

DeLEG: Deep Learning for EpiGenomics data to predict phenotype.



Phenotype, genotype and environment

Phenotype = genotype + environment

$$P = G + E$$

Interaction between genetics and environmental factors

$$P = f(G, E)$$

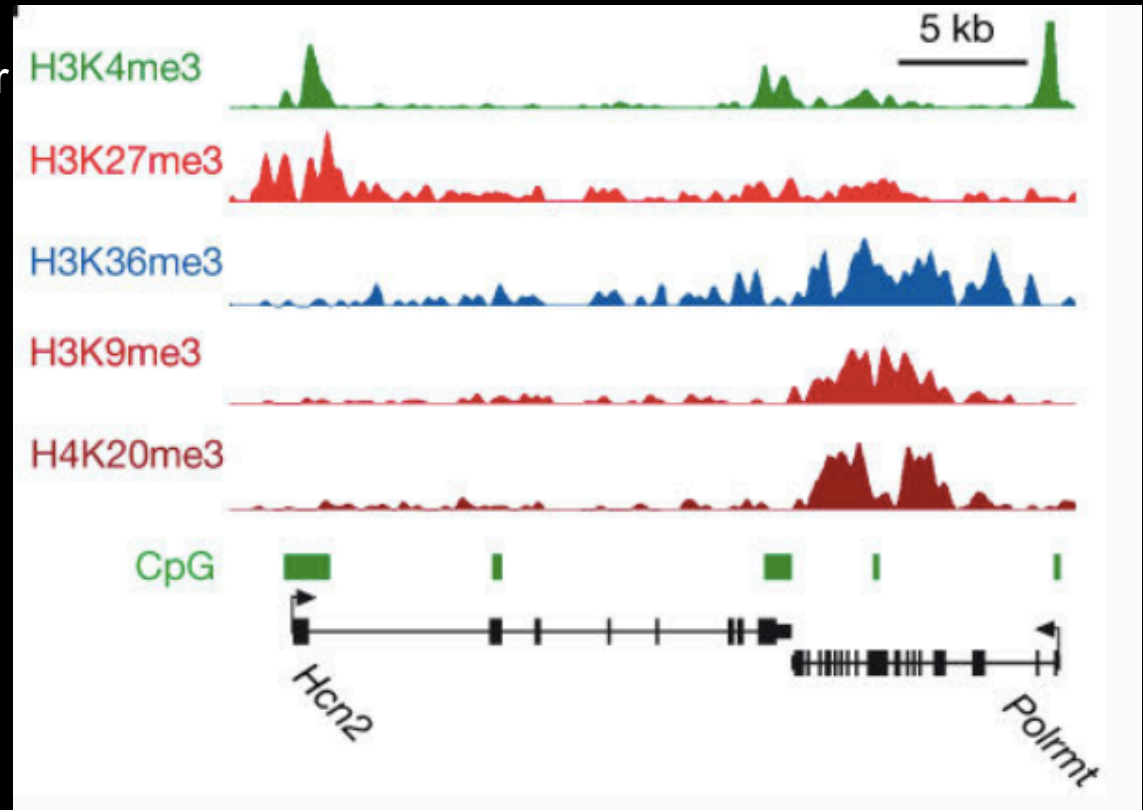


Challenges

1. Finding “important” regions in ChIP-seq data
2. Using the “important” regions for prediction, classification and better understanding of Human Epigenome

HEADLINE:

- Ground truth: “Healthy” and “Disease”
- Training set: 14 subjects in each phenotype
- Validating set: 4 subjects in each phenotype
- Testing set: Any new subject not in the set of 36 subjects



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Workflow

Otsu thresholding for segmentation

Enrichment score of window
around TSS

Train a Conv Neural Network
for classification

Input: ChIP-seq data

Output: “Important” regions
of the ChIP-seq data
corresponding to peaks

Detailed Method: Otsu
thresholding of the ChIP-seq
data to remove background
and filter out peaks. A post-
processing for noise removal
is done following
thresholding.

