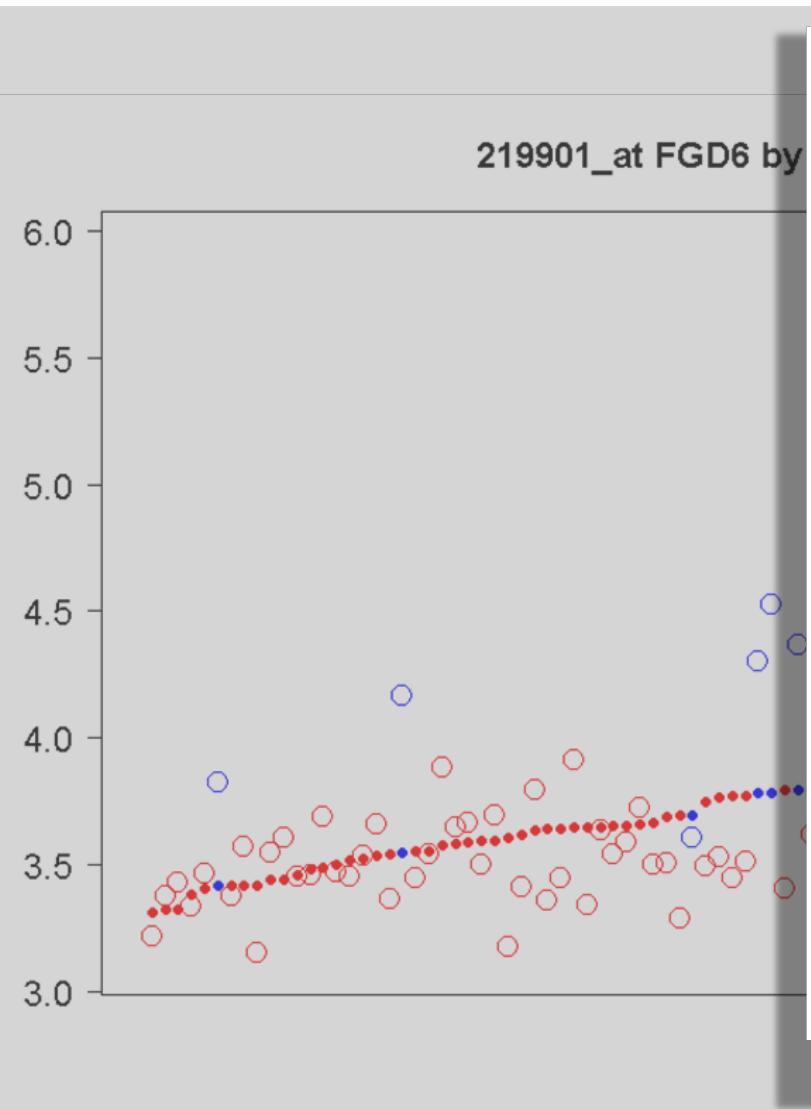


GO term	Macrophage	B cell
	p value	p value
Immune response	1e-36	> 0.05
Chemotaxis	1e-12	> 0.05
Protein folding	> 0.05	1e-10
RNA processing	> 0.05	1e-9
DNA replication	> 0.05	8e-3

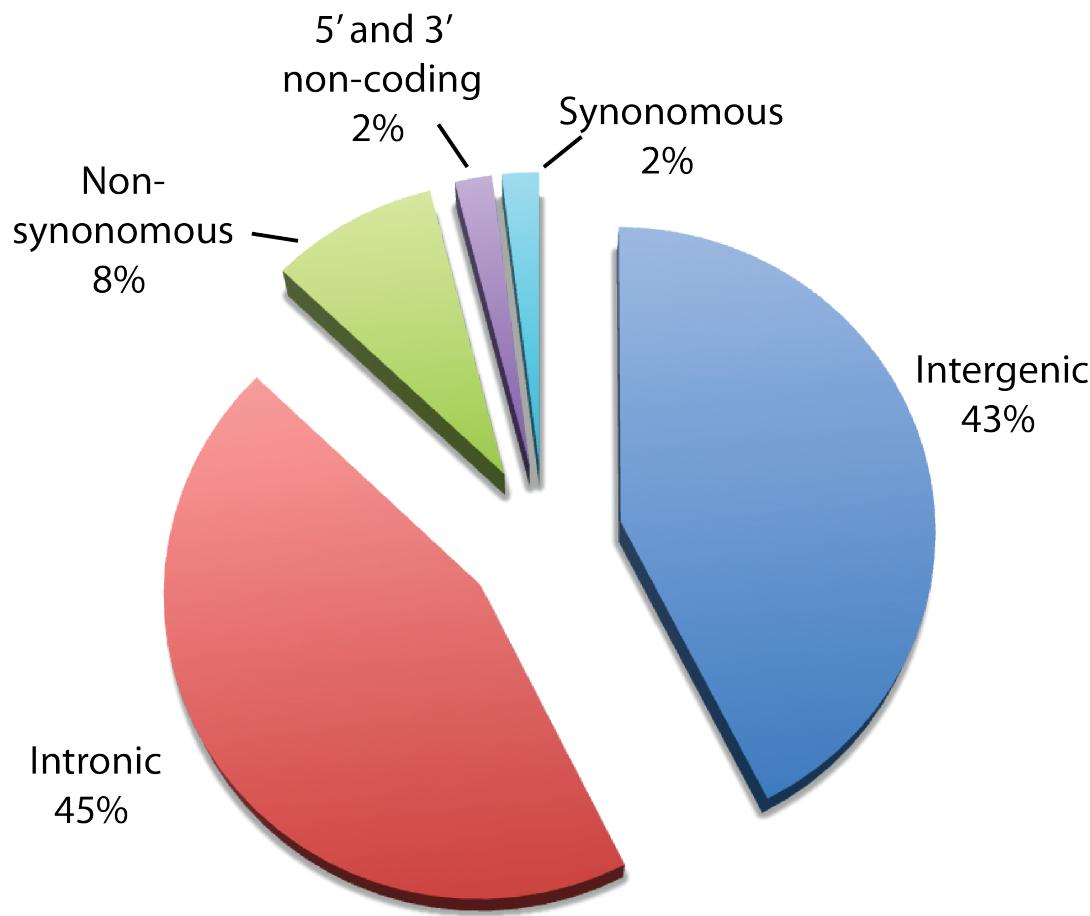
Signal-dependent responses of the same cell type can vary among individuals



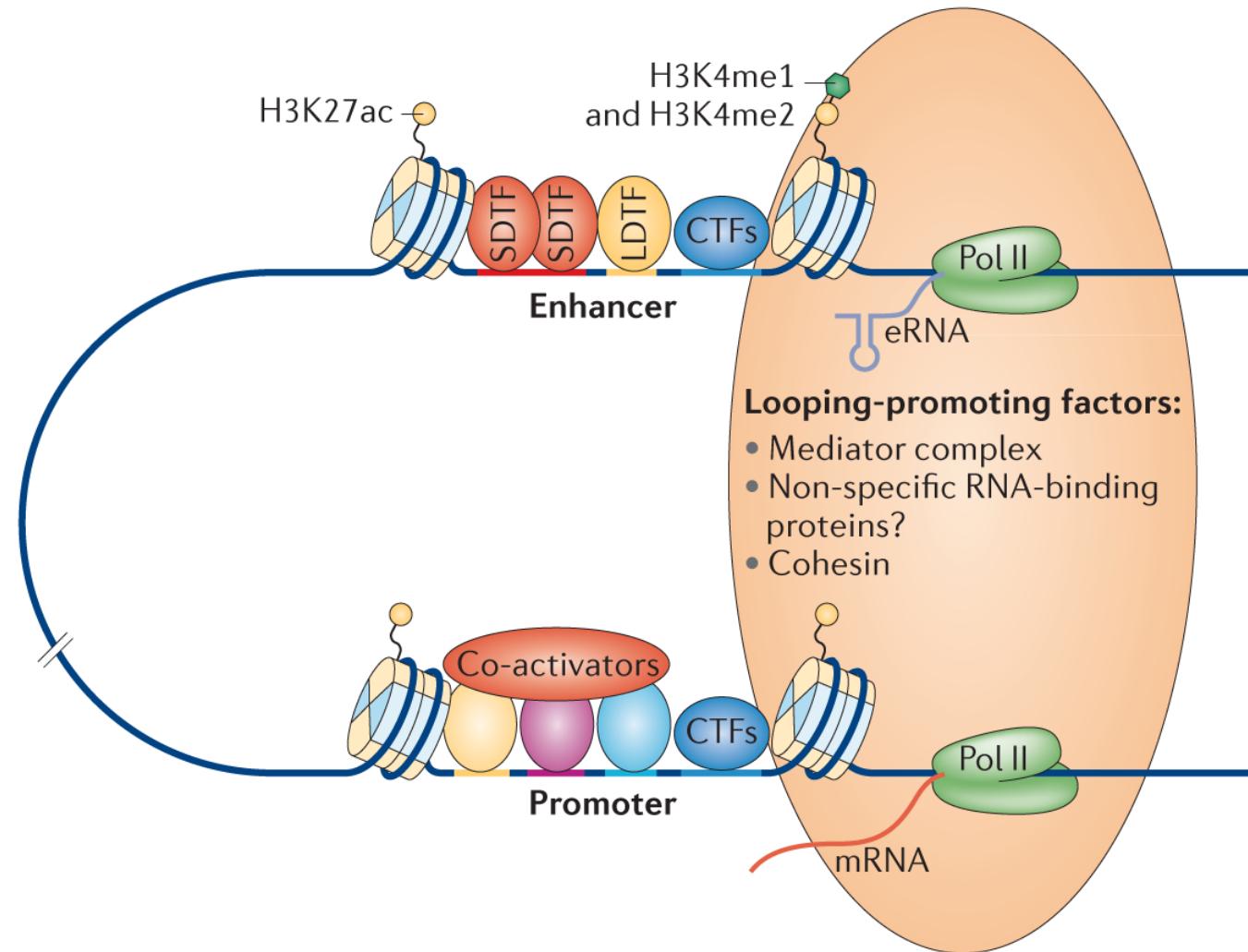
The FGD6 response to Ox-PAPC is associated to a local DNA variant



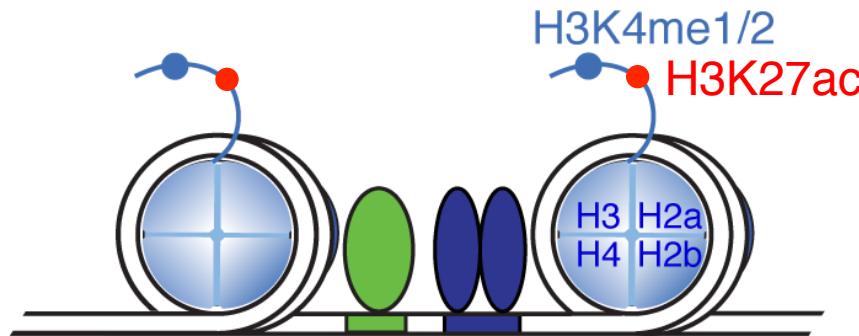
Most GWAS disease/trait-associated SNPs reside in non-coding regions of the genome



Enhancer/promoter interactions establish cell-specific and signal-dependent gene expression

