

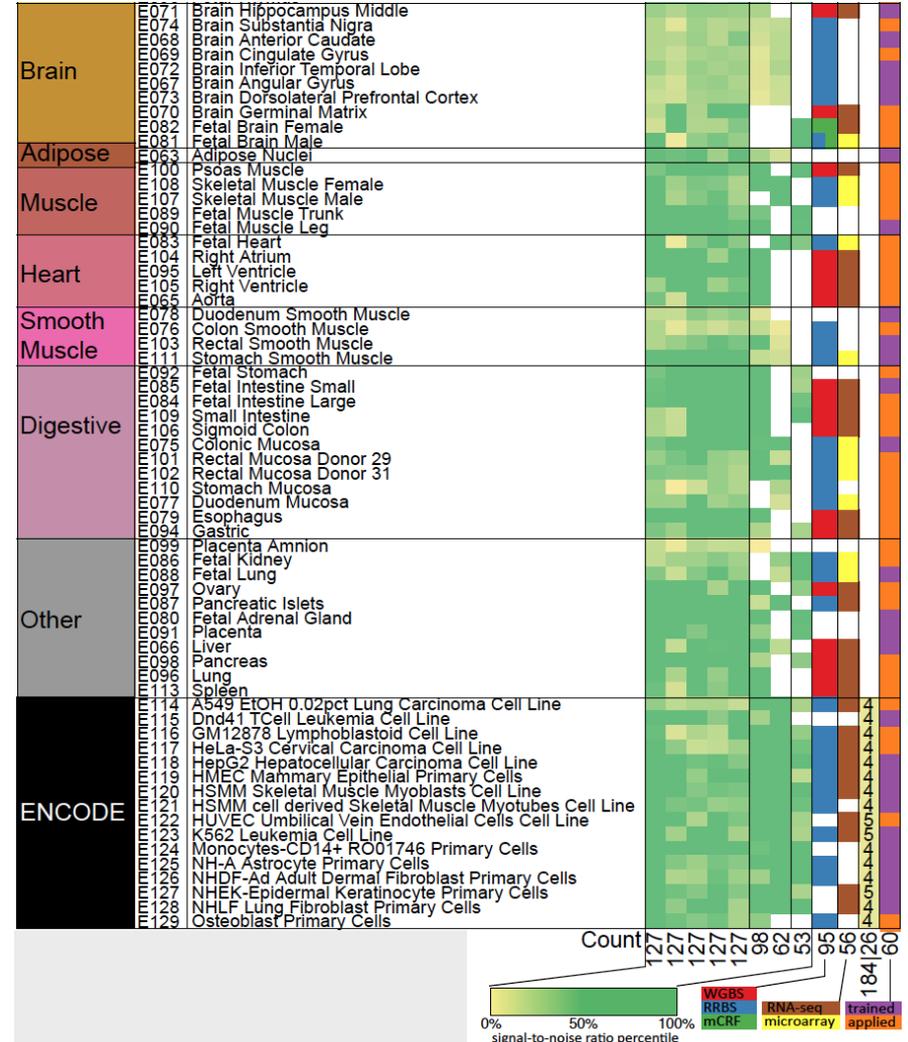
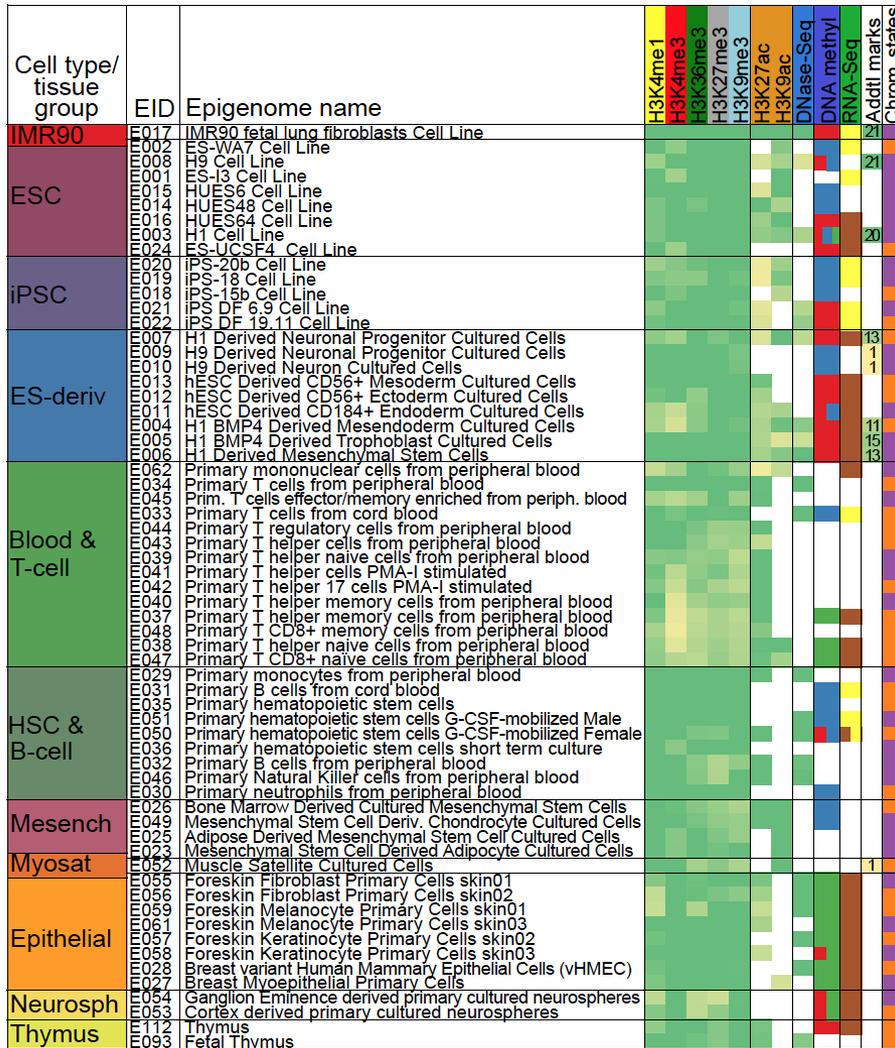
ENCODE3 chromatin state calls based on  
Roadmap Epigenomics ChromHMM models