Input: ChIP-seq data and
peak regions

Output: Enrichment score
from ChIP-seq data of each
bp around TSS for each
gene corresponding to
peaks

Detailed Method: Gene
identification from database
and window extraction (data
manipulation process)

Input: Enrichment score of
fixed length sequences

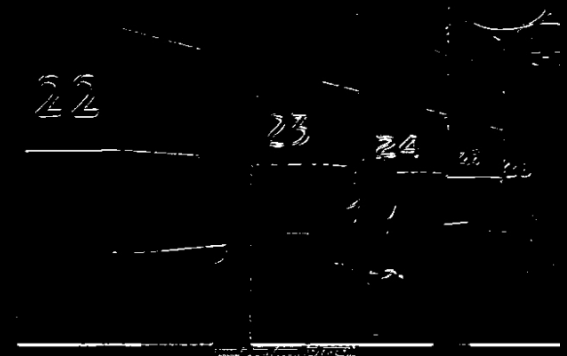
Output: Classification
probabilities for the window
to lie in each phenotype.

Detailed Method: Training a
CNN with windows from 14
subjects in each class,
totalling to about 20k
windows in each class.
Looking back at the learned
features -> further insight

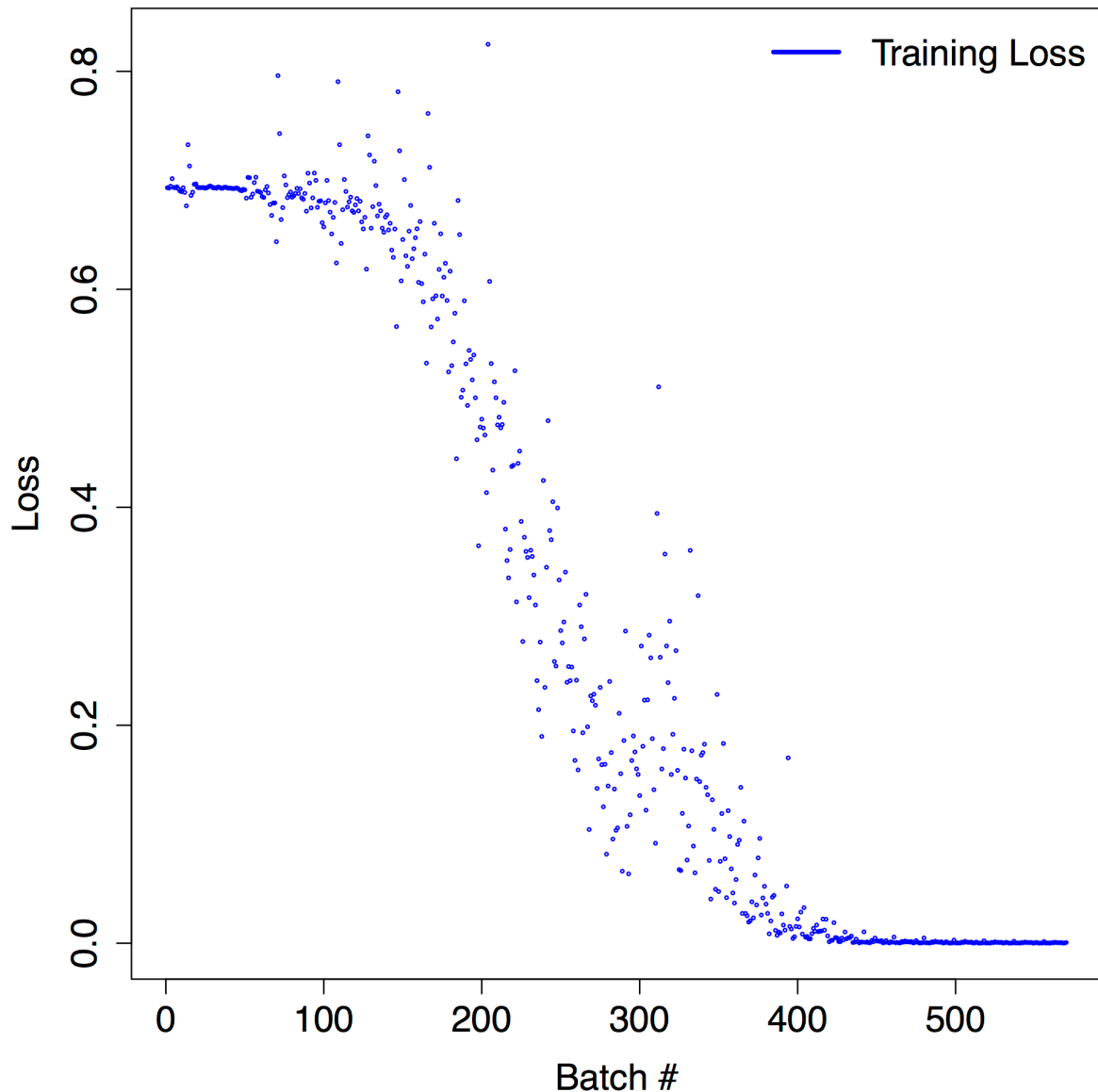
Learning attention from classification data

“ Those who pay attention *learn*,
Those who don't cram ”

- Use Global Average Pooling concept to learn regions in input which caused the network to classify it to a particular class
- Final layer filters know where to look
- Get defining regions - define further research!!



