# Context-Aware Regularization with Markovian Integration for Attention-Based Nucleotide Analysis CARMANIA

M.S. Refahi , M. Abavisani, B.A. Sokhansanj , J.R. Brown and G.L. Rosen







## Motivation: DNA is a Language (with Challenges)

- DNA as Language: Genomic sequences can be treated like natural language
  - Long-Range Dependencies: Many genomic features depend on far-apart elements (enhancer-gene links have a median distance of ~28 kb and can extend up to ~1 Mb)
  - **Long-Context Information:** Some features require large genomic span (e.g., biosynthetic gene clusters).
  - **Repetitive Elements:** Huge portions of genomes are repeats (e.g. **~50% in humans' genome**)

Review Article | Published: 29 November 2011

Repetitive DNA and next-generation sequencing: computational challenges and solutions

Todd J. Treangen & Steven L. Salzberg ☑

Nature Reviews Genetics 13, 36–46 (2012) | Cite this article

34k Accesses | 1536 Citations | 129 Altmetric | Metrics

Article Open access Published: 25 September 2023

# Genome-wide enhancer-gene regulatory maps link causal variants to target genes underlying human cancer risk

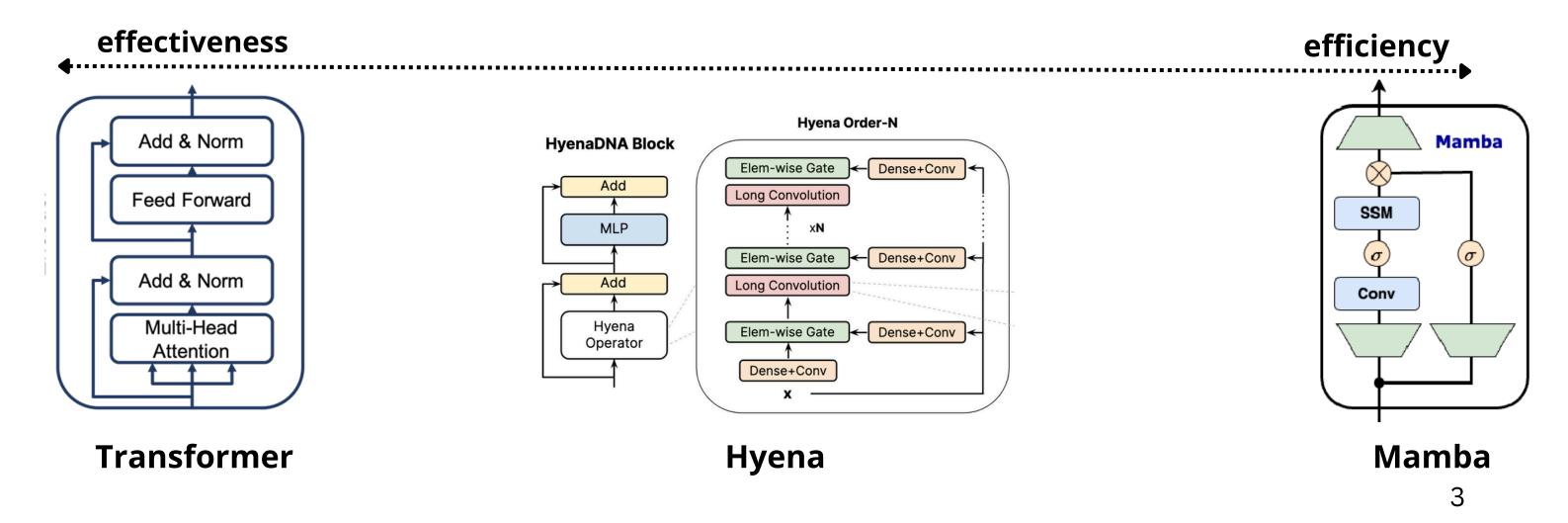
Pingting Ying, Can Chen, Zequn Lu, Shuoni Chen, Ming Zhang, Yimin Cai, Fuwei Zhang, Jinyu Huang, Linyun Fan, Caibo Ning, Yanmin Li, Wenzhuo Wang, Hui Geng, Yizhuo Liu, Wen Tian, Zhiyong Yang, Jiuyang Liu, Chaoqun Huang, Xiaojun Yang, Bin Xu, Heng Li, Xu Zhu, Ni Li, Bin Li, ... Xiaoping Miao 

+ Show authors



## Genomic Language Models

Model	Time Complexity	Max Seq Length	Models
Transformer	O(n²)	<4-10k bp	NT, DNABERT2, glM2, MetaBerta
Hyena	O(n log n)	160k-1m bp	HyenaDNA, EVO
Mamba	O(n)	131k bp	Caduceus



