



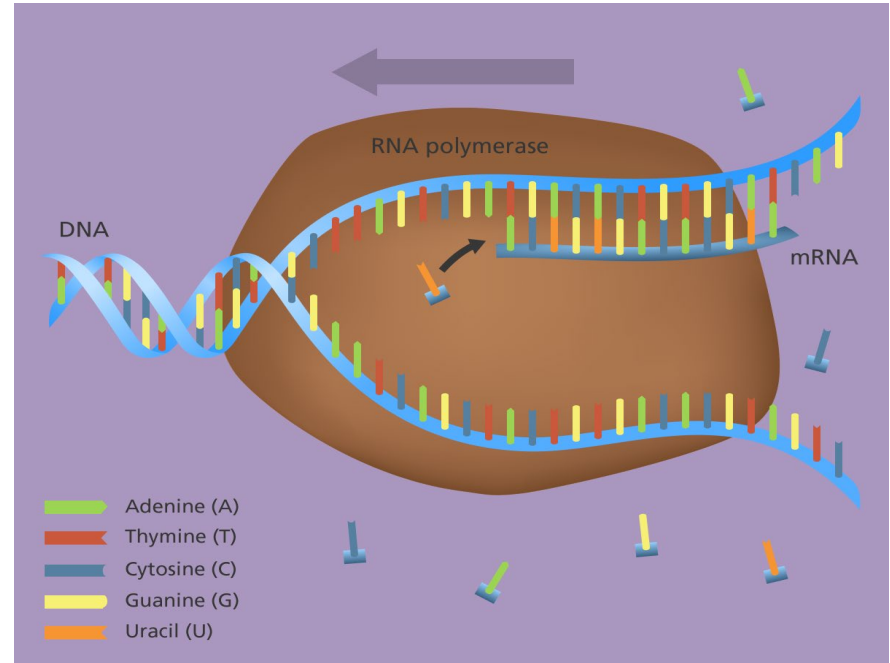
The Identification of Transcriptionally Active Regions

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University of Minnesota - Morris April 20, 2022

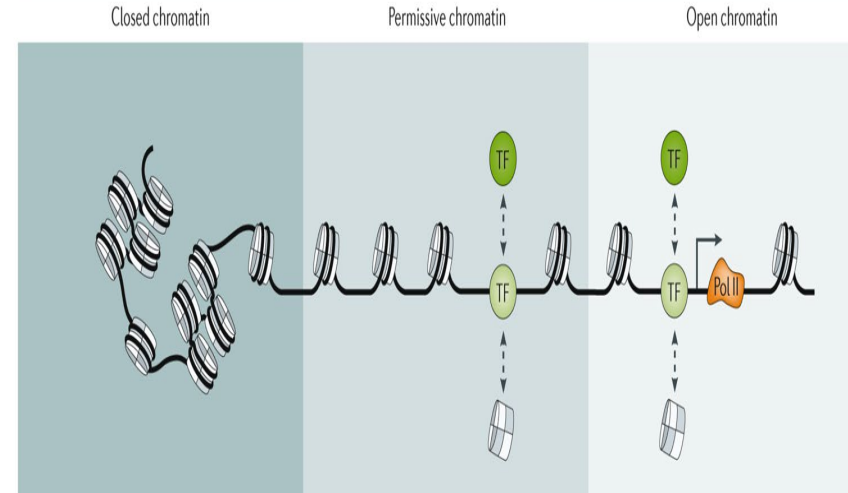
What is Transcription?

- One of the main processes for the flow of genetic information in an organism
- Four Key Components
 - Chromosomes
 - Proteins
 - Genes
 - RNA polymerase
- Transcription factors
- DNA



Chromatin Accessibility

- Chromosomes are made up of chromatin
- There are different types of chromatin
 - Closed accessibility
 - Transcription can not take place
 - Permissive accessibility
 - Regions are permissive of transcription factors
 - Open accessibility
 - Regions can be accessed by transcription factors and the RNA polymerase enzyme accessibility



A figure on the different types of chromatin accessibility. The three blue areas of the image left from right represent closed, permissive, and open accessibility. [Klemm et al., 2019]