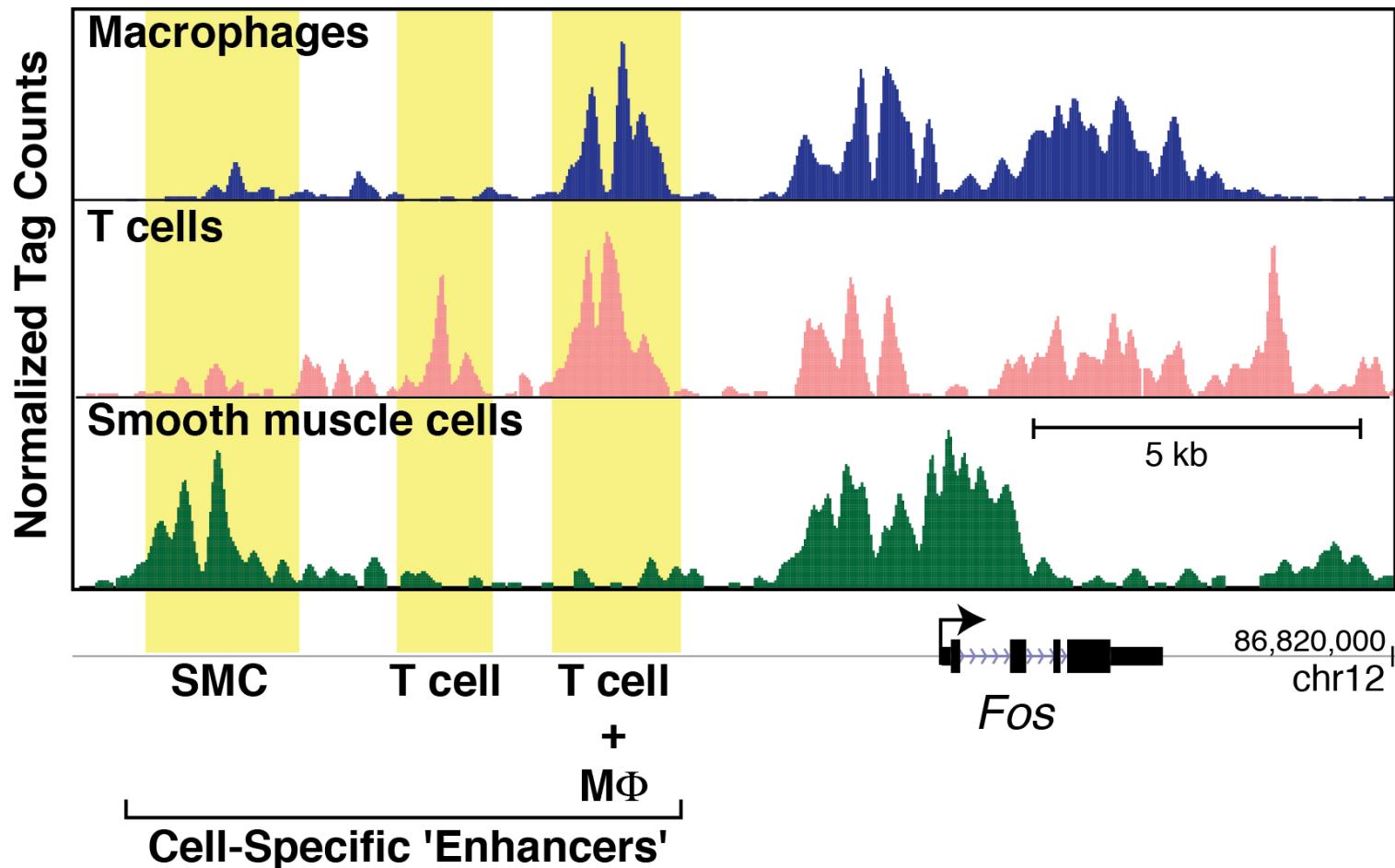


Enhancers are major determinants of cell-specific and signal-dependent gene expression



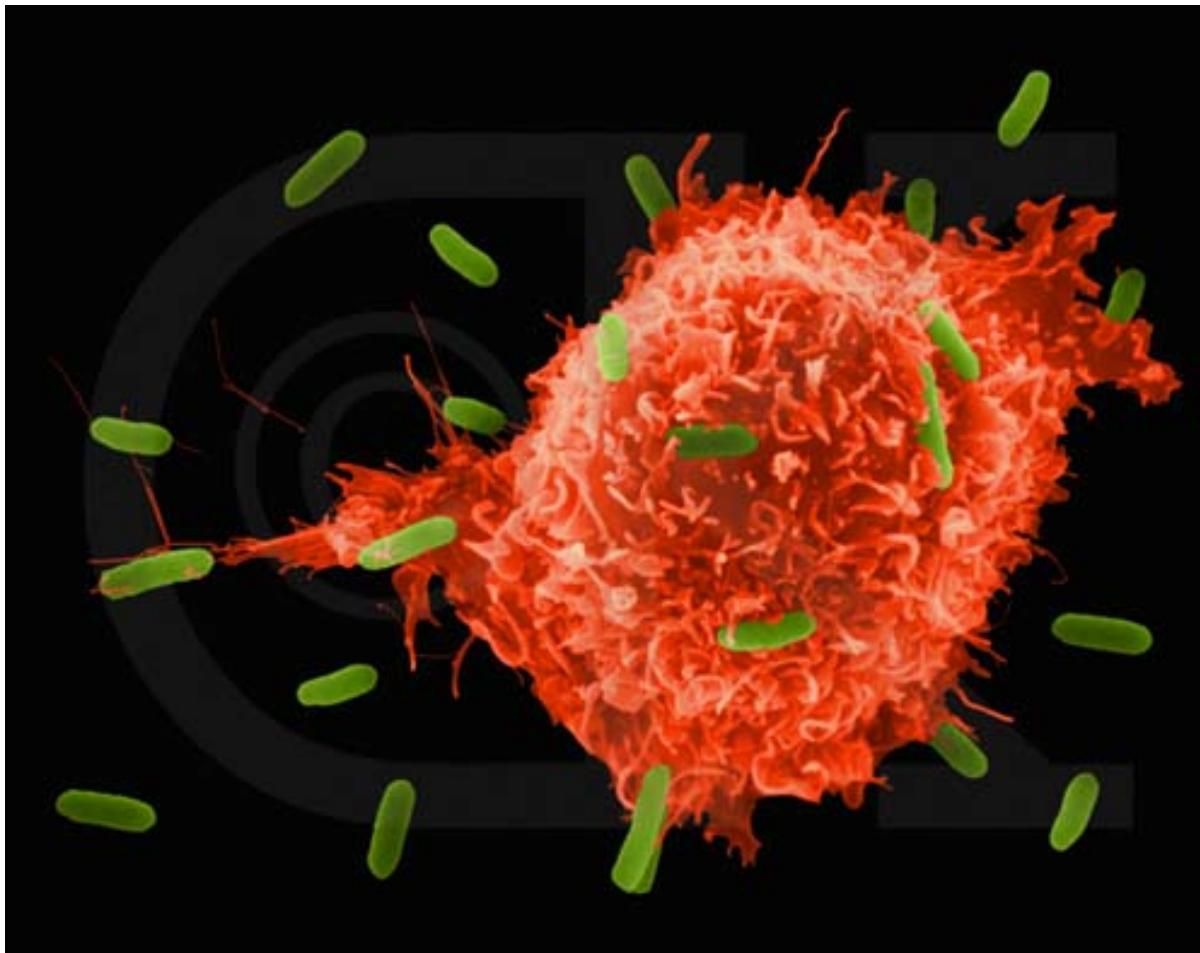
- Act at a distance from promoters to ‘Enhance’ gene transcription
- Function determined by sequence-specific transcription factors
- Exhibit a distinct epigenetic signature; $H3K4me1/2 > H3K4me3, H3K27ac$
- ~ 1 million predicted in the human genome

Cell-specific selection of enhancers

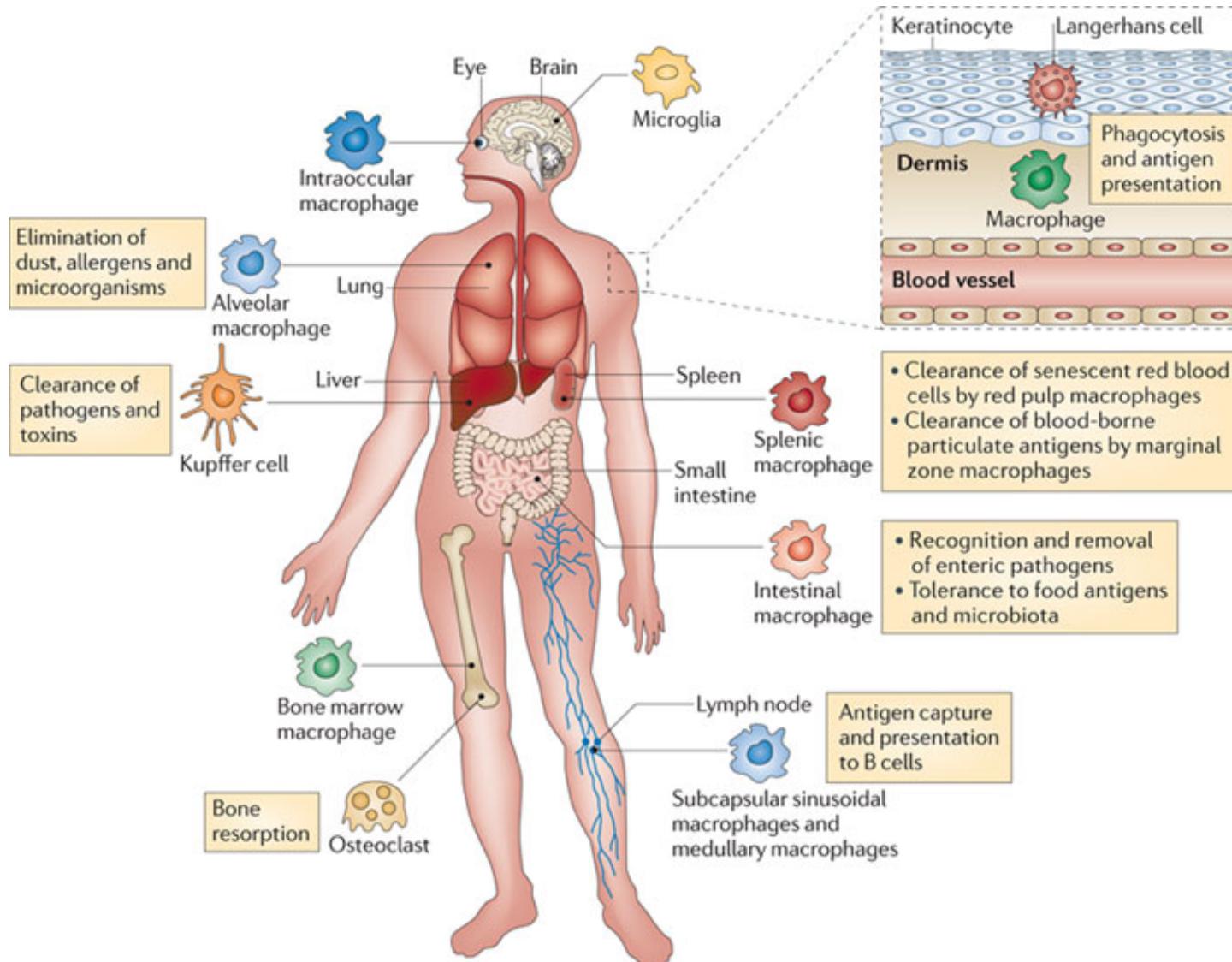


From ENCODE - Combined H3K4me3 and H3K4me1

Macrophages play essential roles in the response to infection and injury

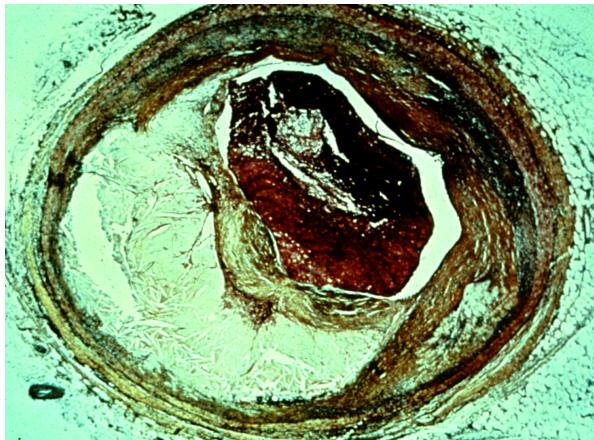


Specialized homeostatic functions of resident tissue macrophages

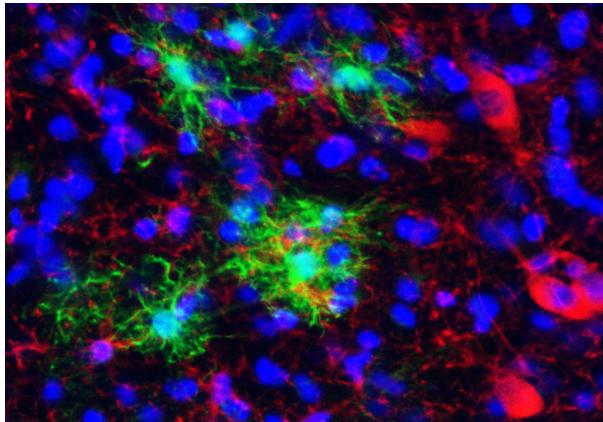


Roles of macrophages in human disease

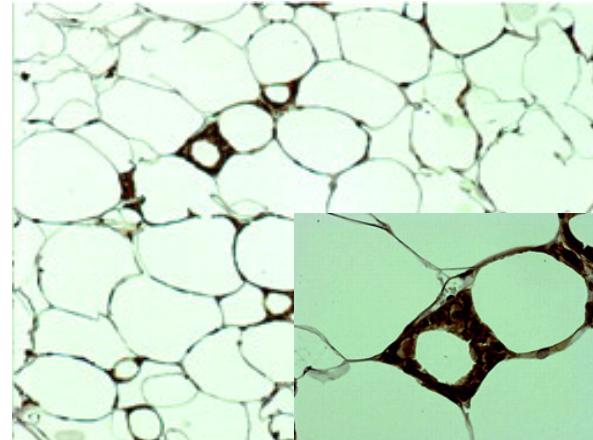
Macrophage foam cells
in atherosclerosis