WM20151112

# Goal: chromatin state calls for ENCODE3 data

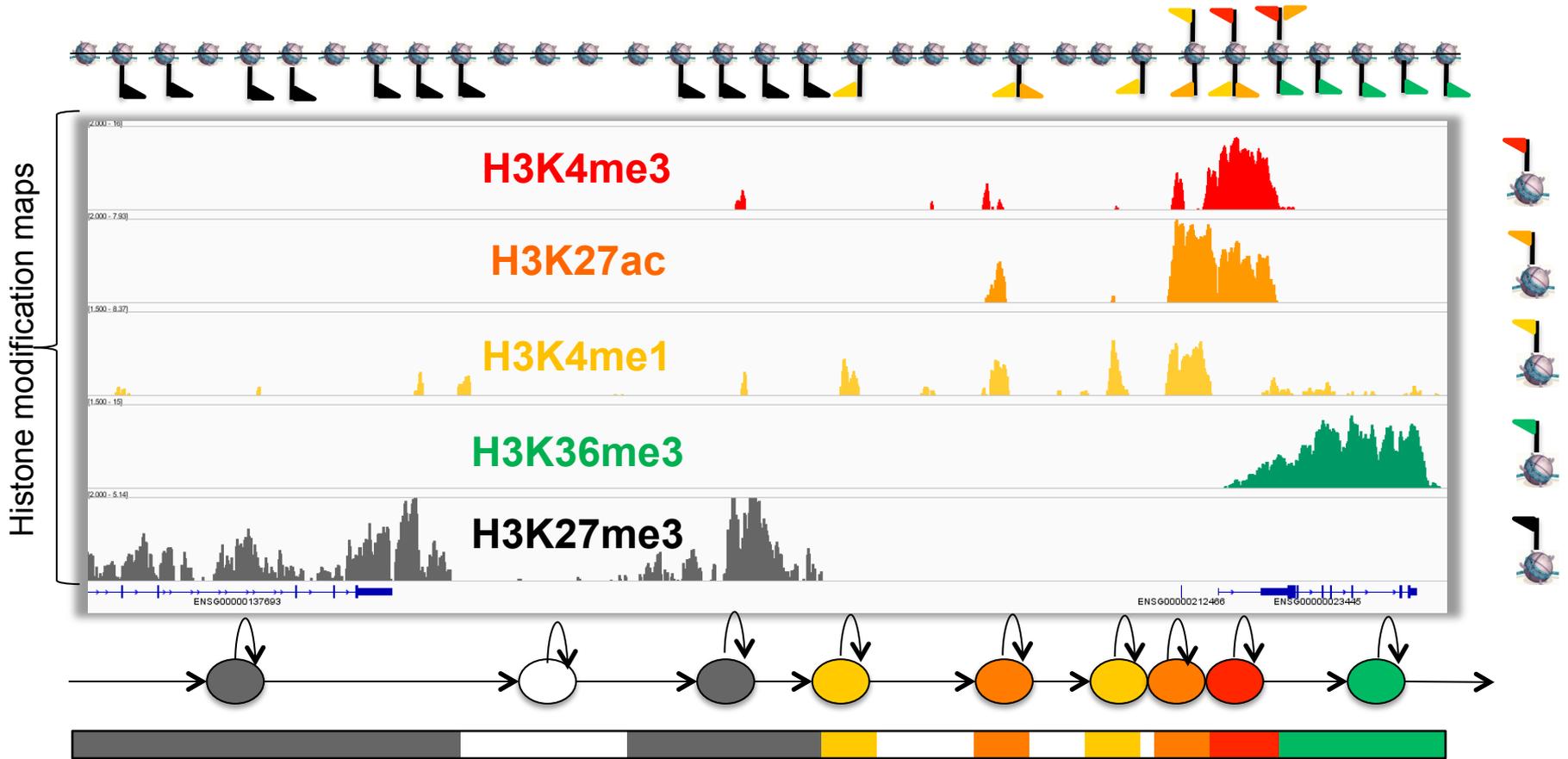
- The latest ENCODE3 ChIP-seq data freeze contained a limited amount of new data. Instead of building new chromatin state models on these data, we may be able to use existing models.
- The Roadmap Epigenomics Project has released chromatin state models across 127 epigenomes, spanning 100+ cell types.
- These models can be applied to incoming ENCODE3 data, provided that these data are processed identically to what was used for Roadmap.
- Benefits: validate ENCODE3 cell types against large body of reference cell types, fairly rapid general QC, identify novel regions.