Diseases and Mutations

- Mutations are changes in a DNA sequence
 - Mistakes during the transcription or translation process
 - Caused by environmental factors
- Diseases are caused by mutations
 - Gene(s)
 - Caused by environmental factors
 - Damaged chromosomes
- Series of open chromatin regions are associated with diseases such as cancer [Wang et al., 2021]
- p53 gene can be related to cancer



Treatments and Medicine

- Identification of disease mechanisms is crucial in understanding
 - What mechanisms are associated with specific diseases
 - Determine potential treatments and medications for specific diseases
- Disease mechanisms can be better understood through studies about accessibility and transcriptionally active regions [Perrin et al., 2021]



Problems with current methods

- Nascent transcription methods (GRO-seq)
 - Identifies nascent transcripts (there are methods that identify mature mRNA)
- Current Issues
 - Expensive
 - Difficult to use
 - Requires large portions of cells
 - Struggle in identifying transcriptionally active regions from large sites
- Issues solved through the use of ATAC-seq and recurrent neural networks
 - Classification problem
 - Sequential data



Outline

- Background Information
 - Recurrent Neural Networks
 - ATAC-seq
 - Nascent Transcription (GRO-seq)
- Implementation
 - Data Attainment
 - Results
- Conclusion



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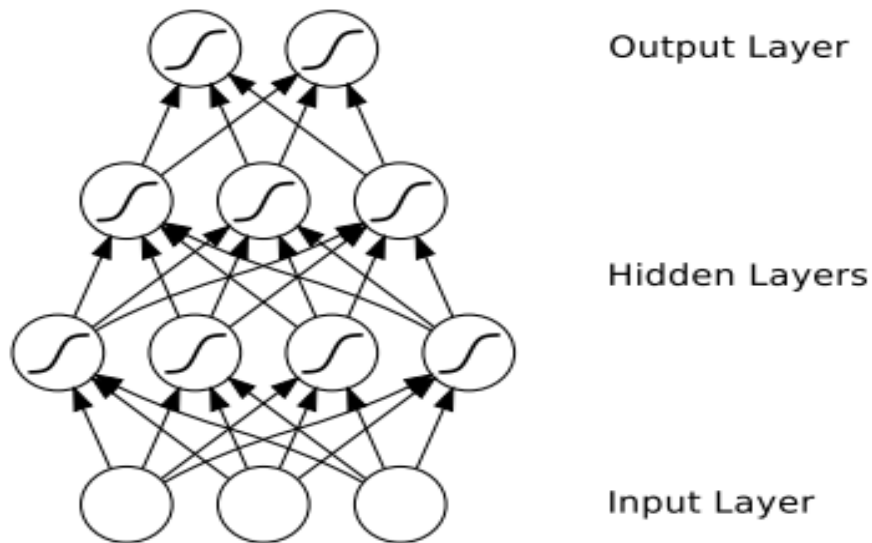


Machine Learning

- Machine learning
 - Study of statistical models that automatically improve themselves through experience and data
- Artificial neural networks
 - Subset of machine learning consisting of computing systems inspired by the physical processes observed in brains
- Feedforward neural networks (FNN)
 - A form of artificial neural networks

Feedforward Neural Networks - Organization

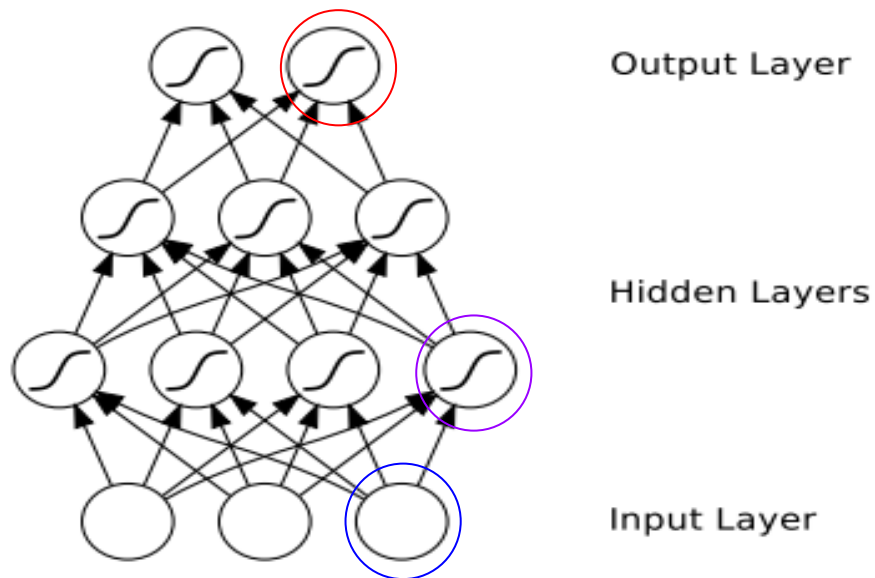
- Organized into
 - Input layer
 - One or more hidden layers
 - Output layer