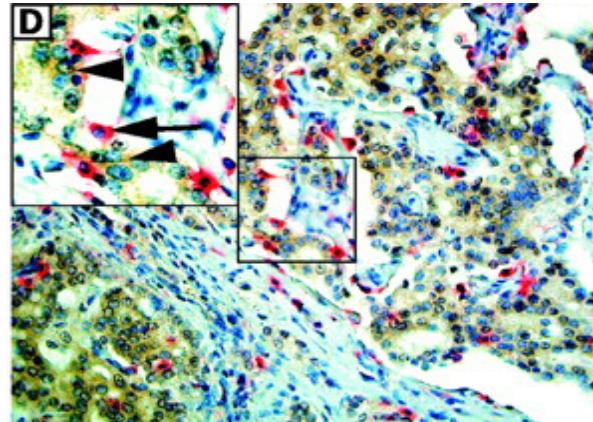
Activated microglia in
neurodegenerative disease



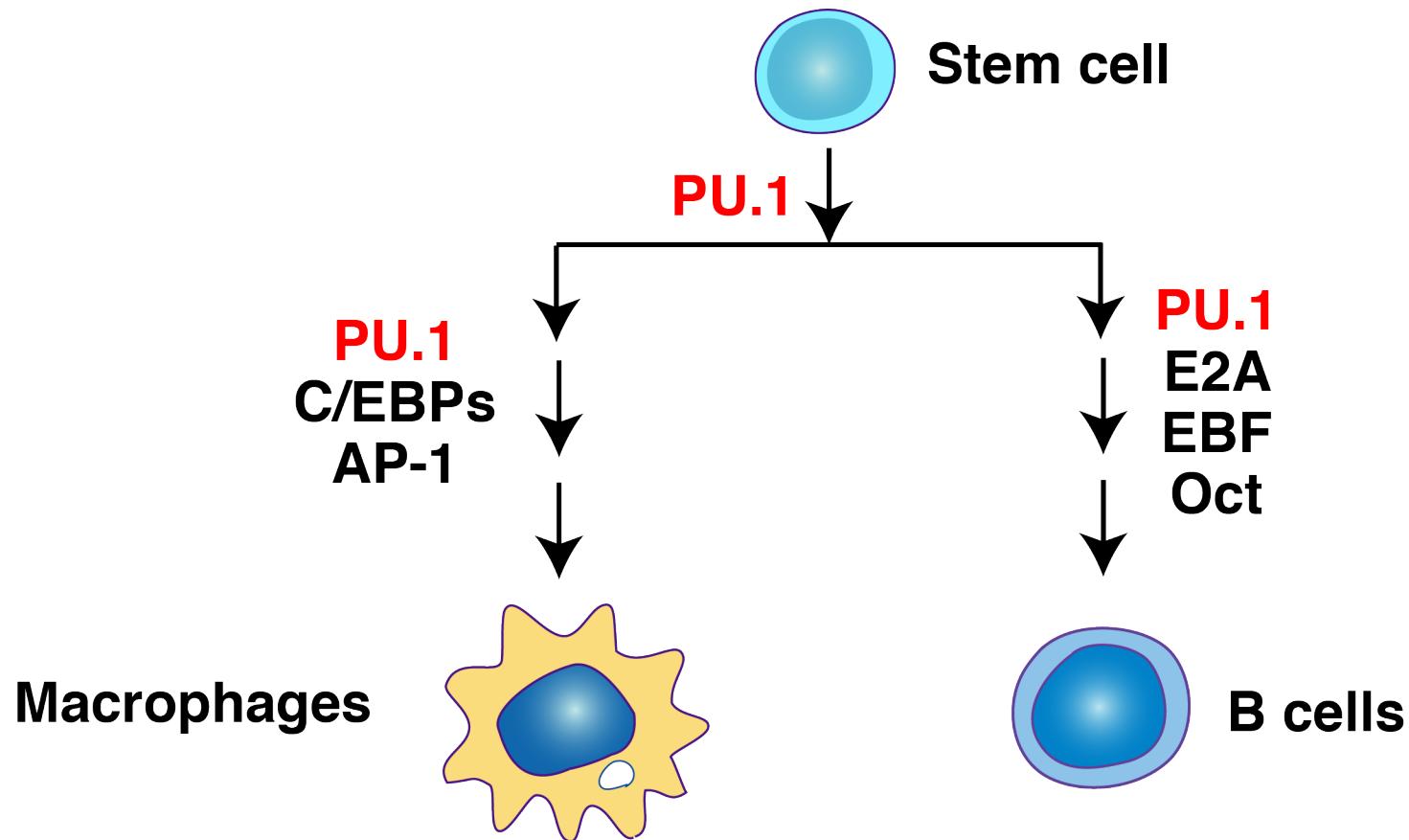
Adipose tissue macrophages
in insulin resistance



Tumor-associated macrophages
in cancer



PU.1 is required for both macrophage and B cell differentiation



Enriched motifs at macrophage and B-cell-specific PU.1 binding sites

Macrophage-specific PU.1 binding sites

 AAAGAGGAAAGTG	PU.1
 ATGACTCA	AP1
 TATTGCGCAA	C/EBP

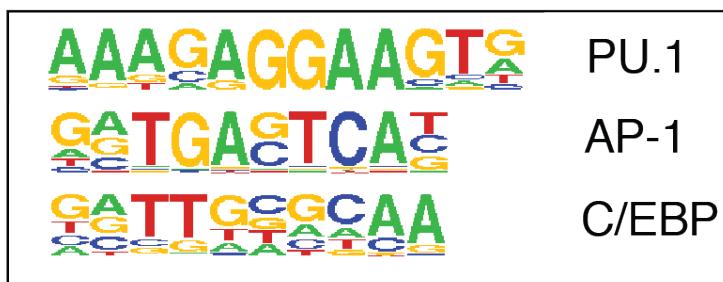
B-cell-specific PU.1 binding sites

 AAAGAGGAAAGTG	PU.1
 CCACACCTGC	E2A
 CCCCTGGGGAC	EBF
 GGGGATTCCCCC	NFkB
 TATGCAAAAT	OCT

PU.1 and C/EBP β occupy the majority of the enhancer-like regions in macrophages

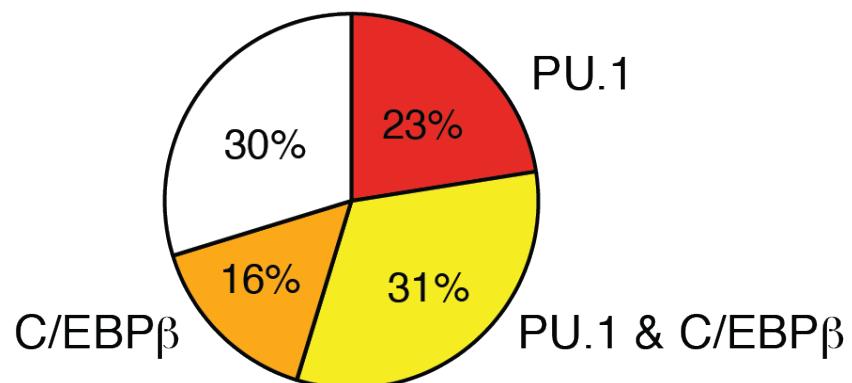
Macrophage H3K4me1 Peaks

Enriched motifs



Macrophage H3K4me1 Peaks

PU.1 and C/EBP β ChIP-Seq



Motifs for lineage determining TFs are enriched in H3K4me1 marked regions of the genome

B cells

	PU.1/ETS 42.9% (17.0%)
	PU.1:IRF 16.4% (6.2%)
	EBF 27.3% (17.1%)
	Runx 19.8% (12.9%)
	E2A 43.9% (31.9%)
	OCT

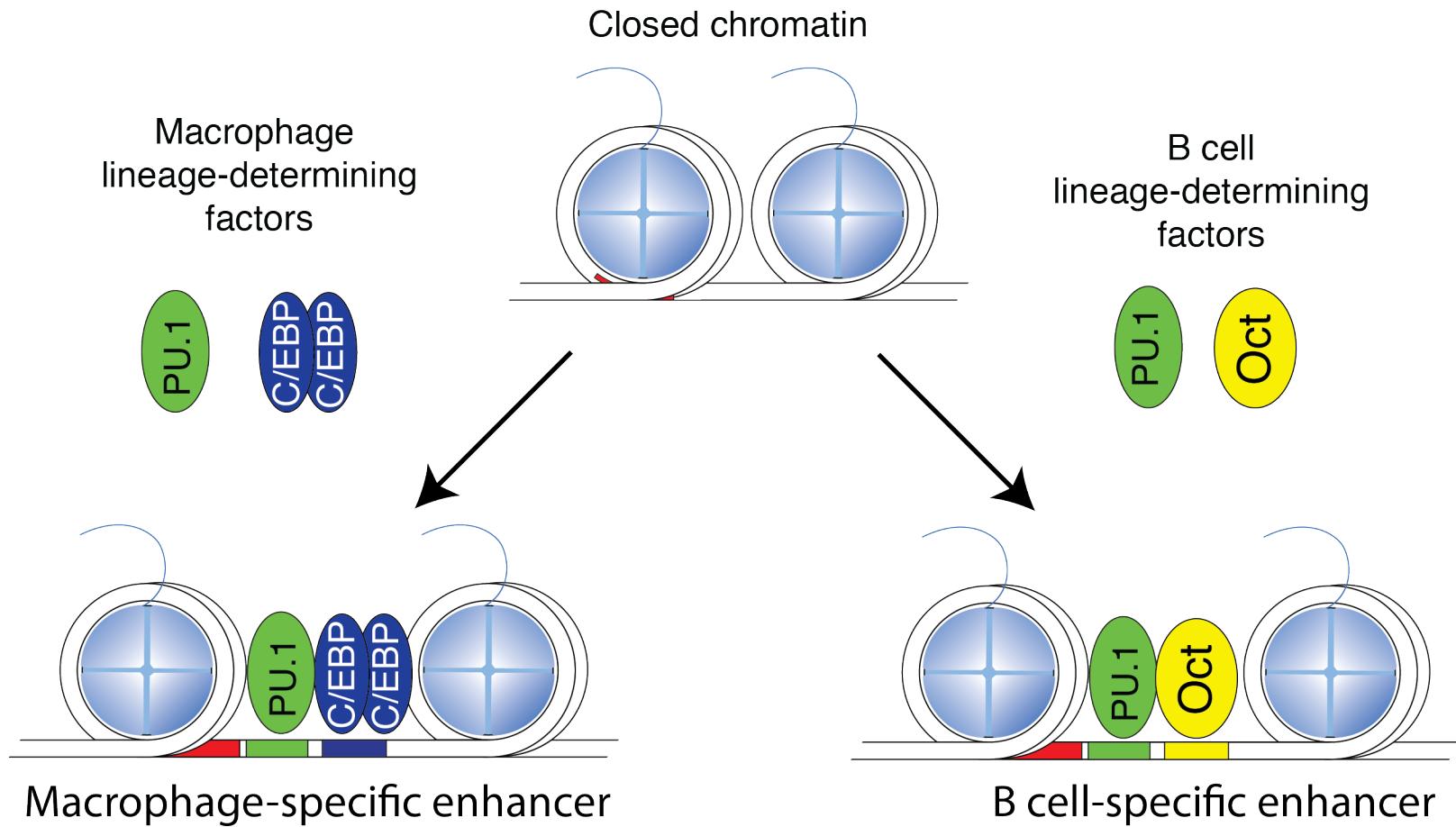
Embryonic stem cells

	KLF 42.7% (23.6%)
	OCT 22.0% (16.6%)
	SOX 53.8% (45.0%)
	Esrrb 29.4% (19.6%)
	OCT:SOX 15.4% (8.5%)