## **LLMs Don't Count! But Biology Needs Counting**

ChatGPT 5 V

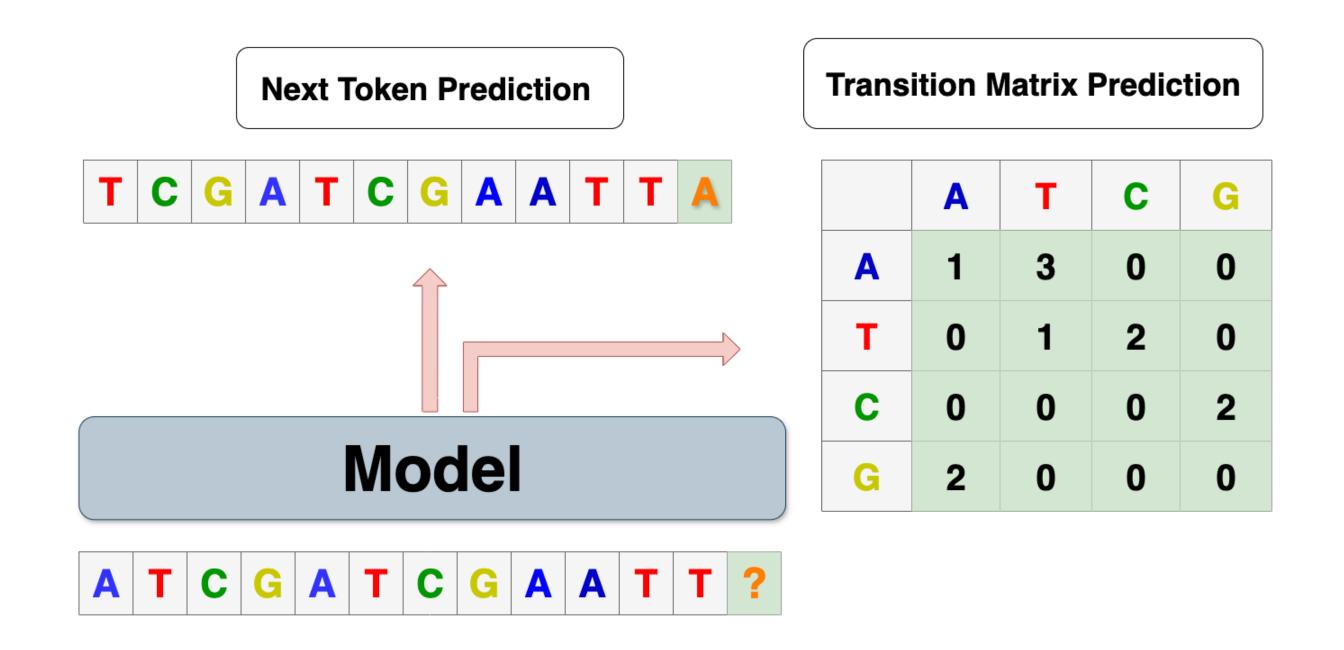
Write a meaningful paragraph that exactly includes at least 10 DNA, 10 protein, and 10 RNA.

In modern biology, DNA serves as the foundational blueprint for life, encoding the instructions that guide the synthesis of RNA and protein. Within every cell, DNA replication ensures the faithful transmission of genetic material, while RNA transcription translates these DNA sequences into functional messages.

Messenger RNA, transfer RNA, and ribosomal RNA each play crucial roles in interpreting DNA information and guiding protein assembly. The resulting protein molecules fold into intricate shapes that define their biological functions. Enzymatic protein complexes regulate DNA repair and RNA processing, ensuring stability across generations. Structural protein frameworks provide cellular integrity, while signaling protein molecules control responses to environmental cues. Ultimately, the interplay among DNA, RNA, and protein forms a dynamic triad—each DNA, RNA, and protein interaction sustaining the continuity and adaptability of life.

**百 6 7 1 2 …** 

#### Can We Force an LLM to Behave like a Word Counter?



## Transition Matrix Loss: Teaching the Model to Count

• **Ground-truth transition matrix:** Row-normalized bigram frequency matrix from the actual sequence.