Feedforward neural network. [Graves, 2012]

Feedforward Neural Networks - Nodes

- Nodes
 - Computational units
 - Receives values from previous nodes
 - Performs computation
 - Outputs values to nodes in the next layer or as a final value
- Each node consists of an activation function
 - Map inputs that vary non-linearly to an output



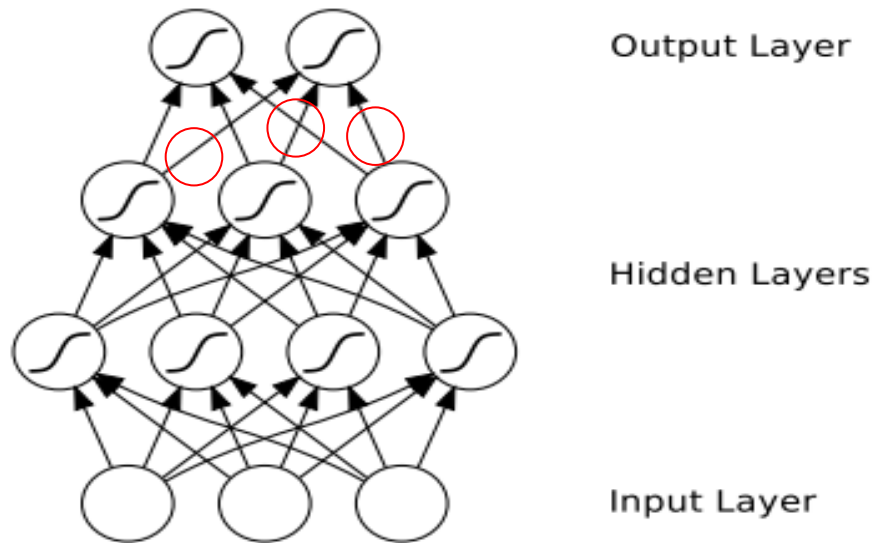
Feedforward neural network. [Graves, 2012]

Feedforward Neural Networks - Edges

- Edges
 - Transmits values to nodes between layers
 - Contains a weight that determines how much

the output value

activation function $\left(\sum_{i=1}^3 x_i * w_i \right)$



Feedforward neural network. [Graves, 2012]

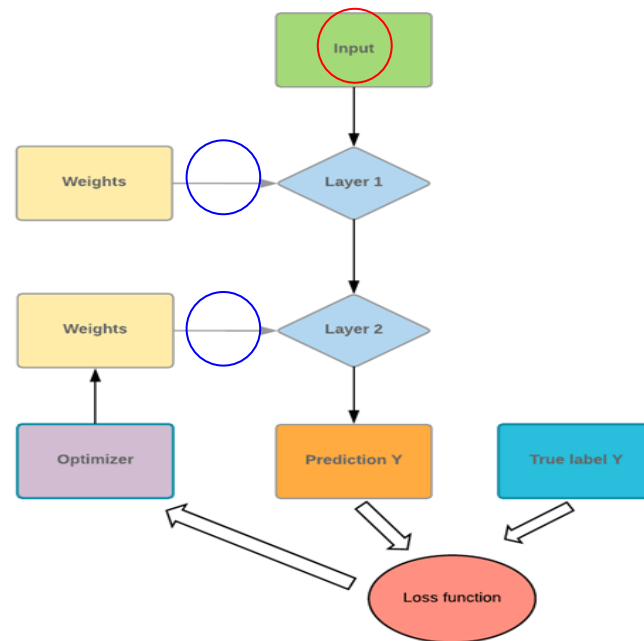


Feedforward Neural Networks - Training Runthrough

- Training dataset
 - Train model to map inputs to an expected outputs for the variable being predicted
- Testing dataset
 - Make estimates on what the expected outputs are for the variable being predicted in the testing dataset
 - Compare predicted and actuals