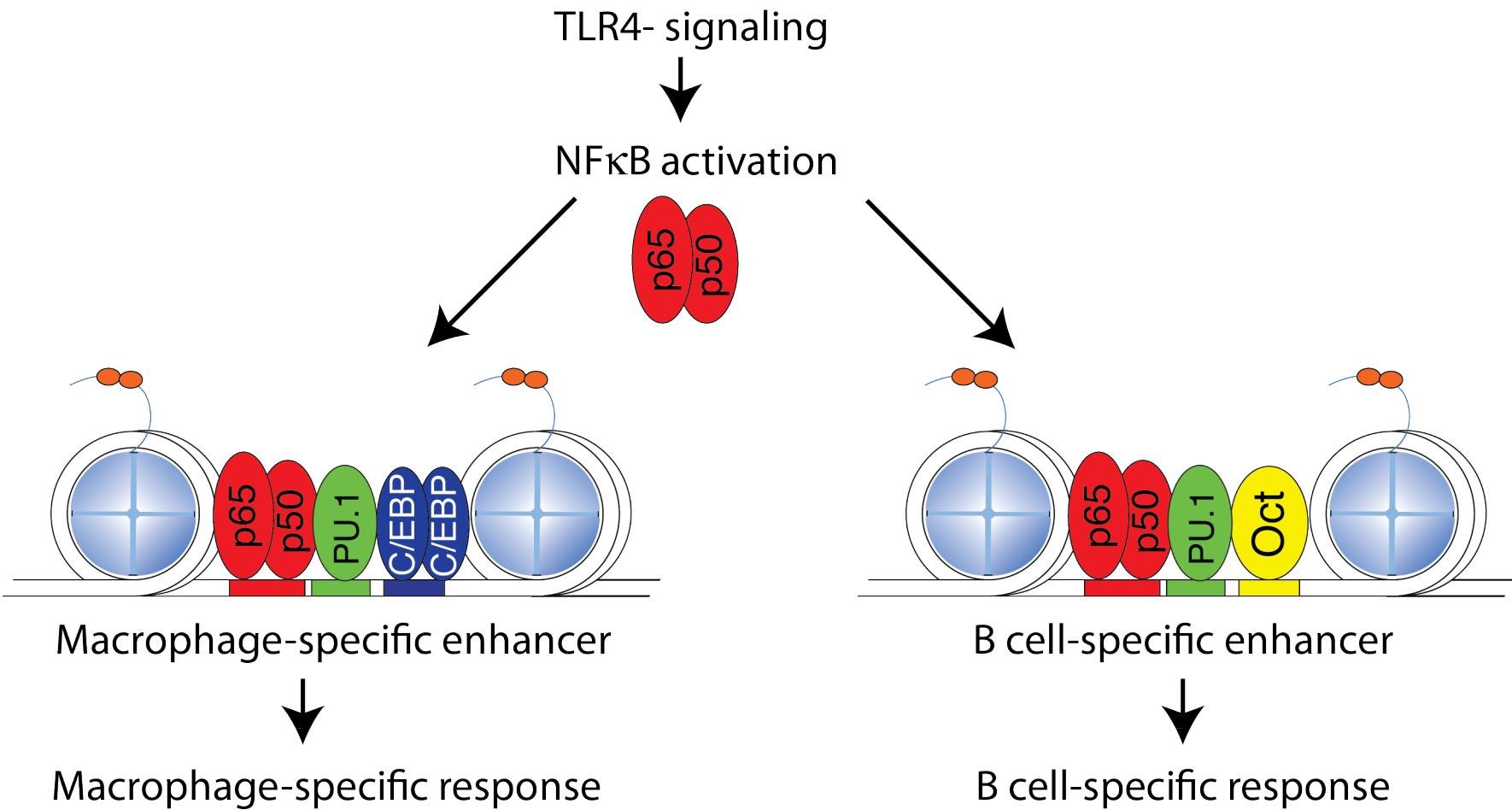
Liver

	HNF4 (DR1) 35.5% (16.8%)
	HNF1 10.4% (4.6%)
	CTCF 4.7% (2.5%)
	NF1-half site 43.2% (31.0%)
	HNF6 25.6% (15.4%)
	C/EBP 33.3% (25.8%)
	FOX (HNF3)

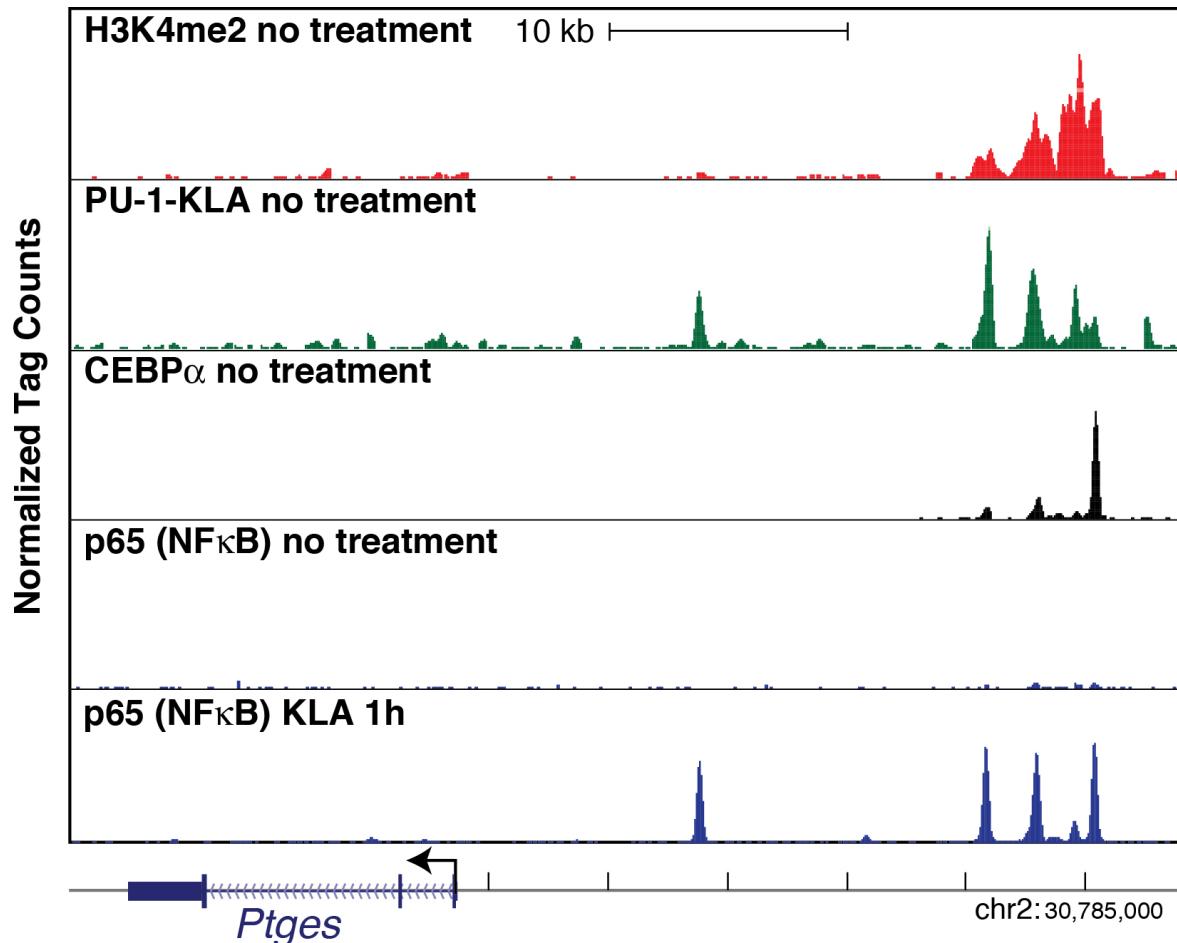
Selecting macrophage and B cell-specific enhancers



Selecting macrophage and B cell-specific enhancers

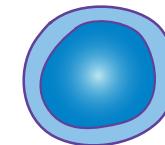
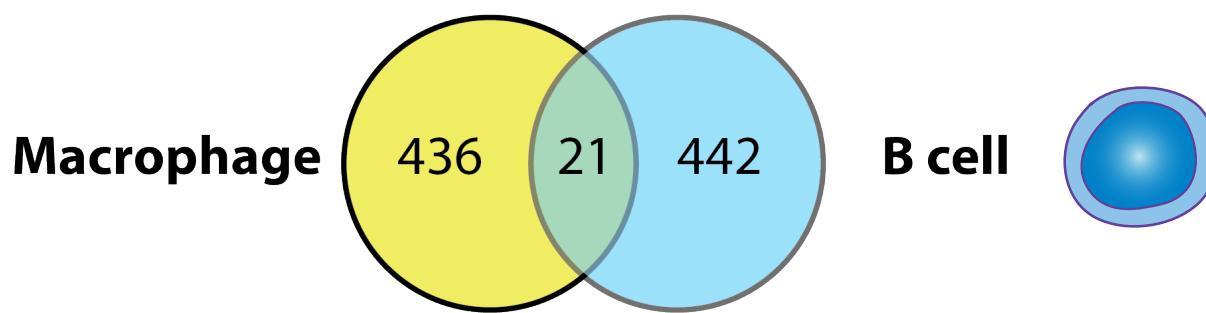
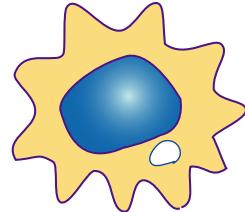


Binding of p65 (NF κ B) to poised enhancer like regions near *Ptges*



Transcriptional responses to the same signal can be cell type-specific

mRNAs increased > 4-fold 1h after TLR4 ligation



GO term	Macrophage p value	B cell p value
Immune response	1e-36	> 0.05
Chemotaxis	1e-12	> 0.05
Protein folding	> 0.05	1e-10
RNA processing	> 0.05	1e-9
DNA replication	> 0.05	8e-3

A collaborative – hierarchical model for enhancer selection and activation

Cell Fate-determining factors (i.e, lineage-determining TFs)

- Expressed in cell-restricted combinations
- Collaborate with each other and additional factors to bind to DNA and initiate nucleosome remodeling
- Establish cellular identity and exhibit reprogramming potential



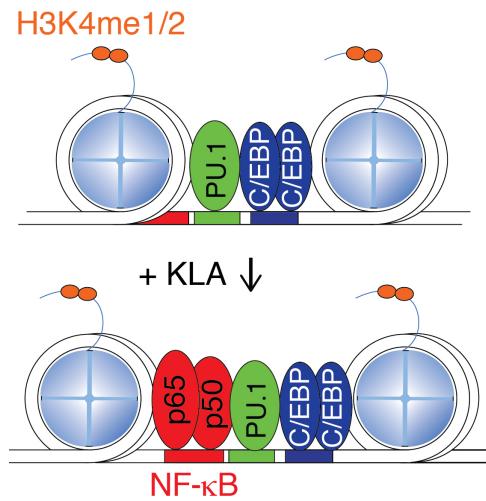
Cell State-determining factors (i.e., Signal-dependent TFs)

- Broadly expressed
- Primarily localize to pre-existing ‘primed’-enhancers
- Confer responsiveness to internal and external signals

Effects of natural genetic variation support a collaborative/hierarchical model



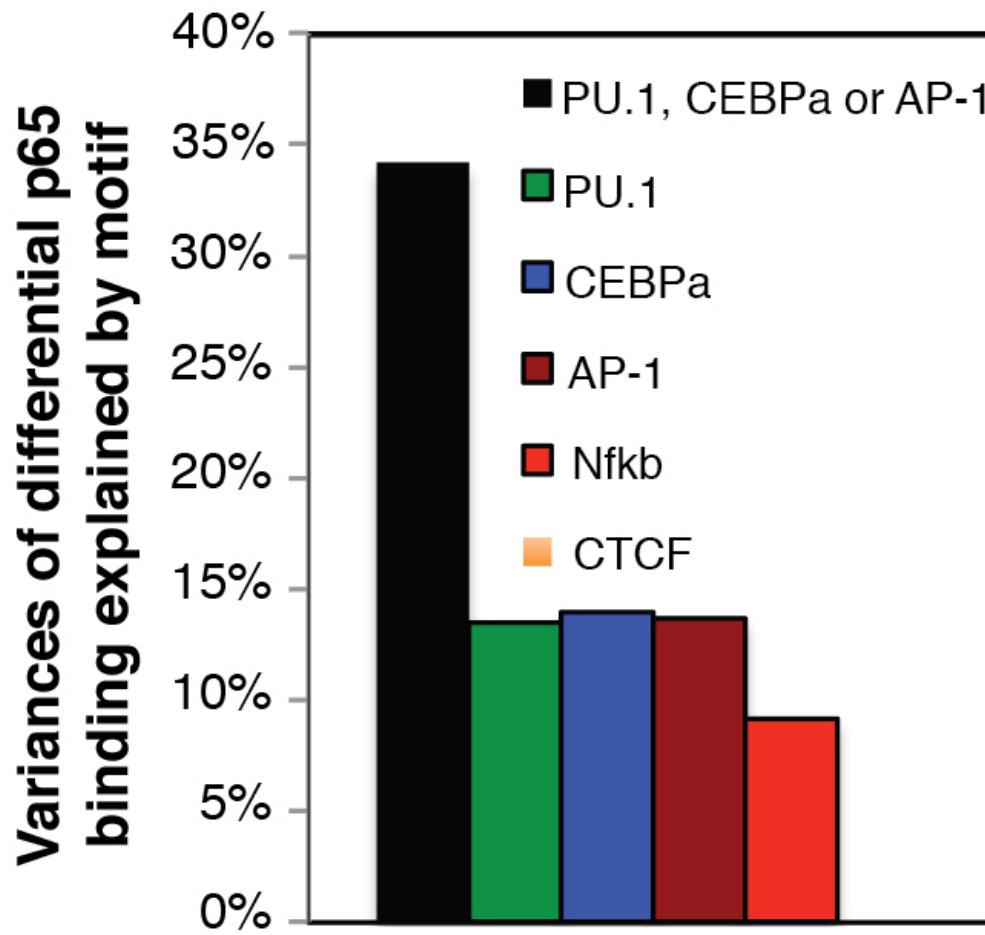
C57BL/6



BALBc ~4m SNPs

- Mutations in PU.1 motifs reduce nearby binding of C/EBP
- Mutations in C/EBP motifs reduce nearby binding of PU.1
- Mutations in NF κ B motifs rarely reduce binding of nearby PU.1 or C/EBP
- Mutations in PU.1 or C/EBP motifs frequently reduce binding of NF κ B