	A	Т	C	G	
A	5	5	5	5	
T	0	1	1	1	
С	5	1	2	2	
G	1	1	2	1	

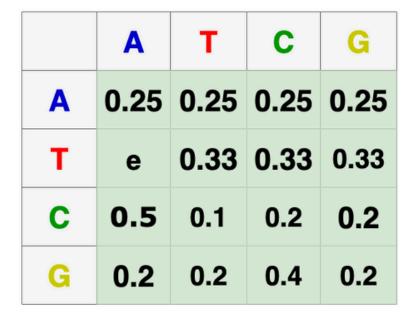
	A	Т	С	G
A	0.25	0.25	0.25	0.25
T	е	0.33	0.33	0.33
С	0.5	0.1	0.2	0.2
G	0.2	0.2	0.4	0.2

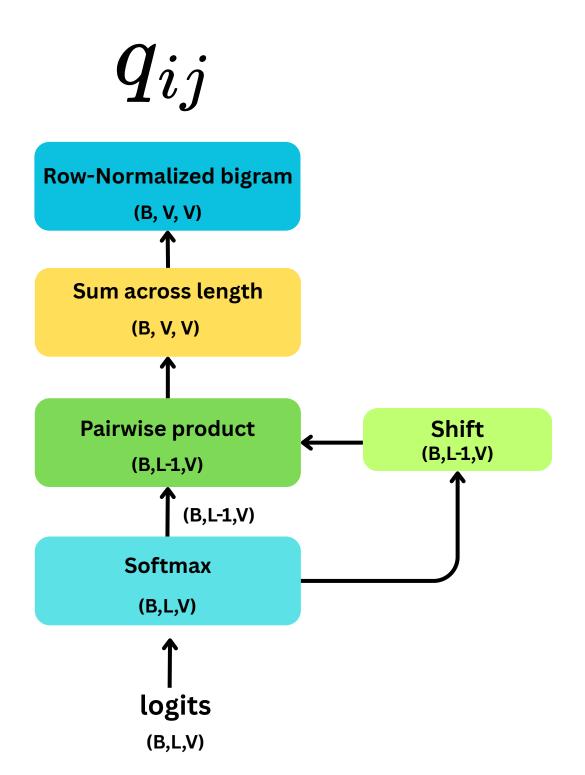
### Transition Matrix Loss: Teaching the Model to Count

- **Ground-truth transition matrix:** Row-normalized bigram frequency matrix from the actual sequence.
- **Predicted transition matrix:** Computed from model outputs via pairwise probability products.

## $p_{ij}$

	A	Т	C	G	
A	5	5	5	5	
T	0	1	1	1	
С	5	1	2	2	
G	1	1	2	1	





### Transition Matrix Loss: Teaching the Model to Count

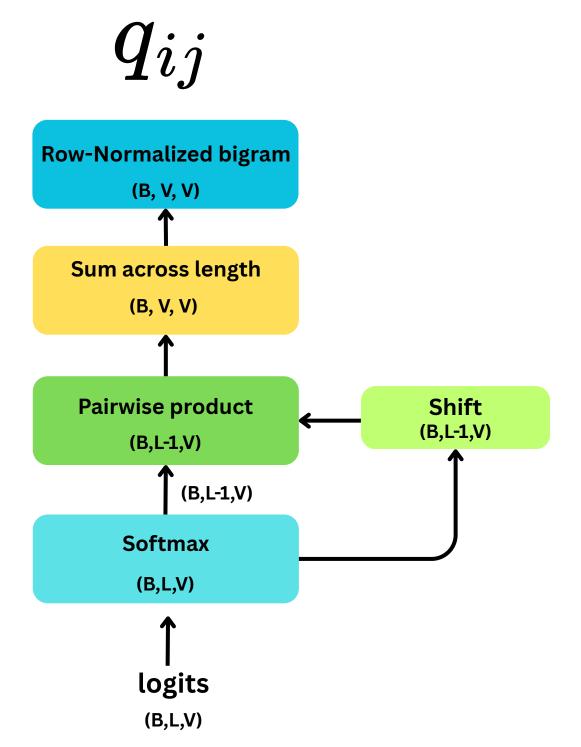
- **Ground-truth transition matrix:** Row-normalized bigram frequency matrix from the actual sequence.
- **Predicted:** Computed from model outputs via pairwise probability products.
- KL divergence loss:

$$\mathcal{L}_{ ext{TM}} = \sum_{i,j} p_{ij} \log rac{p_{ij}}{q_{ij}}.$$

 $p_{ij}$ 

	A	T	С	G
A	5	5	5	5
Т	0	1	1	1
С	5	1	2	2
G	1	1	2	1

	A	T	С	G
A	0.25	0.25	0.25	0.25
Т	е	0.33	0.33	0.33
С	0.5	0.1	0.2	0.2
G	0.2	0.2	0.4	0.2



### Markovian Nature of In-Context Learning in Transformers

#### **Key Point:**

**In-context learning** in **Transformers** often follows a **Markovian** pattern .

The model primarily depends on **a few preceding tokens**, not the entire sequence history.