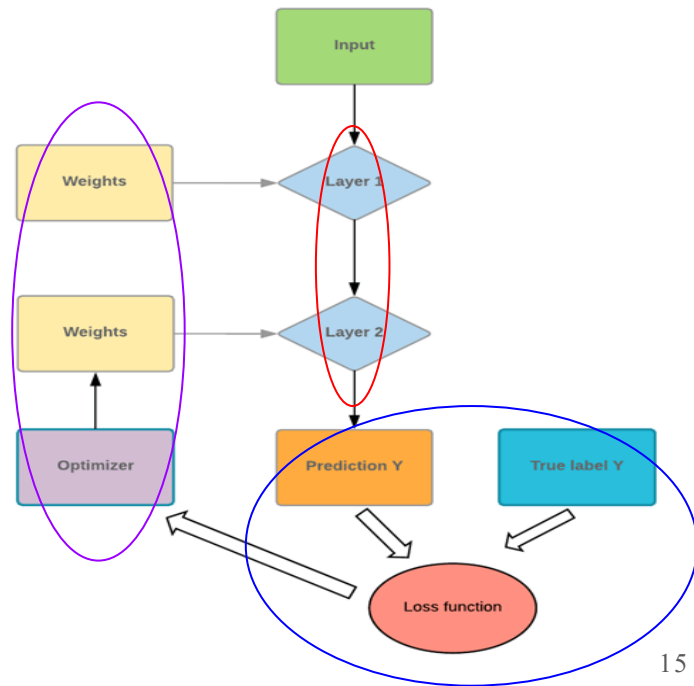
Feedforward Neural Networks - Training Runthrough cont.

- Provide the training dataset to the model
 - Choose the output variables (what is being predicted)
 - Choose the input variables (what is used to determine the outputs)
- Assign random weights



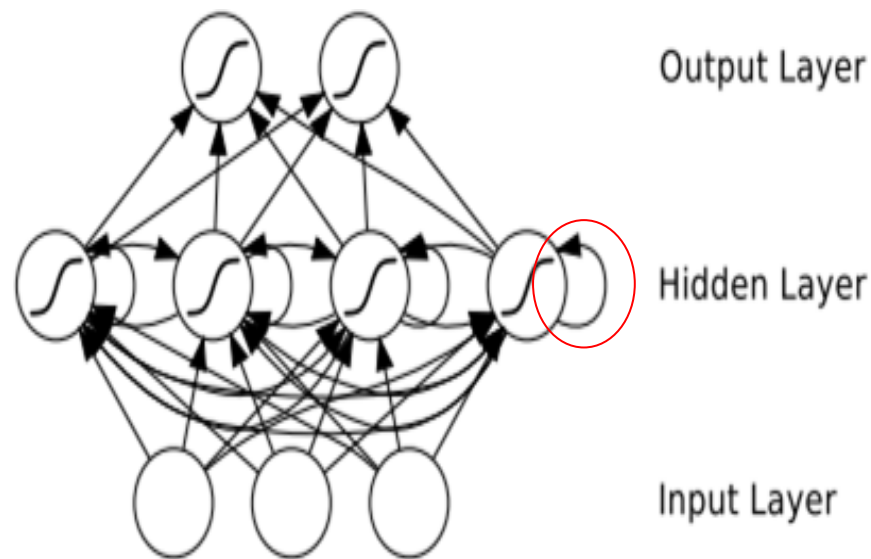
Feedforward Neural Networks - Training Runthrough cont.

- Forward Propagation
 - Mapping input values to an output value
- Loss Function
 - Determines the distance between the expected outputs and predicted outputs
- Back propagation
 - Done through an optimizer
 - Tuning the weights to lower the value produced by the loss function
- Forward and back propagation are ran until
 - Loss function value reaches a sufficient value



What are Recurrent Neural Networks?