# The Roadmap Epigenomics Project



Performed a wide variety of epigenomics assays to increase our understanding of the role of epigenetics in gene regulation

# Learning genome-wide chromatin state maps

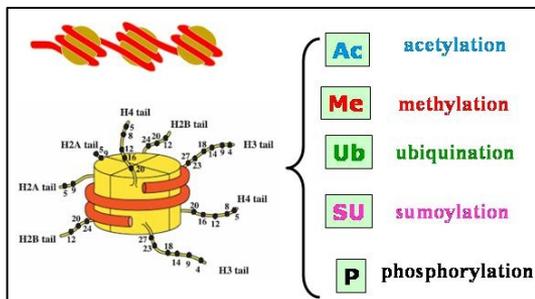


## Chromatin States: Hidden Markov model

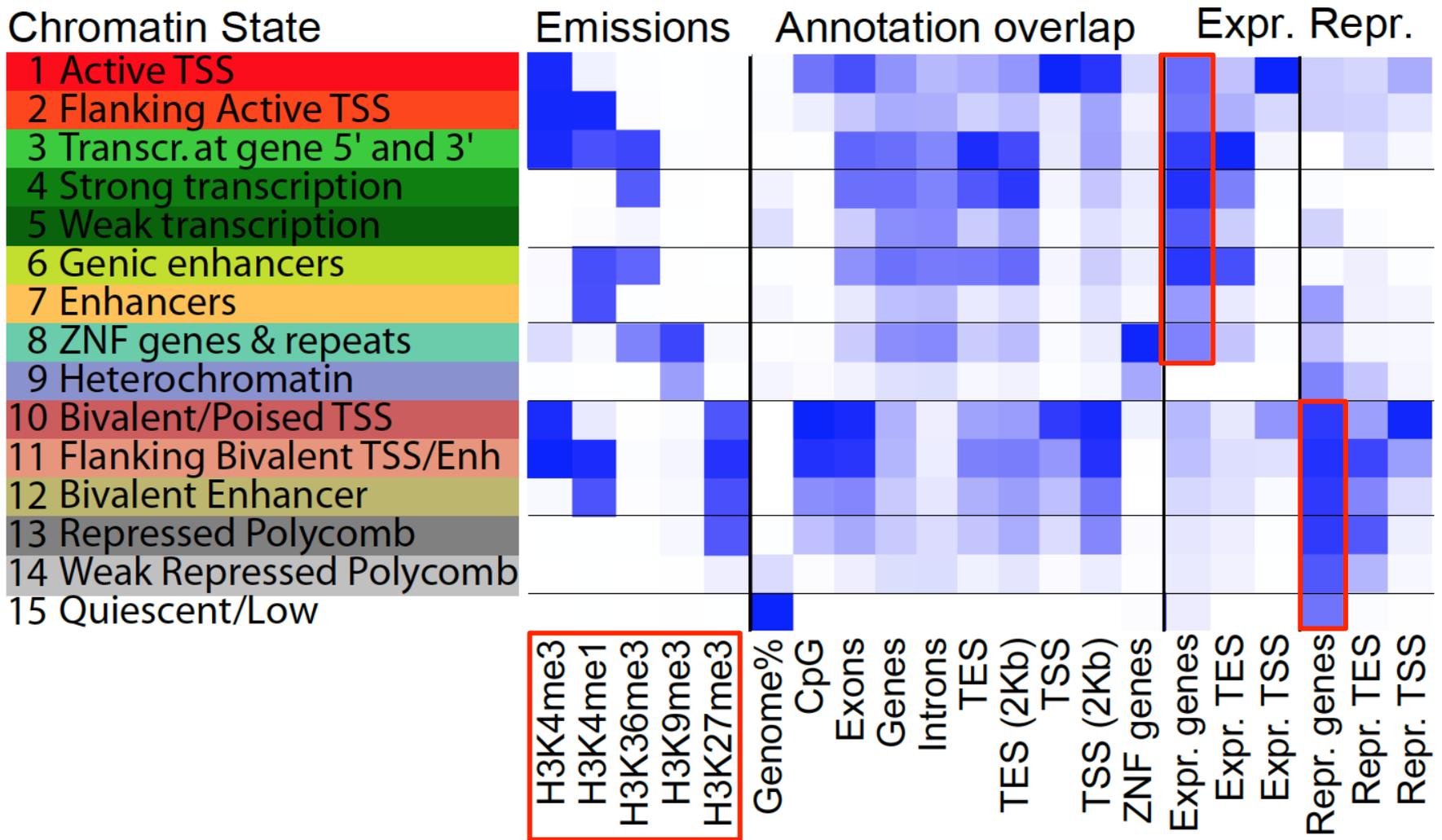
ChromHMM: automating chromatin-state discovery and characterization