Effect of motif mutations on p65 binding



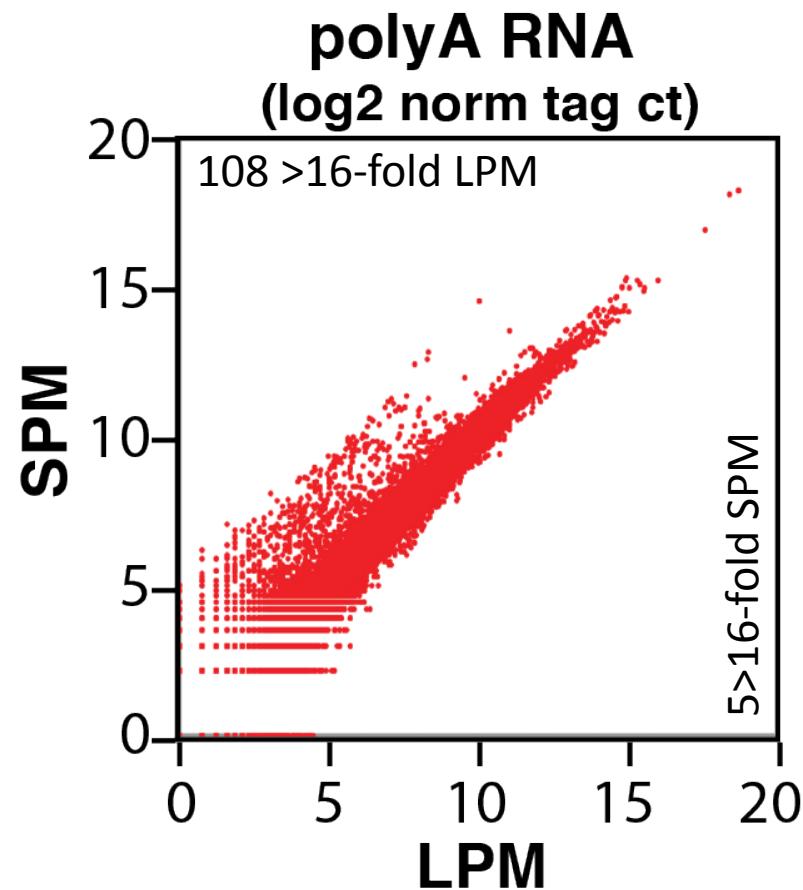
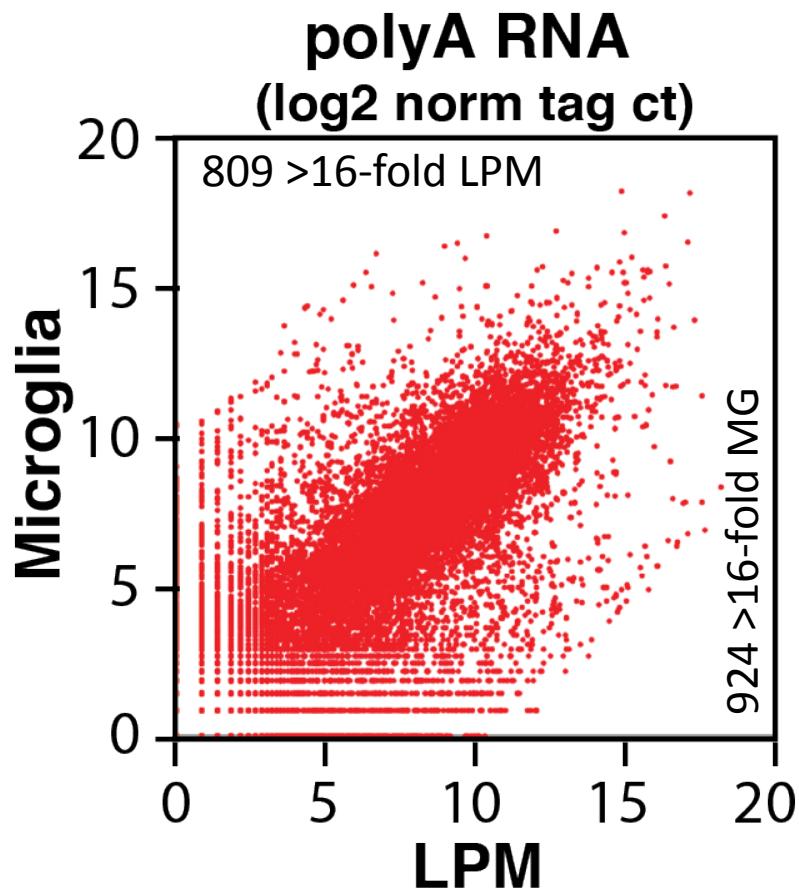
A collaborative – hierarchical model for enhancer selection and activation

- Supported by gain and loss of function experiments and effects of mutations in transcription factor binding sites (e.g., Heinz et al., Mol Cell, 2010, Heinz et al., Nature 2013, Kaikkonen et al., Mol. Cell, 2013)
- Model is vastly oversimplified and has poor predictive power
- Fails to explain how new enhancers are selected
- Fails to account for functions of the majority of transcription factors expressed in macrophages
- Relevance to *in vivo* populations of macrophages is unclear

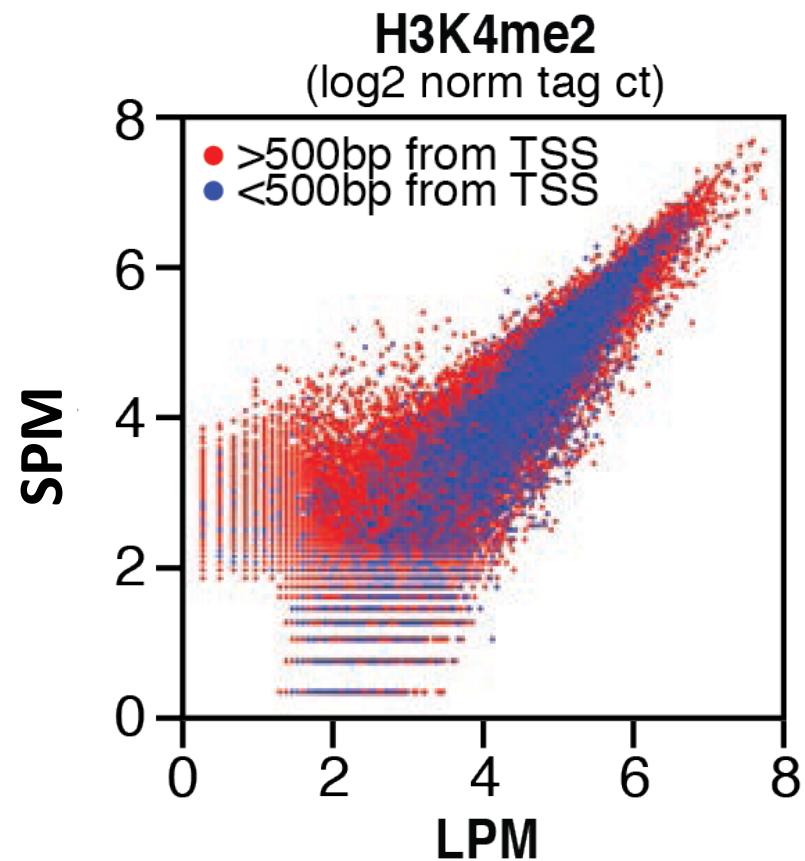
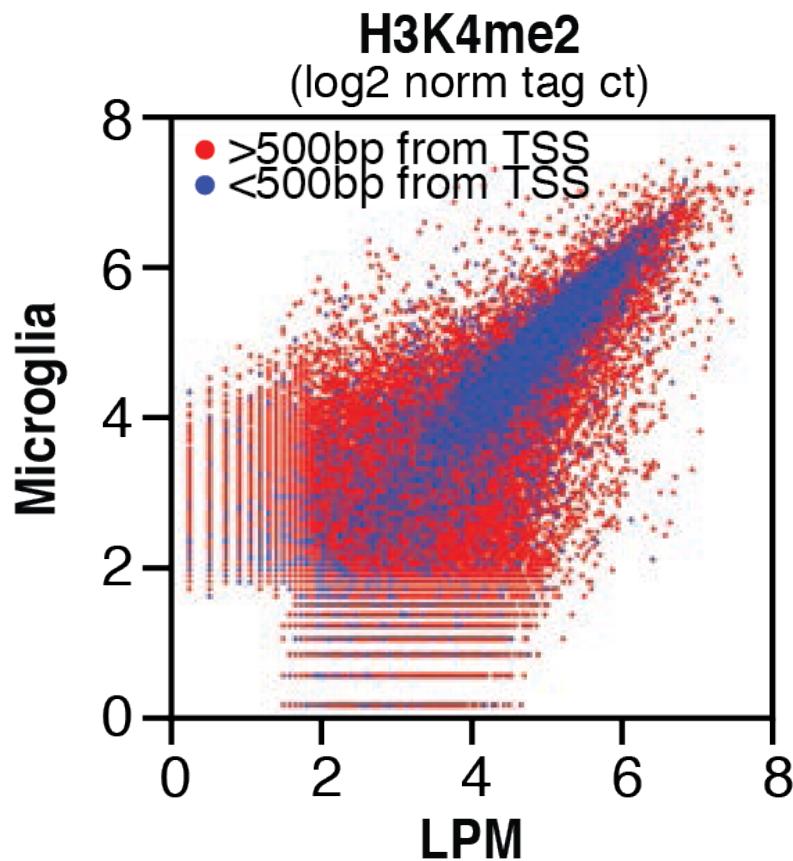
A comparison of tissue resident macrophage subsets

Cell type	Environment
Microglia (MG)	Brain/TGFβ
Large perit. MΦ (LPM - MHCII^{low})	Peritoneum
Small perit. MΦ (SPM - MHCII^{med})	RA

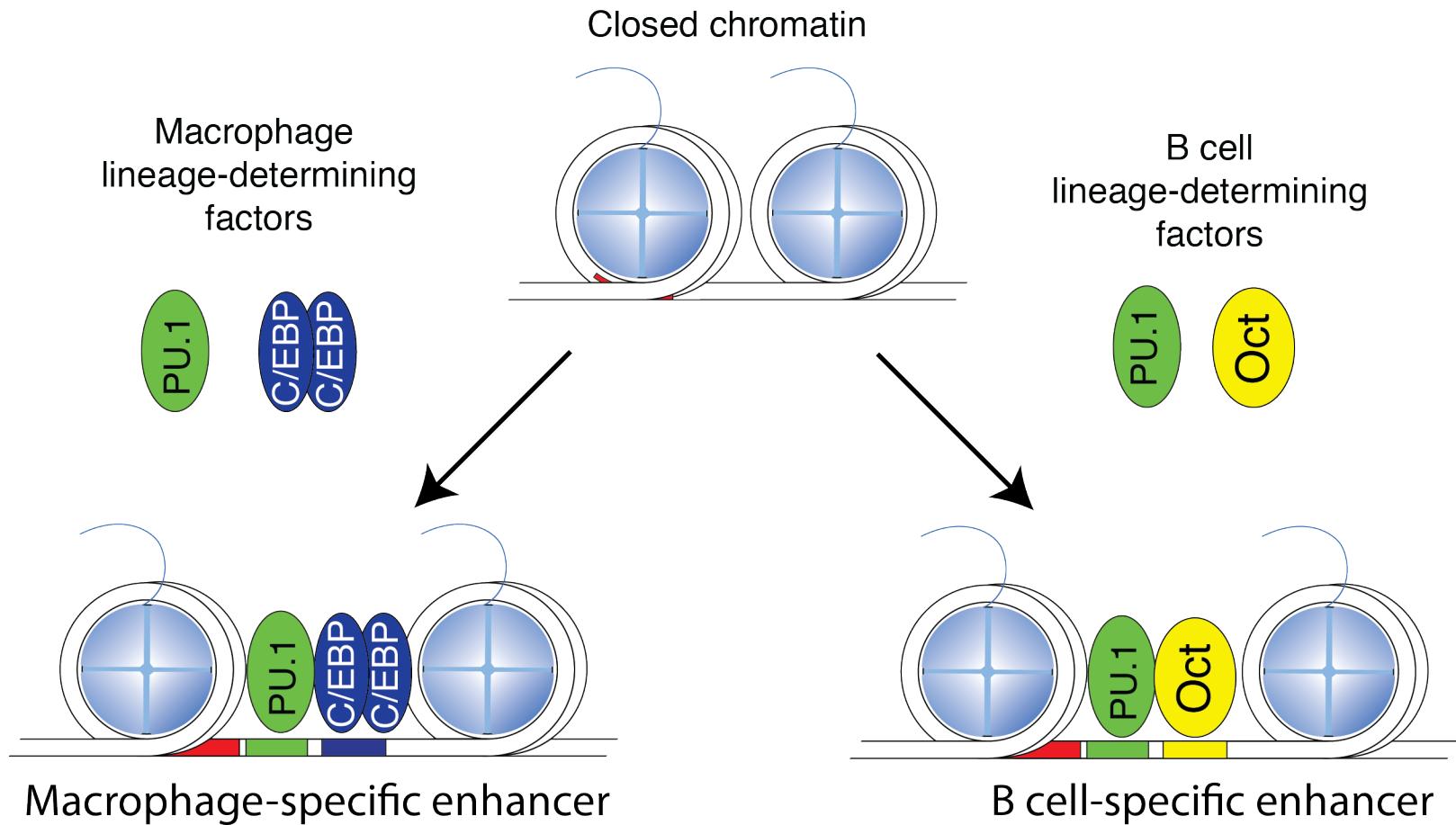
Macrophage transcriptomes can be highly divergent dependent on tissue of residence