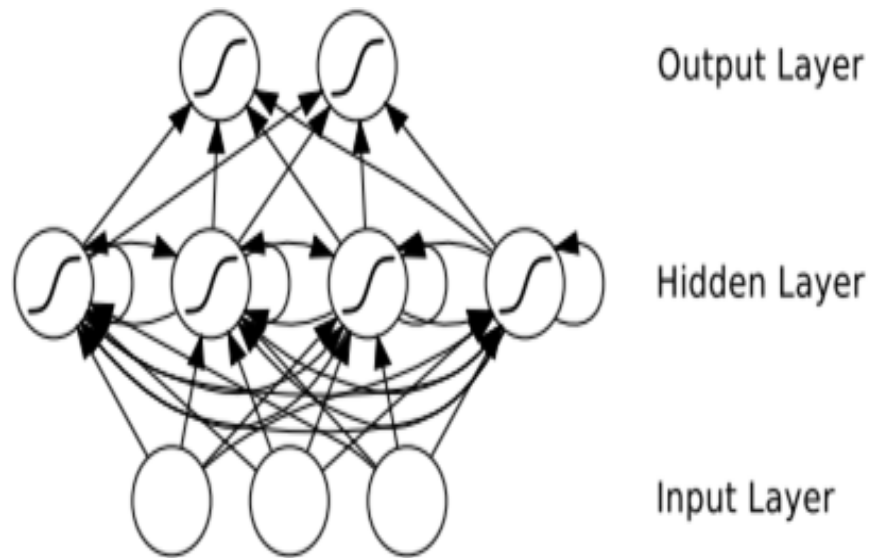
- Derived from Feedforward Neural Networks
 - Handles sequential data
 - Loops allow each previous observation to have an affect on the input in the next observation



Recurrent neural network. [Graves, 2012]

What is the Vanishing Gradient Problem?

- Vanishing Gradient Problem
 - Back propagation has little effect on the weights in earlier layers
- Gated recurrent unit
 - Solves the problem through the use of an update and reset gates
 - Decide what information is relevant to the output and can retain past information



Recurrent neural network. [Graves, 2012]



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What is ATAC-seq?

- Determines the accessibility of chromatin
- Two parts of the methods
 - ATAC-seq method
 - Pipeline
 - Series of steps involving tools to analyze the ATAC-seq reads



Short read runs

- DNA sequencing
 - Technique that determines the exact sequence of organic molecules in a DNA sequence
- Short read runs (reads)
 - A set of raw sequences after amplifying and sequencing copied short fragments of the DNA



Attainment of ATAC-seq Reads

- Sequencing adapters
 - Short single strands of synthetic
- Tn5 transposase
 - Inserts sequencing adapters at accessible regions of the genome
- Genome
 - The set of all genetic information about an organism
- Reads are stored in a database with annotations
 - Where the reads are from
 - The study where the genome was sequenced
 - The species that was sequenced

DNA or RNA



ATAC-seq Pipeline - Preprocessing

- Download the reads and annotations
- Check quality of the reads
- Determine presence of sequencing adapters
- Trim reads
 - Remove low quality reads
 - Sequencing adapters



ATAC-seq Pipeline - Mapping

- Human reference genome
 - Database on one idealized individual of a species
- Map the reads to the reference genome of interest
 - Matches read to most similar regions