Jason Ernst & Manolis Kellis

*Nature Methods* **9**, 215–216 (2012) | doi:10.1038/nmeth.1906

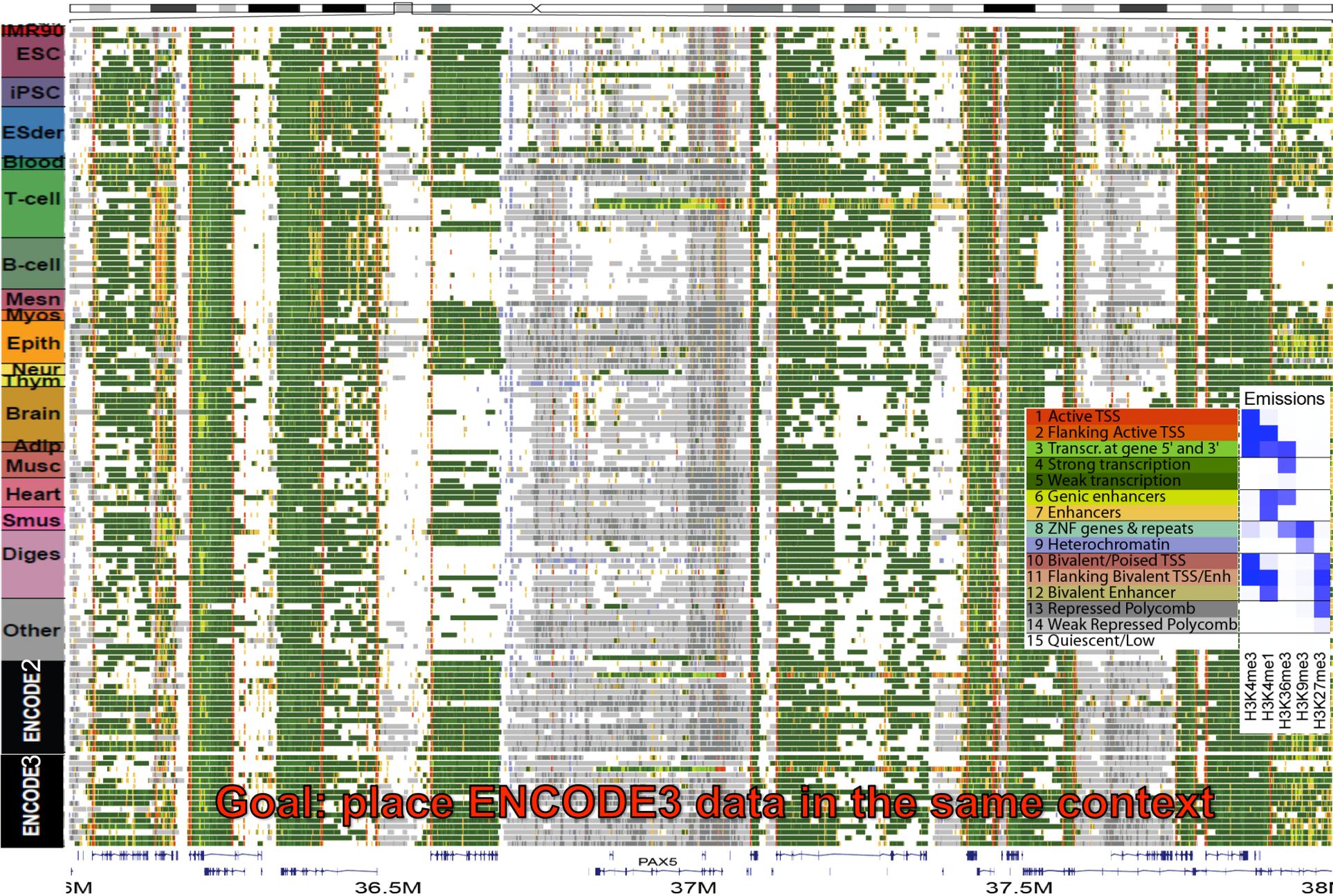


The 5 core histone modification marks have been used to build a ChromHMM model with 15 chromatin states



States make a clear distinction based on transcriptional activity

# Chromatin state annotations across 127 epigenomes



# ENCODE3 human ChIP-seq data on the DCC portal

## Summary for project file set ENCSR301PBV

Status: released

**Assay(s):** ChIP-seq  
**Accession:** ENCSR301PBV  
**Description:** Uniformly processed ENCODE3 human histone ChIP-seq October 2015  
**Project type:** Evaluation  
**Biosample term name:** Loucy, DOHH2, OCI-LY7 ← 3 cell types only?  
**Biosample type:** immortalized cell line  
**Organism:** human  
**Lab:** Zhiping Weng, UMass  
**External resources:** None submitted

## Files in project file set ENCSR301PBV

Visualize Data

### Raw data

Accession	Originating dataset	File type	Biological replicate	Technical replicate	Read length	Run type	Paired end	Mapping assembly	Lab	Date added
ENCF002ASS <a href="#">Download</a> 1.14 GB	ENCSR623GZY	fastq	2	1	36 nt	single-ended			Bradley Bernstein, Broad	2014-03-19
ENCF002ATI <a href="#">Download</a> 1.1 GB	ENCSR005SXO	fastq	1	1	36 nt	single-ended			Bradley Bernstein, Broad	2014-03-19
ENCF002ATK <a href="#">Download</a> 1 GB	ENCSR447ZGY	fastq	2	1	36 nt	single-ended			Bradley Bernstein, Broad	2014-03-19

etcetera

