



# Interpretable prediction of DNA replication origins in S.cerevisiae using attention-based motif discovery

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## INTRODUCTION

DNA replication: Biological process of producing two identical replicas of DNA from one original DNA molecule.

DNA replication origins: Genomic locations initiating replication.

#### S.Cerevisiae Genome:

- Many Origins
- Known Locations
- ➤ Identified Consensus Sequences

## **ARS**

ACS

- Autonomously Replicating Sequence in
- the S.Cerevisiae genome
- 200-1000 bp

## ARS Consensus Sequences

- Recognized by the Origin Recognition Complex (ORC)
- 12-17 bp

#### 5'-WWW-WTTTAYRTTTW-GTT-3'

(Theis and Newlon, 1997 [1]) 'W' denotes A or T, 'Y' denotes C or T, 'R' denotes G or A.

#### **Problem:**

- How to discriminate origin sequences from non-origins the by Genome LLMs?
- How to interpret the results?
- Is there anything on top of ACS which determines the replication origin?

## **RESULTS**

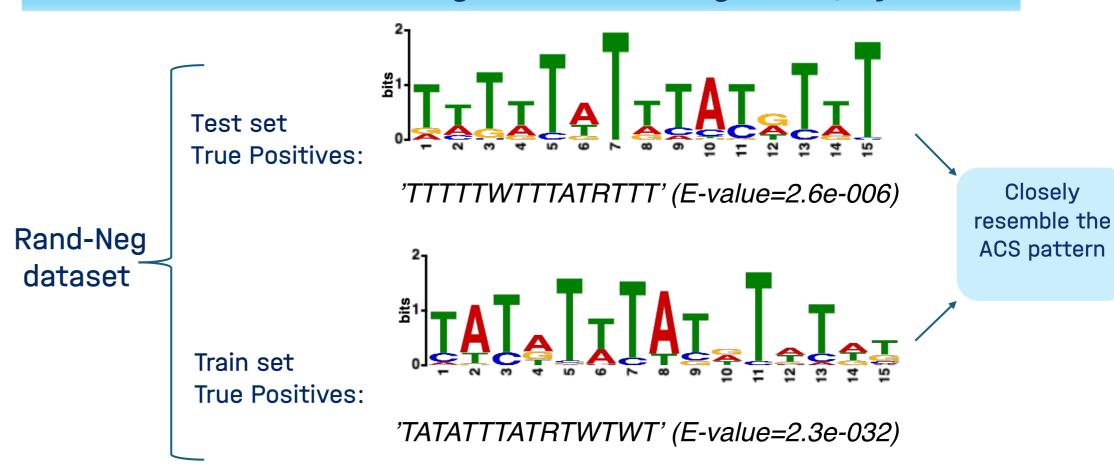
Performance Analysis

DATASET	ACC	AUC
Random-Neg	0.83	0.90
ACS-Neg	0.74	0.82
Shuffled-Neg	0.90	0.96
Block-5-Shuffled-Neg	0.77	0.86

ACS pattern is an important feature for model

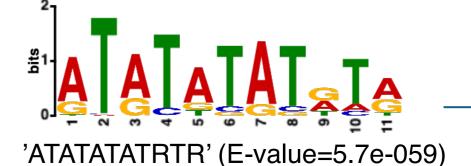
Ability of model in capturing both local and global sequence dependencies

## Motifs Discovered from High-attention Fragments, by MEME



ACS-Neg dataset:

Train set **True Positives** 



A/T bases, the pattern associated with the intergenic origins [5]

MEME motif

Discovery [4]

## **METHOD**

**DNABERT:** A multi-layer Transformers encoder following the same training process as BERT [2].

- Pre-trained on Human Genome data.
- Overlapping k-mers tokenization.
- Four balanced datasets are curated based on OriDB [3]
- Confirmed Origins < 500 bp, as positive</li> instances.
- Negative instances (non-origins) subsampled by multiple levels of complexity as follow:

Subsample randomly from the genome, 500 bp instances, non-overlapping positive ranges

Select nonreplicating ACS motifs, extend them to 500 bp, nonoverlapping positive ranges

Random-Neg **O**riDB

ACS-Neg **O**riDB

ACS-Neg dataset: presence of ACS matches in both origin and non-origins, making discrimination more challenging

The aim: identifying discriminative factors beyond ACS motifs.

**O**riDB

Shuffling randomly origin sequences to generate negative samples, breaking both **local** and **global** patterns

Shuffling randomly blocks of 5 bp of origin sequences to generate negative samples, disrupting only the **global** patterns

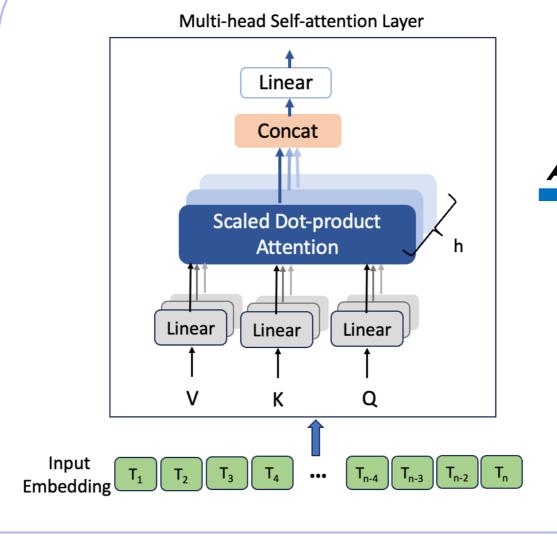
Block-5-Shuffled-Neg

# Fine-tuning **DNABERT** Data Explainability Engineering

## **CONCLUSION:**

- We developed an interpretable pipeline that integrates DNABERT attention maps with the MEME motif discovery tool.
- Motifs identified from high-attention regions validate DNABERT's ability in capturing biologically relevant sequence features through its attention mechanism.

**Attention-based** Motif discovery



- --- origin end Peaks Position in Sequence
- A: Attention scores being extracted from the multi-head attention layer for each sequence,
- B: Motif discovery targets are fragments of 20 bps around peaks (red crosses) of the attention scores.

The aim of subsampling non-origin instances by shuffling origins : ✓ Assessing importance of the nucleotide order for model to identification origins.

## **REFERENCES**

Shuffled-Neg

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