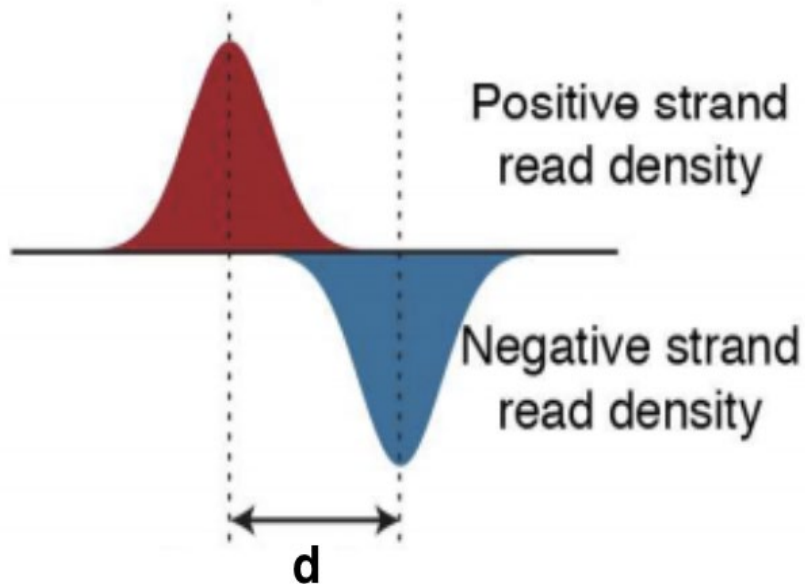


ATAC-seq Pipeline - Filters

- Filters are imposed on the mapped reads to remove
 - Uninformative reads
 - Duplicate reads
 - Low mapping quality reads
 - Reads that are not properly paired
- Insert size is checked
 - Length of space between the sequencing adapters
 - Determine the quality of reads

ATAC-seq Pipeline - Peak Calling

- Determine density distribution
 - Both strands of the DNA
 - On the two density distributions for both strands of the DNA
- Regions with significant differences
 - Peaks (open chromatin regions)





What is Nascent Transcription?

- Identifies nascent transcripts
- Two parts of the methods
 - Nascent transcription (GRO-seq)
 - Pipeline similar to ATAC-seq



Attainment of Nascent Transcription Reads

- Halts transcription
- Nucleus
 - Controls and regulates the activities of cells
 - Isolated from a cell population of interest
- Organic molecules labeled with tags (markers) are added
- Transcription is then restarted
 - Nascent transcript molecules are labeled
- Reads are produced with annotations



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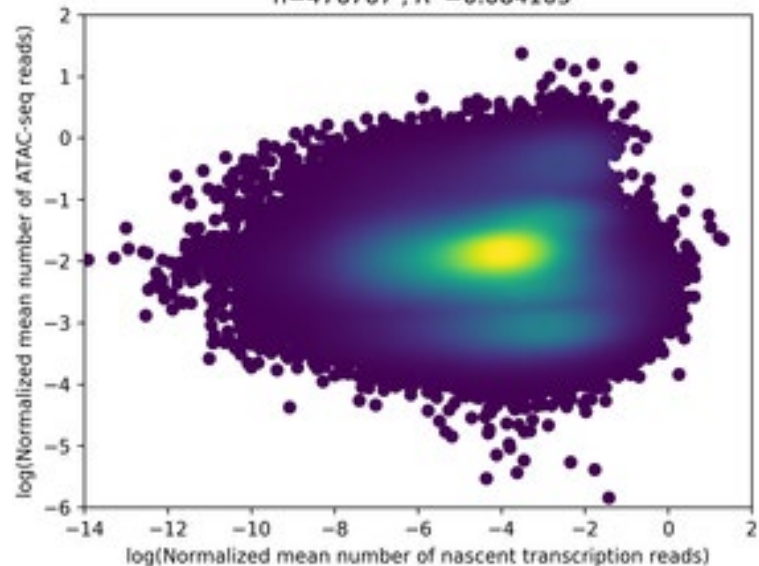
Implementation - Data Attainment and Pipeline

- Gene Expression Omnibus (GEO)
 - Database that consists of data on gene expression and DNA sequencing
- ATAC-seq and nascent transcription (GRO-seq) reads from GEO
 - Filtered under the criteria of the reads having matching cell types and diseases such as cancer
 - Nine sets of reads were attained
- Pipelines
 - Approximately half a million open chromatin regions were identified from the ATAC-seq pipeline
 - 29 percent of open chromatin regions were labeled as transcriptionally active from the nascent transcription (GRO-seq) pipeline
 - Used as the expected outputs

Implementation - Correlation Between Accessibility and Nascent Transcription

- Low correlation between accessibility and nascent transcription
 - Each point is a open chromatin region
 - Comparing the mapped open chromatin reads from ATAC-seq and nascent transcription (GRO-seq)

Relation between nascent transcription coverage and ATAC-seq coverage
 $n=476767$, $R^2=0.084105$