<https://www.encodeproject.org/projects/ENCSR301PBV/>

Navigated to via:

<https://www.encodeproject.org/search/?type=Project&lab.title=Zhiping+Weng%2C+UMass&status=released>



# ENCODE3 chromatin state calls

- Current ENCODE3 data I'm working with:
  - 19 cell types (next slide)
  - 6 epitopes (K4me1/3, K27me3/ac, K9me3, K36me3) + WCE
- Strategy:
  - process all mapped sequencing reads of ENCODE3 data in a similar manner as done for Roadmap Epigenomics.
  - Remove duplicates & pool data across replicates.
  - Aim for at least 30M reads per cell type / epitope combination. If more, subsample to 30M.
  - Call chromatin states for processed ENCODE3 data, using Roadmap Epigenomics models.

# QC: number of (non-duplicate) reads in millions

## Broad pipeline

## Roadmap/ENCODE3 pipeline

56	116	63	61	57	66	80	Fibroblast-of-arm
102	56	95	98	93	101	72	PC9
76	69	65	61	66	98	252	H1-derived-neurons
96	68	81	89	83	79	135	MM.1S-Myeloma-cell-line
34	42	55	55	69	135	97	Neuroectoderm
87	80	79	92	85	77	104	PC3-prostate
29	40	44	61	59	50	82	Epiblast
55	83	97	66	41	38	76	Radial-Glia-VZ
47	62	58	55	84	35	86	MCF-7
61	48	56	57	42	58	116	DOHH2
14	62	75	77	45	71	68	Neuroepithelial
135	39	83	40	43	39	43	SuDHL-6
49	55	49	54	44	111	43	Oci-Ly-3
83	40	50	47	53	56	61	Karpas-422
48	42	38	36	16	42	145	Oci-Ly-7
39	45	40	41	69	50	62	LOUCY
33	73	32	33	31	30	48	Oci-Ly-1
28	29	37	37	32	21	103	KOPT-K1-KOPzero
8	15	8	6	16	11	77	A673
H3K27me3	H3K9me3	H3K36me3	H3K4me1	H3K27ac	H3K4me3	WCE	