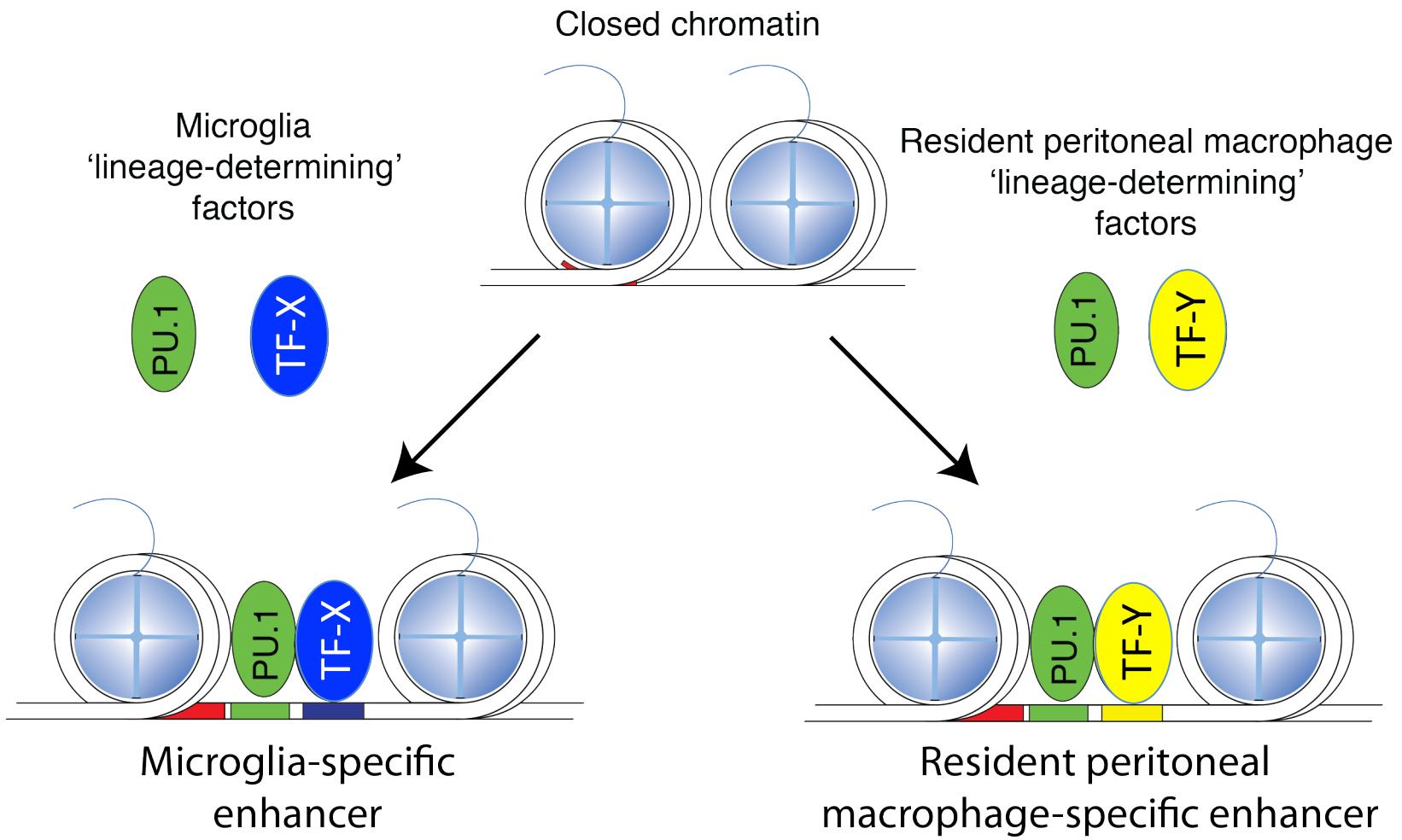
Macrophage enhancer landscapes differ depending on their source



Selecting macrophage and B cell-specific enhancers



Selecting microglia- and RPM-specific enhancers



What TFs collaborate with PU.1 to select microglia and RPM-specific enhancers?

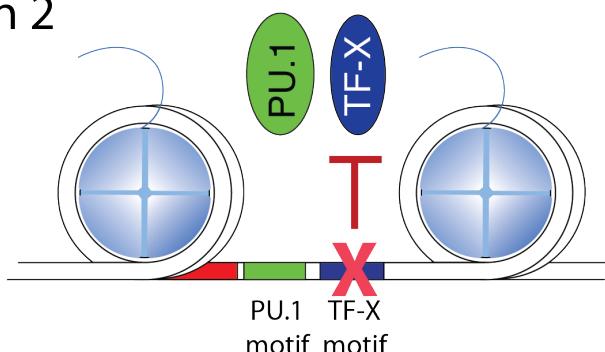
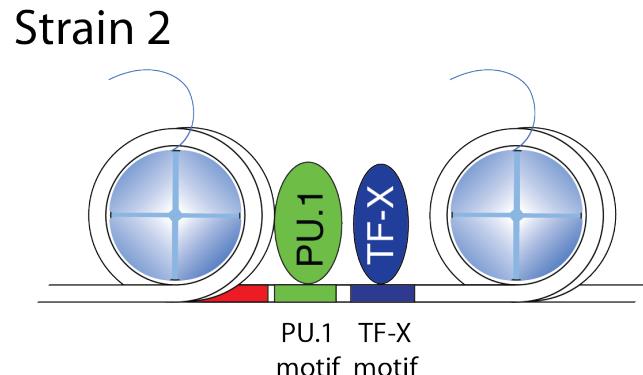
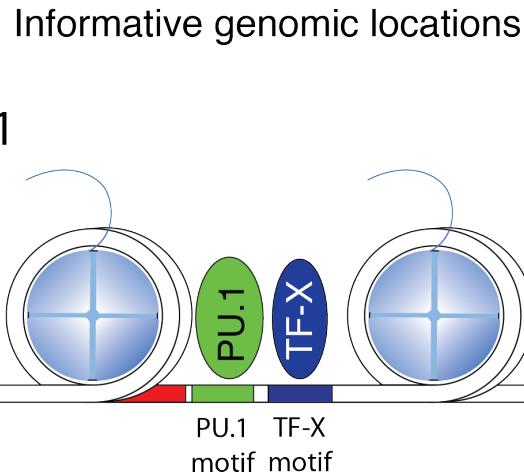
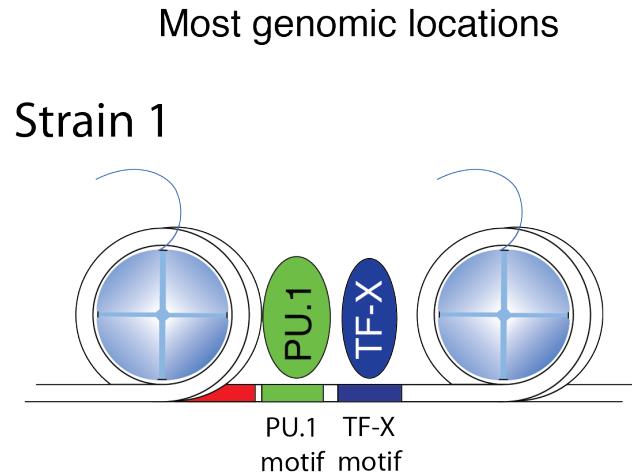
- Strategy :
 - Use the vast natural genetic variation provided by inbred strains of mice as an *in vivo* ‘mutagenesis screen’



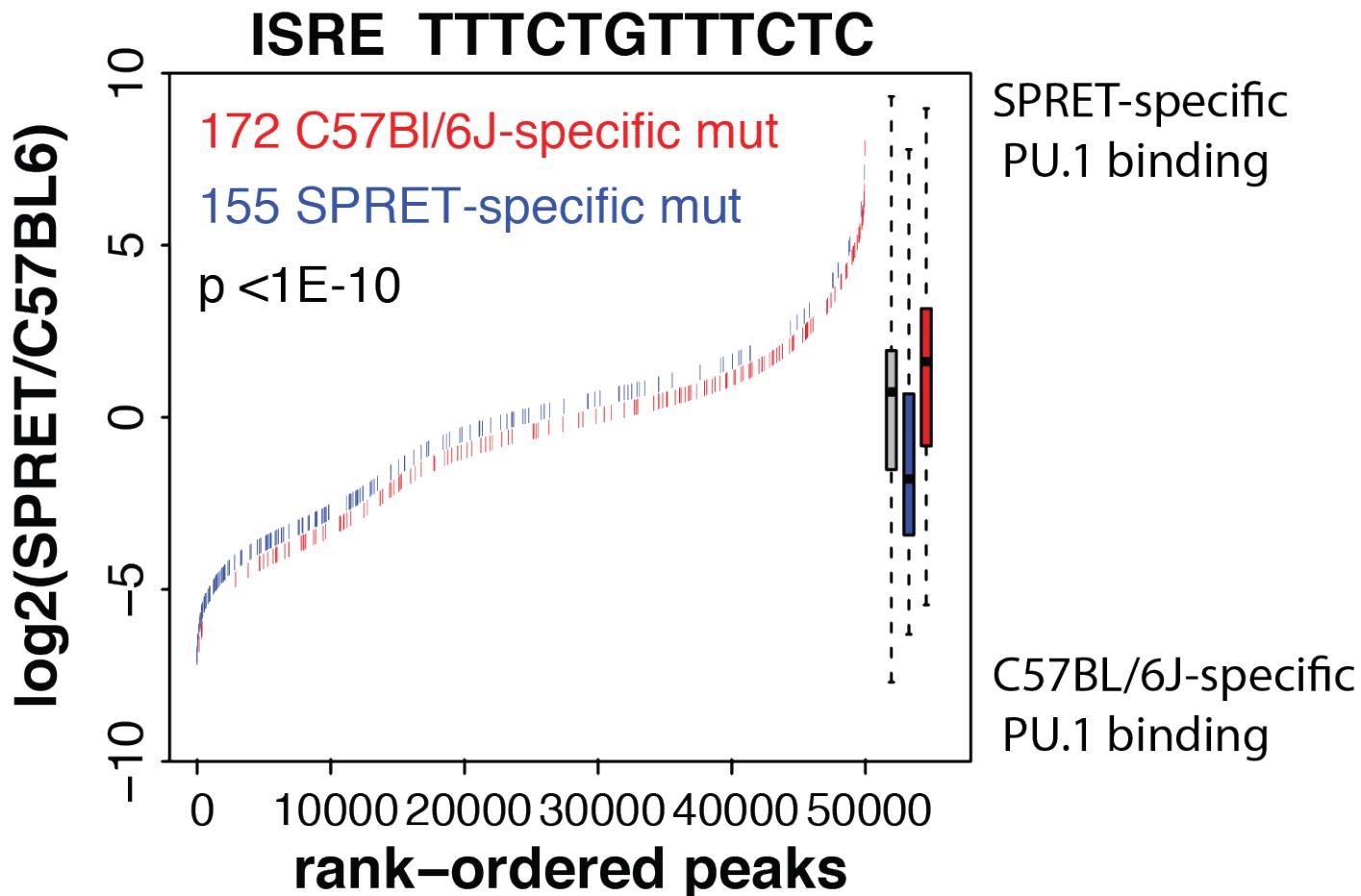
Inbred Strain	SNPs relative to C57BL/6J
BALB/cJ	~4 million
NOD/ShiLt/J	~5 million
SPRET/EiJ	~40 million