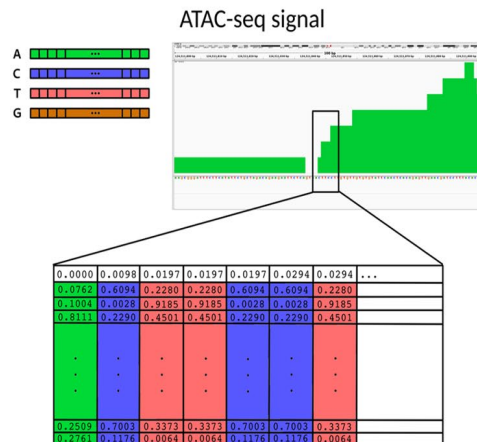


Relationship between accessibility and nascent transcription.
[Tripodi et al., 2020]

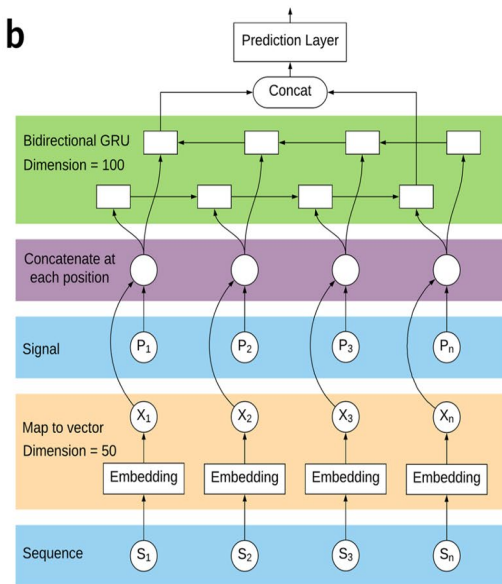
Implementation - Vector Encoding

- Developed a vector encoding
 - Sequence
 - For each open chromatin region
 - Signal
 - Number of mapped ATAC-seq reads for each open chromatin region divide by millions mapped

a



b



Vector encoding signal and sequence to summarize each open chromatin regions. [Tripodi et al., 2020]

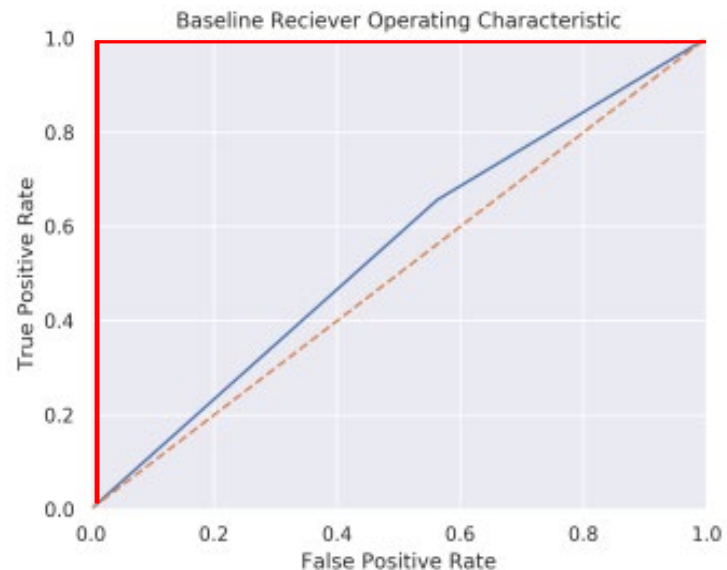


Implementation - Measures

- True positive rate
 - $TP / (TP + FN)$
 - True Positive (TP)
 - Predicted as transcriptionally active and actually is
 - False Negative (FN)
 - Predicted as not transcriptionally active but is actually is
- False positive rate
 - $FP / (FP + TN)$
 - False Positive (FP)
 - Predicted as transcriptionally active but actually is not
 - True Negative (TN)
 - Predicted as not transcriptionally active and actually is not
- F1 Score
 - $TP / (TP + (\frac{1}{2})(FP+FN))$
 - Measure of a model's accuracy on a dataset

Implementation - Baseline

- Developed a baseline method
 - Predicts the probability of whether an open chromatin region is transcriptionally active or not based on the distribution of mapped ATAC-seq reads per open chromatin region labeled as transcriptionally active or not



Baseline method for classifying open chromatin regions as transcriptionally active or not. [Tripodi et al., 2020]