93	177	99	104	101	112	126	Fibroblast-of-arm
116	100	107	112	113	116	56	PC9
60	79	50	50	54	87	232	H1-derived-neurons
75	68	78	70	66	62	124	MM.1S-Myeloma-cell-line
25	30	57	44	120	164	76	Neuroectoderm
69	61	62	77	72	63	84	PC3-prostate
39	44	56	69	63	61	104	Epiblast
58	61	75	78	47	29	58	Radial-Glia-VZ
37	46	44	45	81	28	90	MCF-7
49	35	44	46	34	48	89	DOHH2
9	42	56	62	58	57	52	Neuroepithelial
105	28	64	31	33	29	33	SuDHL-6
38	40	37	44	36	91	33	Oci-Ly-3
63	28	37	38	43	43	48	Karpas-422
35	31	29	30	14	37	112	Oci-Ly-7
28	32	31	34	55	40	49	LOUCY
26	53	24	28	26	27	37	Oci-Ly-1
22	16	27	29	26	18	79	KOPT-K1-KOPzero
7	9	7	7	15	15	63	A673
H3K27me3	H3K9me3	H3K36me3	H3K4me1	H3K27ac	H3K4me3	WCE	

# Processing pipeline

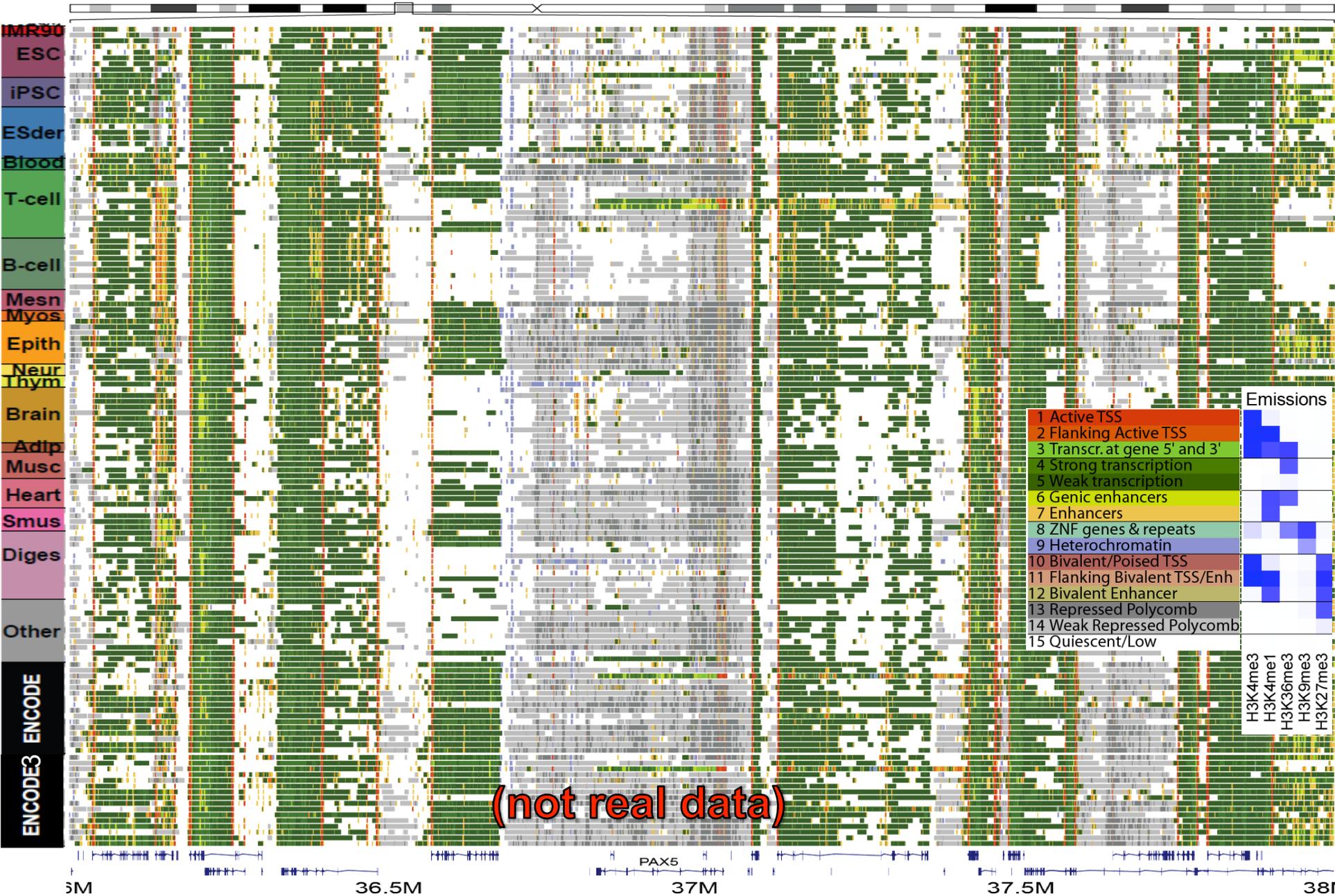
Largely based on code from Anshul and the (draft?) ENCODE3 processing pipeline\*