Use of natural genetic variation to identify motifs for collaborative TFs

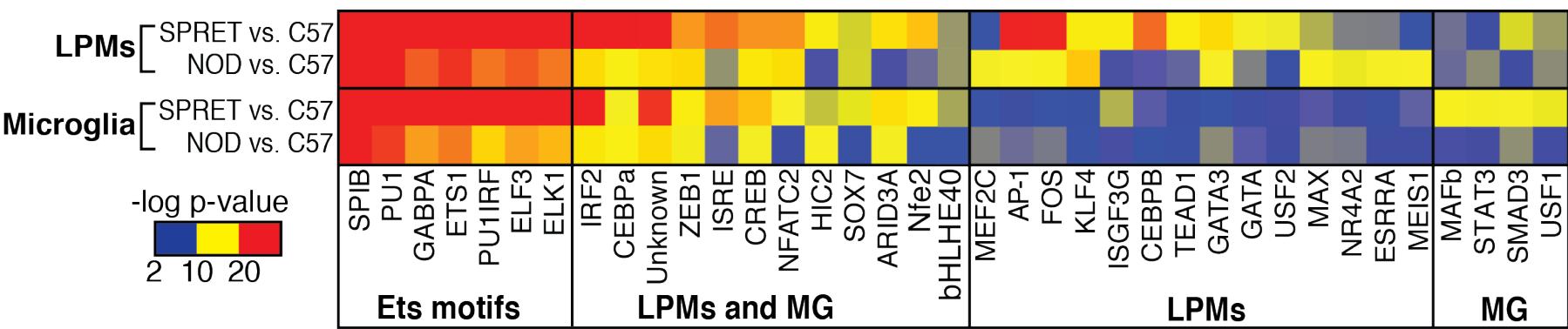
Transcription Factor X is a collaborative partner for PU.1 at a subset of PU.1 binding sites in the genome



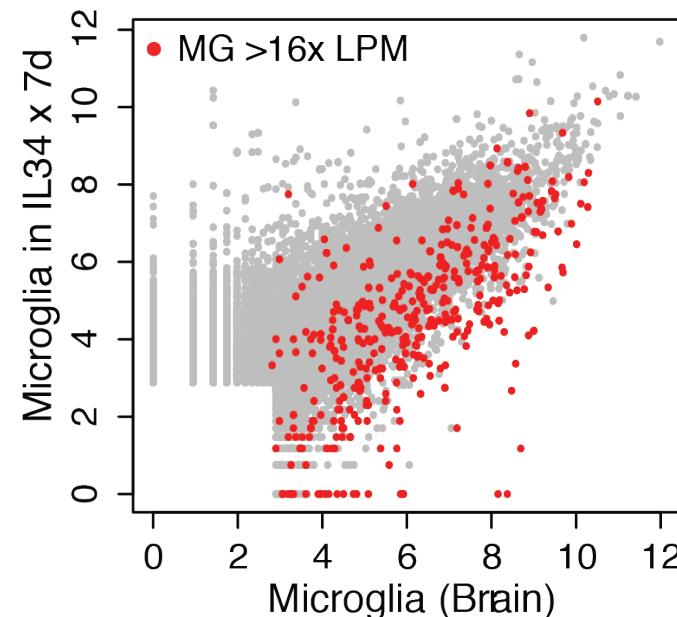
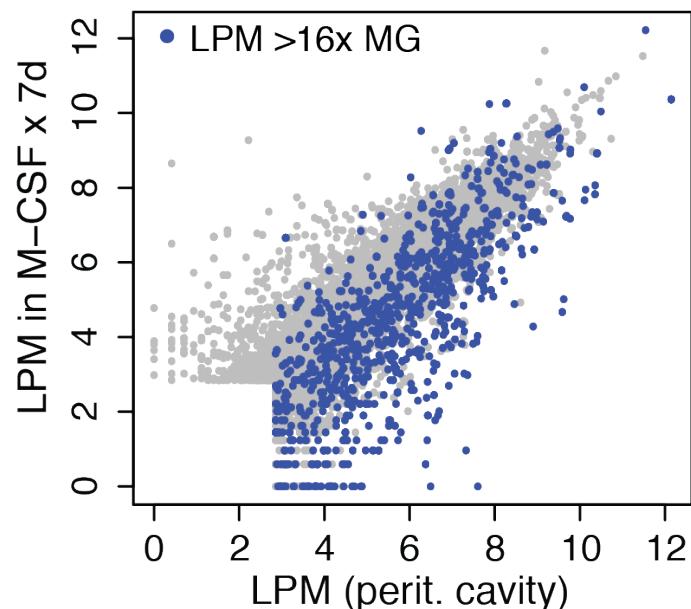
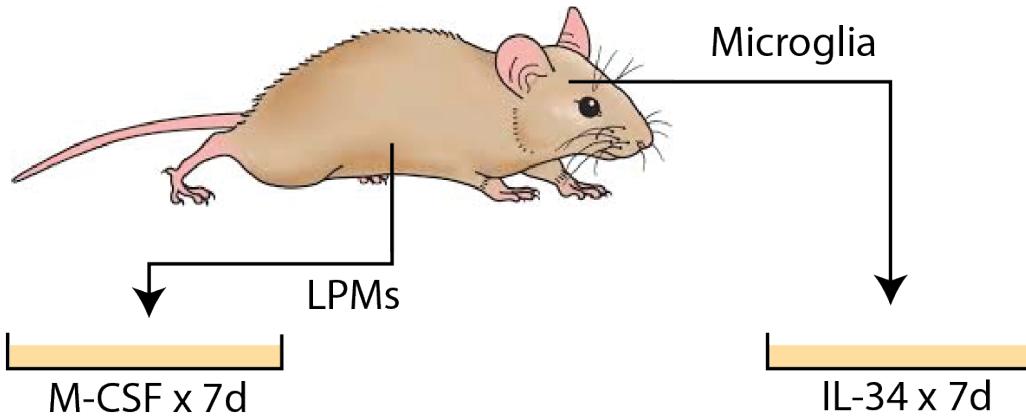
Mutations in nearby ISRE motifs are highly correlated with strain-specific PU.1 binding



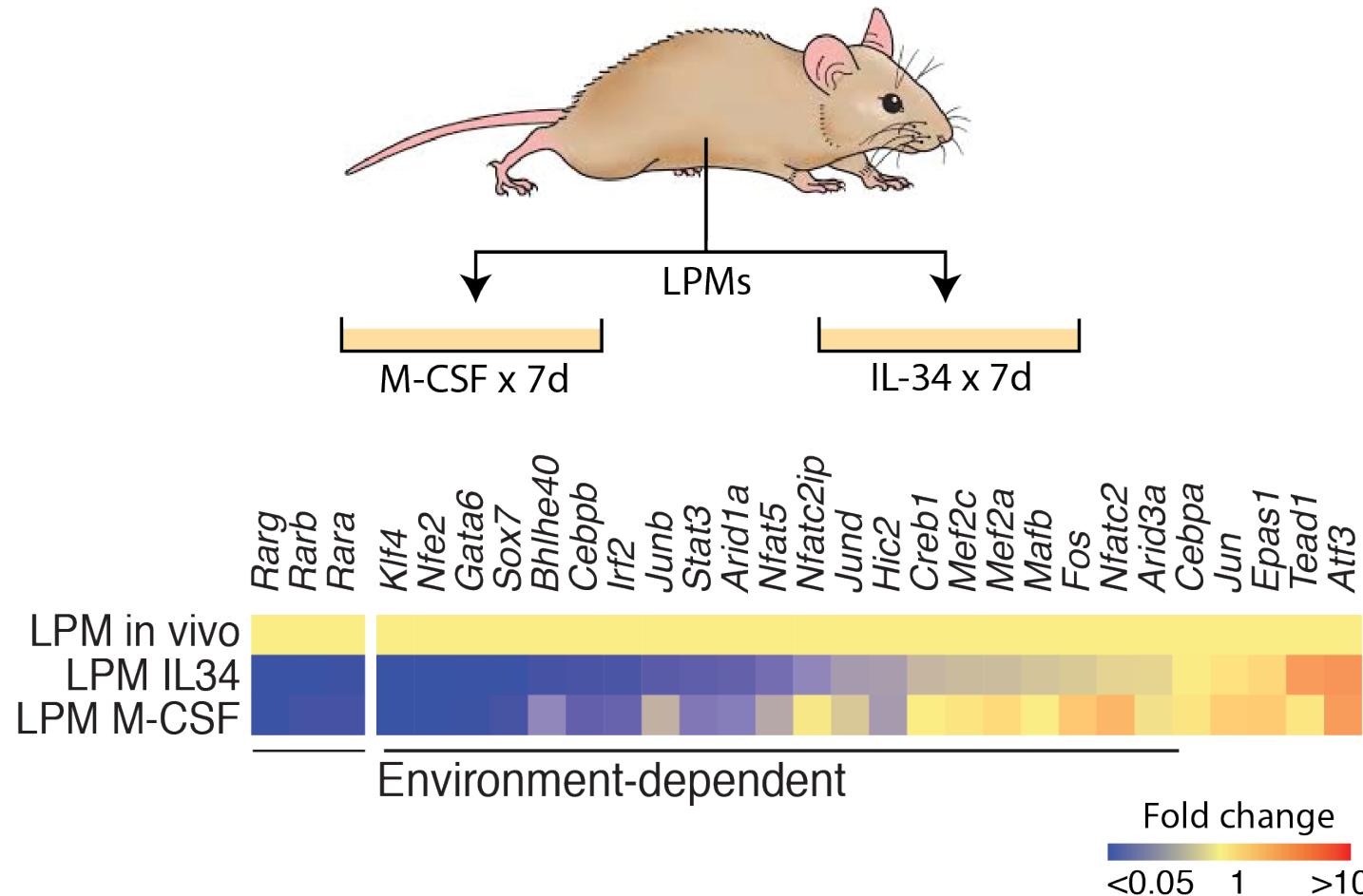
Discovery of TF motifs associated with common and subset-specific binding of PU.1



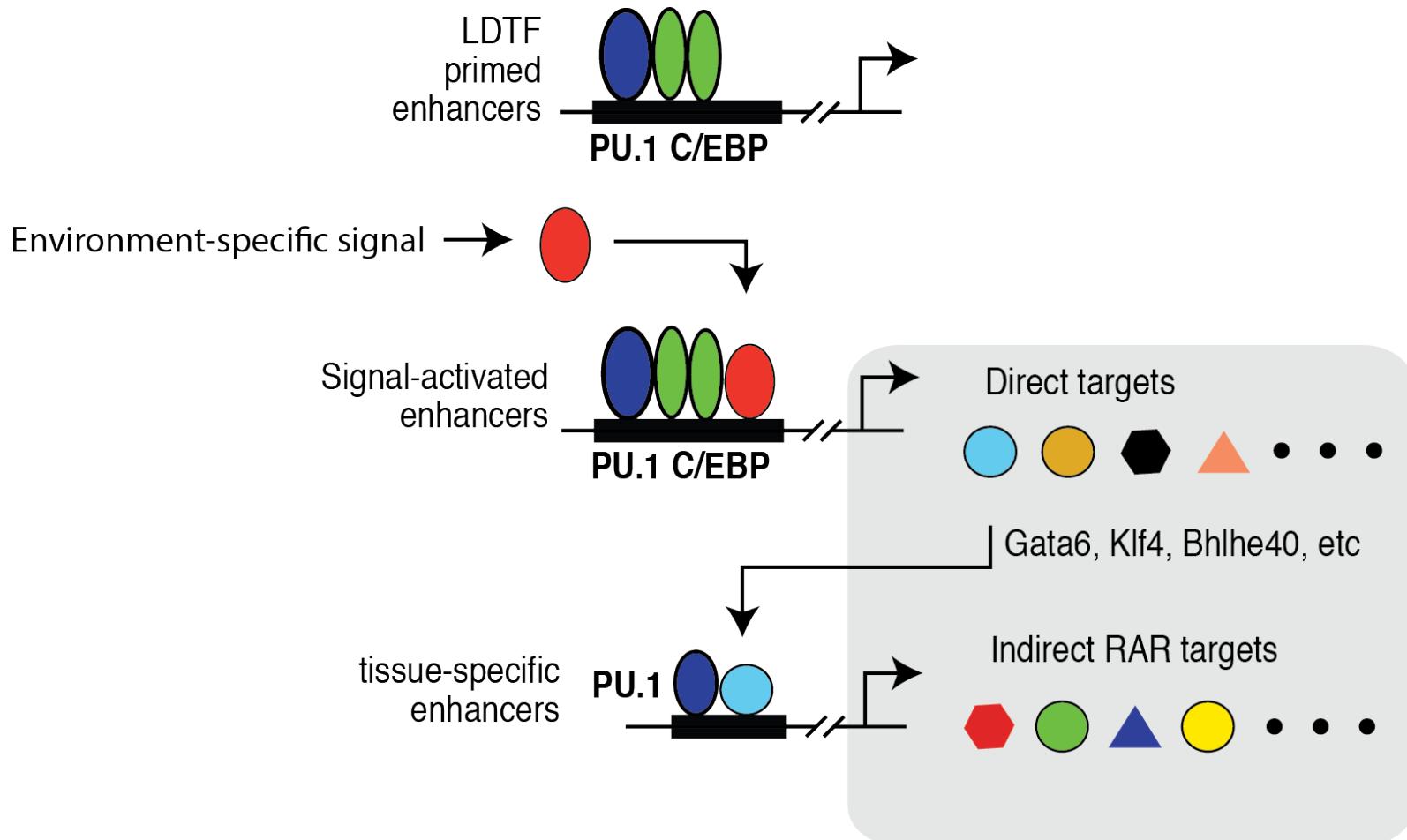
Macrophage identities require constant environmental signals