Implementation - RNN Model Template

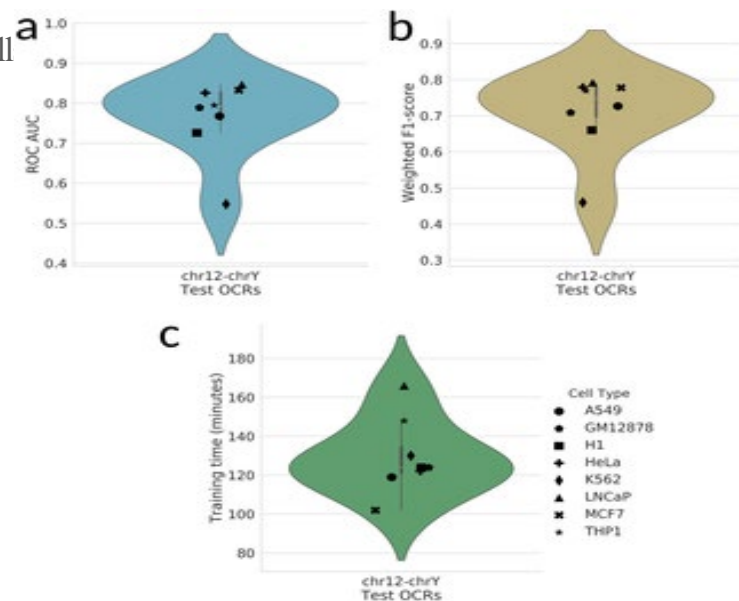
- Leave-one-out training(LOOT)
 - Model trained eight times
 - Leaves a cell type out each time
 - Trained for each of the cell types
- Data split
 - 90 percent training data
 - 10 percent testing data

Cell Type	chr1	chr2	...	chr11	chr12	...	chr21	chr22	chrX	chrY
A549						
GM12878						
H1						
HeLa						
LNCaP						
MCF7						
THP1						
HCT116						

Training, validating, and testing sets. [Tripodi et al., 2020]

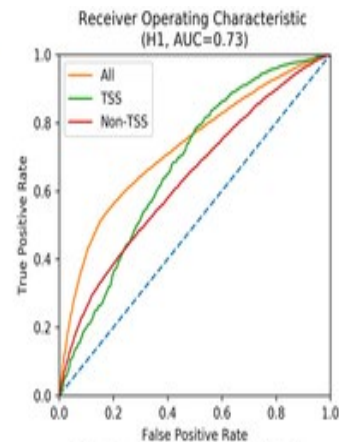
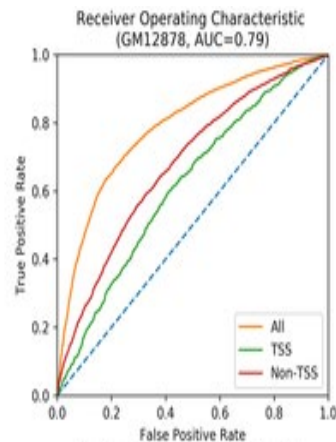
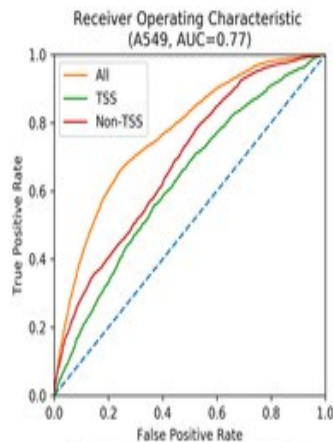
Implementation - RNN Performance

- ROC AUC score is above 70 percent for majority of the cell
 - K562 had the worse quality
- F1 score
 - F1 score is above 65 percent for majority of the cells
- Training time
 - The RNN models for each of the cell types took over 100 minutes



Implementation - RNN Performance cont.

- Transcription start site (TSS)
 - The location where the first DNA nucleotide is transcribed into RNA
 - TSS only represent a small fraction of all transcriptionally active regions



Error analysis using ROC curve for comparing predicted vs actuals. [Tripodi et al., 2020]



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Conclusion

- Transcriptionally active regions can be identified accurately through the use of ATAC-seq and recurrent neural networks
- Maximizes the benefits of ATAC-seq so that disease mechanisms can be more efficiently studied



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References

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Questions?