Processing steps:

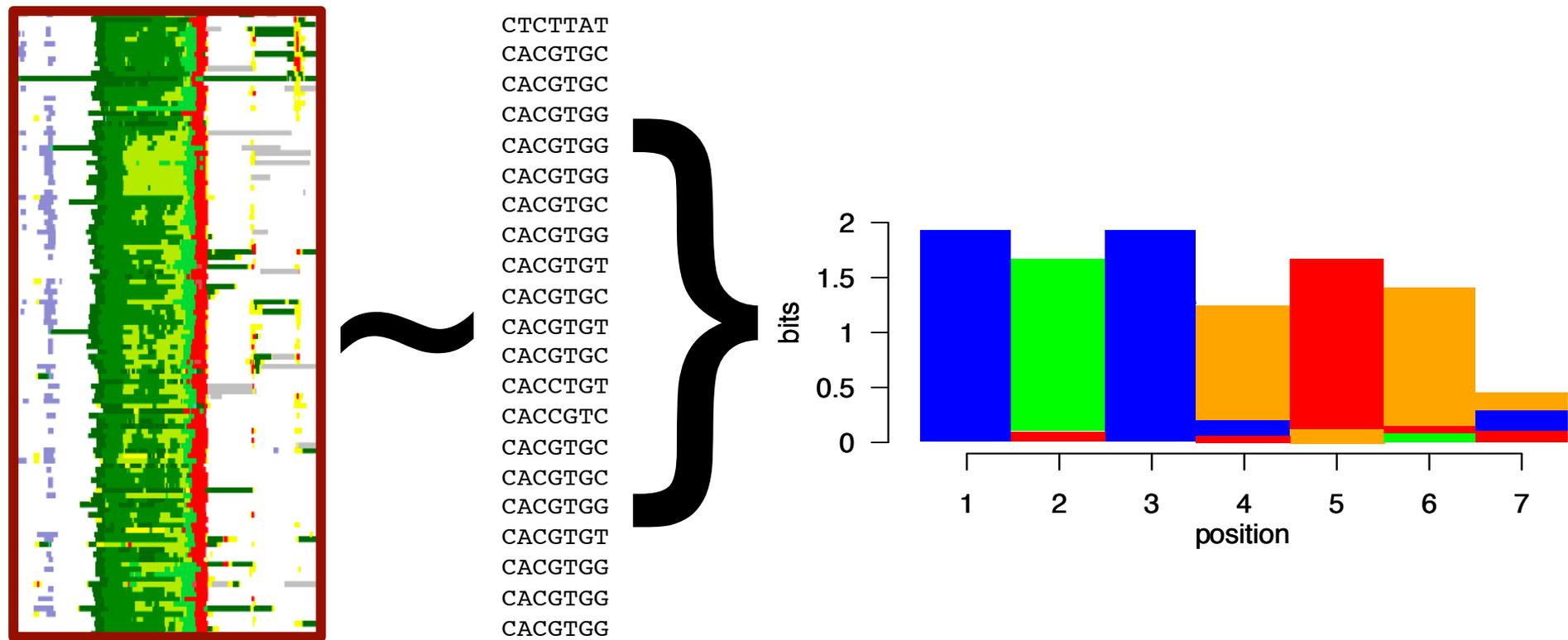
1. Remove un/mis-mapped, low quality reads, mark & remove duplicates, compute library complexity QC, filter for mappability
2. Pool reads across replicates and (if needed) sub-sample to 30M reads, use phantomPeakQualTools to obtain QC info
3. Binarize data using ChromHMM BinarizeBed, create segmentation based on existing Roadmap models and output states per 200bp bins