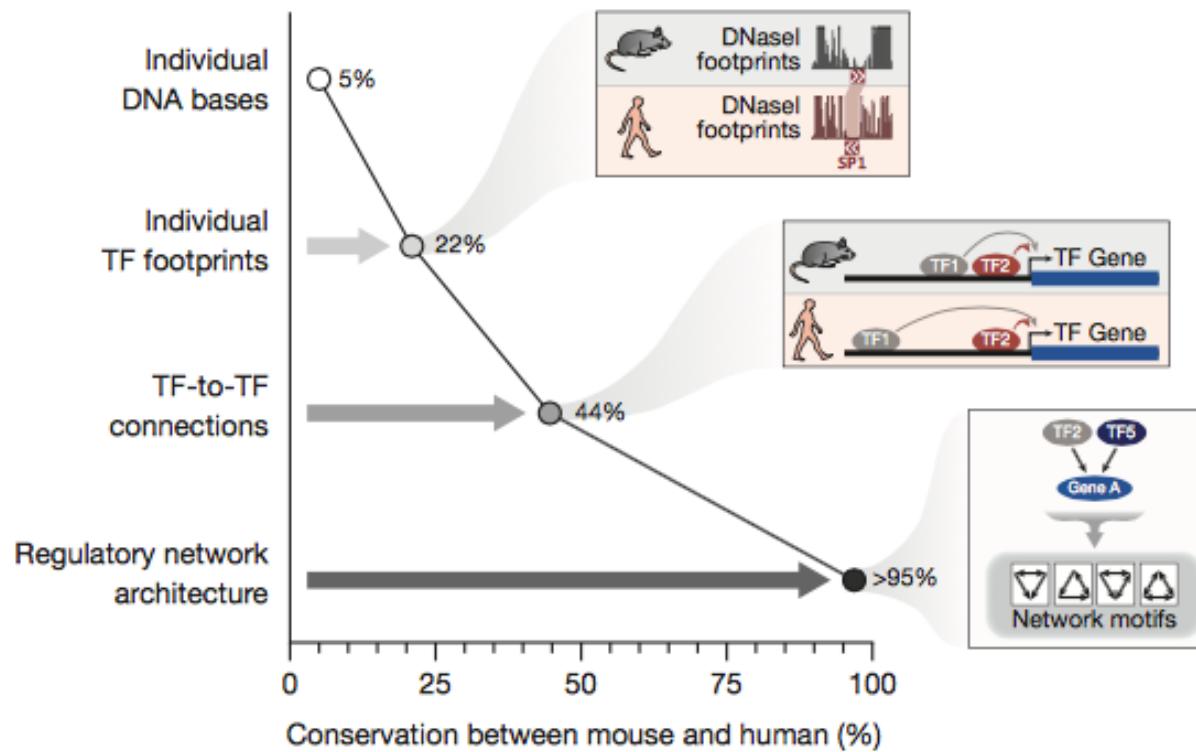
Environment controls expression of TFs that collaborate with PU.1



Environment drives selection and function of enhancers controlling macrophage identities



Conservation of trans-acting circuitry during mammalian regulatory evolution

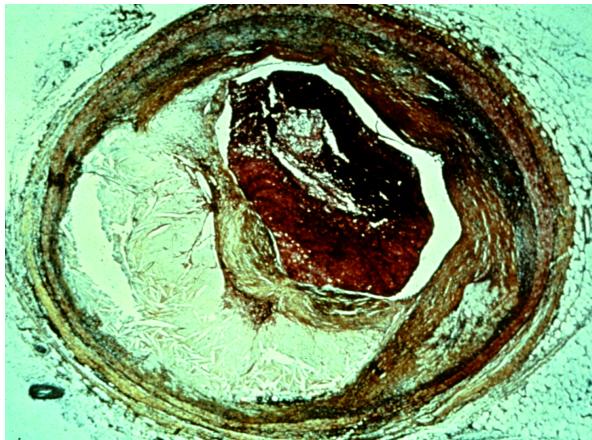


Some take home points

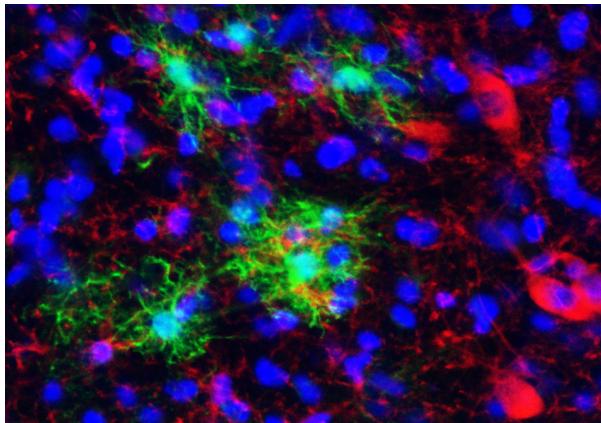
- Knowledge of the enhancer landscape of a cell reveals much about that cell's identity and regulatory potential
- Enhancer landscapes enable prediction of key lineage determining transcription factors and sites of action of signal-dependent factors
- Transcription factor binding maps inform analysis of genetic variation
- Natural genetic variation can be exploited to discover regulatory networks that drive cell-specific gene expression

Understanding and modifying roles of macrophages in human disease

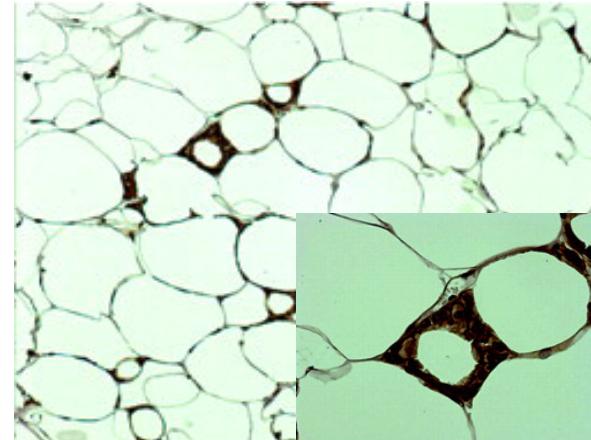
Macrophage foam cells
in atherosclerosis



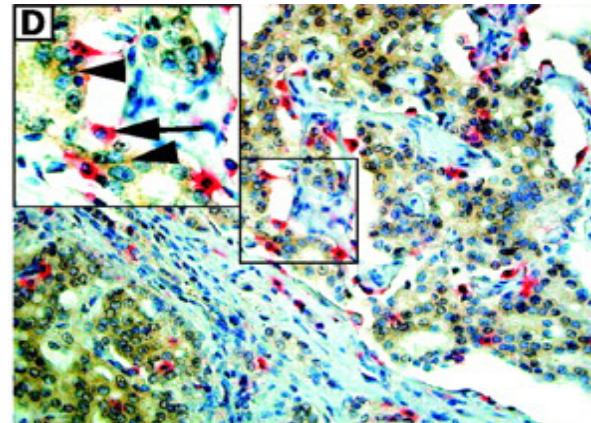
Activated microglia in
neurodegenerative disease



Adipose tissue macrophages
in insulin resistance



Tumor-associated macrophages
in cancer



Going forward

- Improve methodology to define regulatory networks in specific cell types within complex tissues
- Determine effects of cell-autonomous and non-autonomous disease mechanisms on enhancer selection and function
- Consideration of regulatory networks as complex phenotypes for therapeutic modulation

Thanks!

- David Gosselin
- Verena Link
- Casey Romanoski
- Sven Heinz
- Chris Benner
- Dawn Eichenfeld
- Ty Troutman
- Michael Lam



Kings College

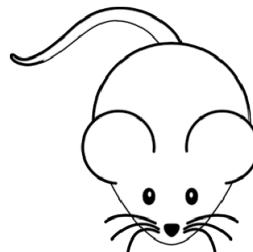
- Frederic Geissmann
- Hanna Gardiner

Exploiting natural genetic variation to understand enhancer selection and function

C57BL/6J

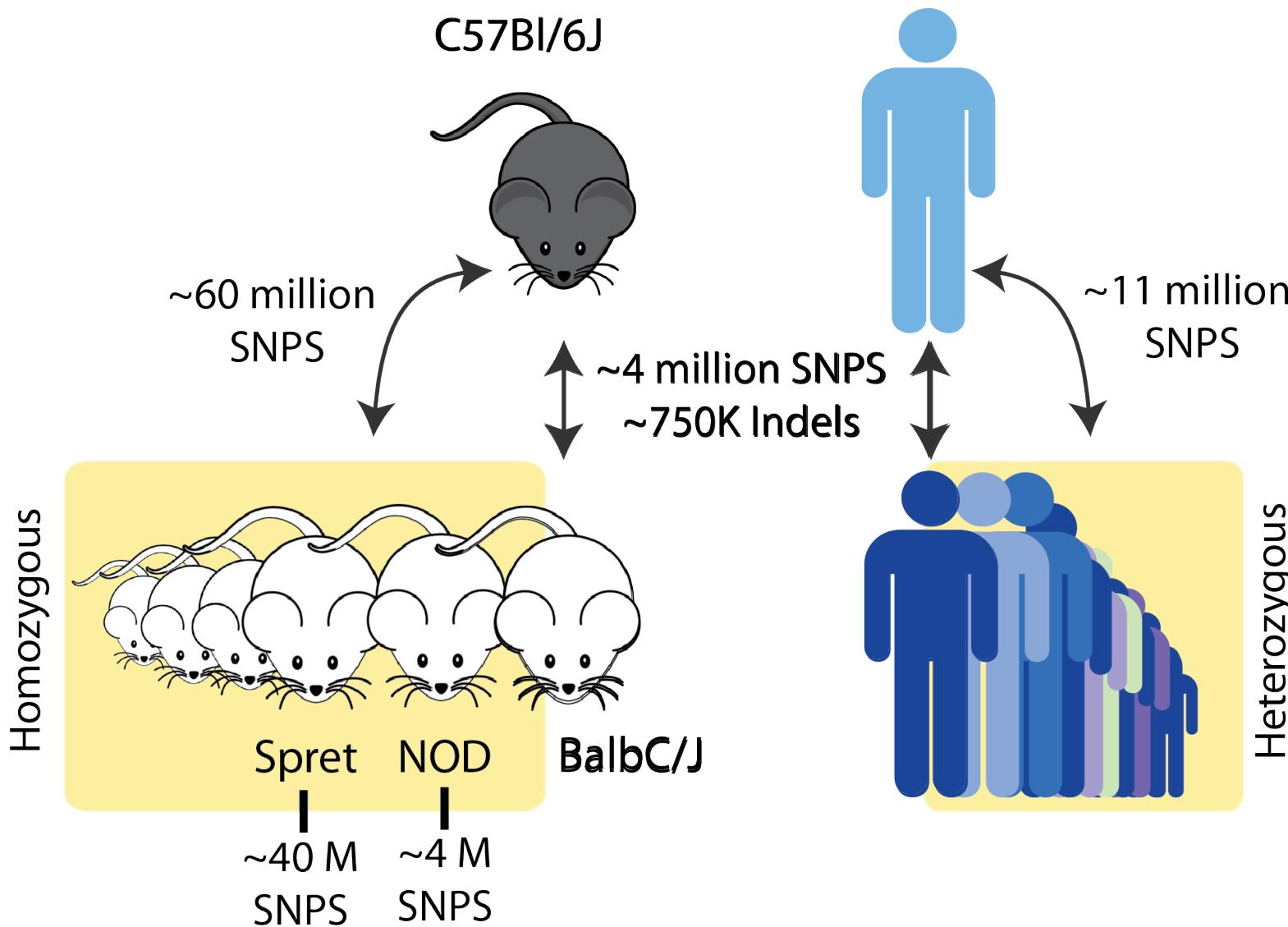


↔ ~4 million SNPs
~750K Indels



BalbC/J

Exploiting natural genetic variation to understand enhancer selection and function



Gene expression in the same cell type can vary among individuals