Supplementary Material Results



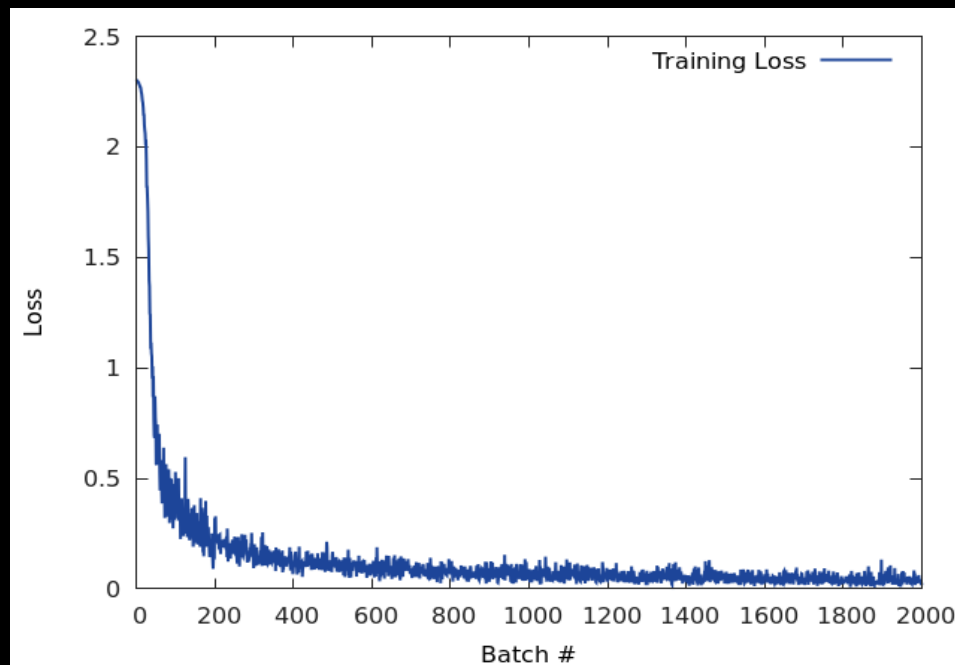
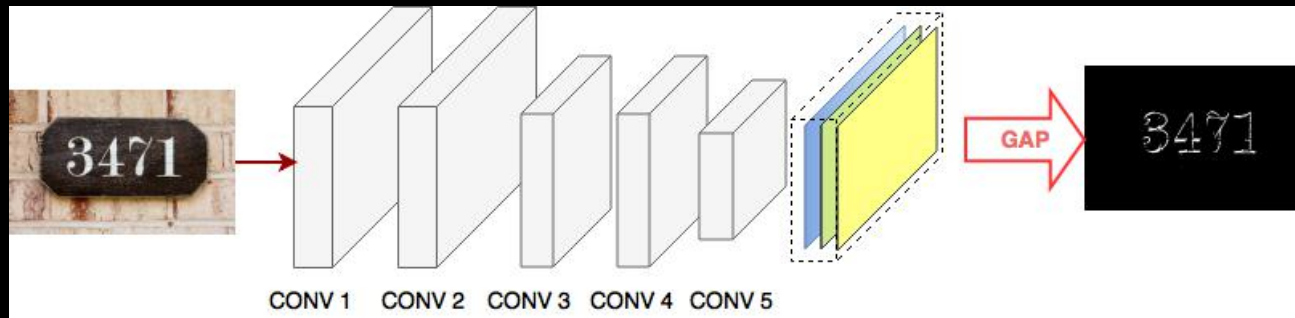
Testing accuracy

percentage correct 81.831664812755 %

FER 70.564186426819 %

INF 91.180461329715 %

Supplementary material - Results



Class	Accuracy
0	99.59%
1	99.47%
2	99.90%
3	98.51%
4	99.08%
5	98.32%
6	98.23%
7	98.64%
8	97.43%
9	97.92%
Overall	98.73%