\*: [https://docs.google.com/document/d/1IG\\_Rd7fnYgRpSlqrlfuVIAz2dW1VaSQThzk836Db99c/edit#heading=h.9ecc41kilcvq](https://docs.google.com/document/d/1IG_Rd7fnYgRpSlqrlfuVIAz2dW1VaSQThzk836Db99c/edit#heading=h.9ecc41kilcvq)

# Bonus: **epi**logos for ENCODE3 data

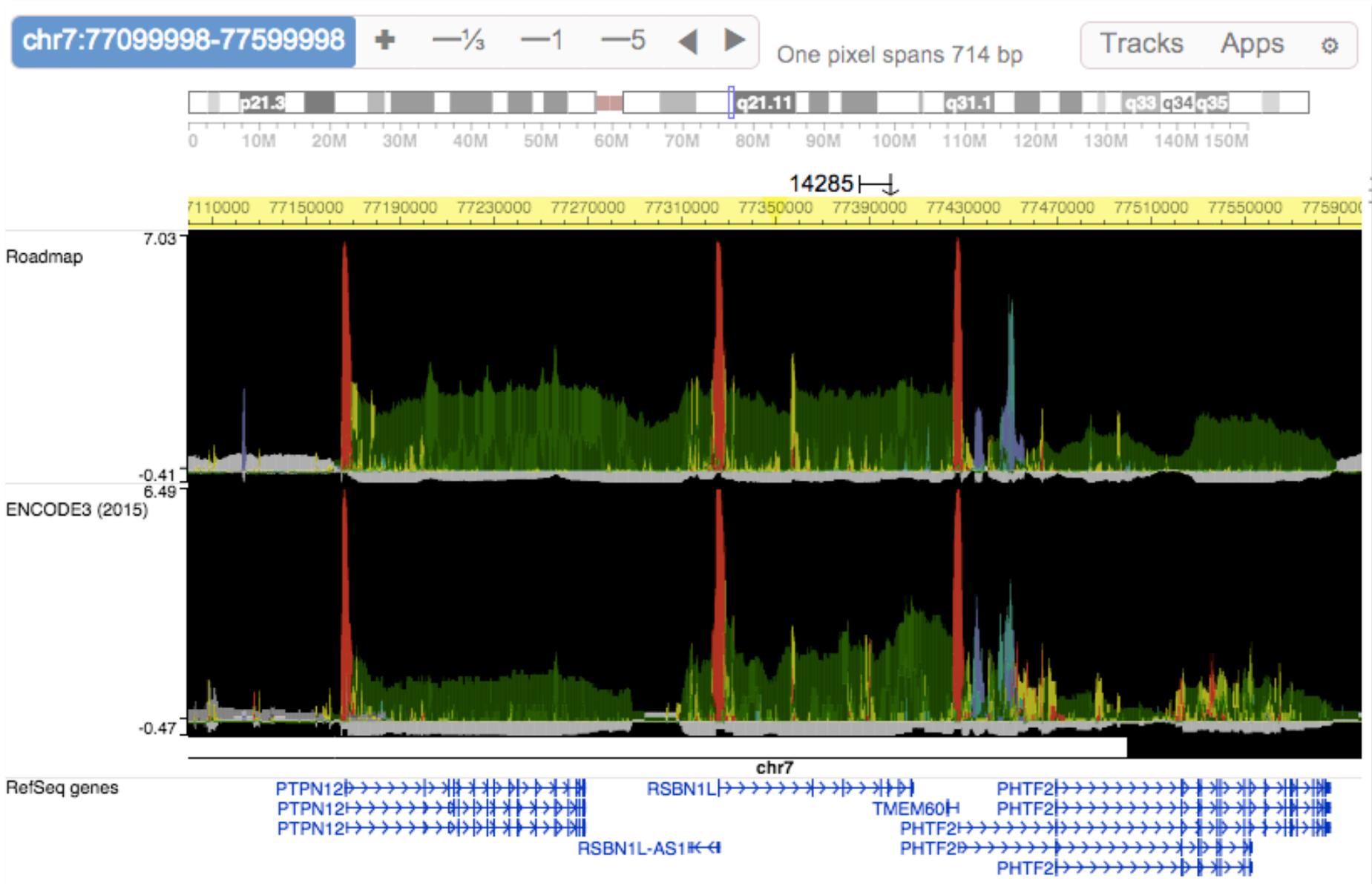


Chromatin state calls across many epigenomes can be viewed as an alignment of sequences with a finite alphabet



There are good ways of modeling such alignments: logos!  
*Information content* of a region, considering background

# ENCODE3 data epilogos



<http://tinyurl.com/encode-vs-roadmap>