Enables use of sliding-window attention instead of full attention ?!

#### From Self-Attention to Markov Models: Unveiling the Dynamics of Generative Transformers

M. Emrullah Ildiz<sup>1</sup> Yixiao Huang<sup>1</sup> Yingcong Li<sup>1</sup> Ankit Singh Rawat<sup>2</sup> Samet Oymak<sup>1</sup>

<sup>1</sup> University of Michigan, Ann Arbor {eildiz, yingcong, oymak}@umich.edu, yixiao.huang@my.cityu.edu.hk

<sup>2</sup> Google Research NYC ankitsrawat@google.com

#### The Evolution of Statistical Induction Heads: In-Context Learning Markov Chains

Ezra Edelman\*
University of Pennsylvania
ezrae@cis.upenn.edu

Nikolaos Tsilivis\* New York University† nt2231@nyu.edu

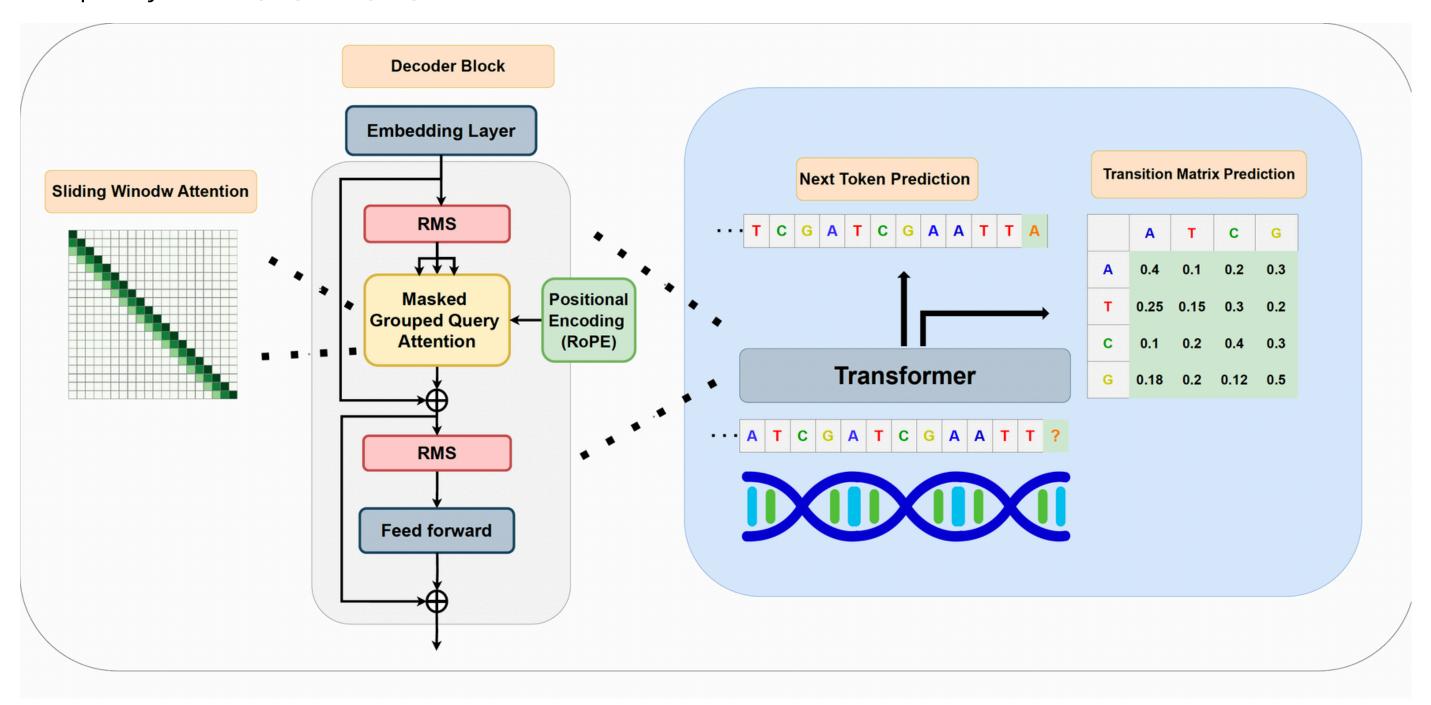
Benjamin L. Edelman
Harvard University
bedelman@g.harvard.edu

Eran Malach
Harvard University
emalach@g.harvard.edu

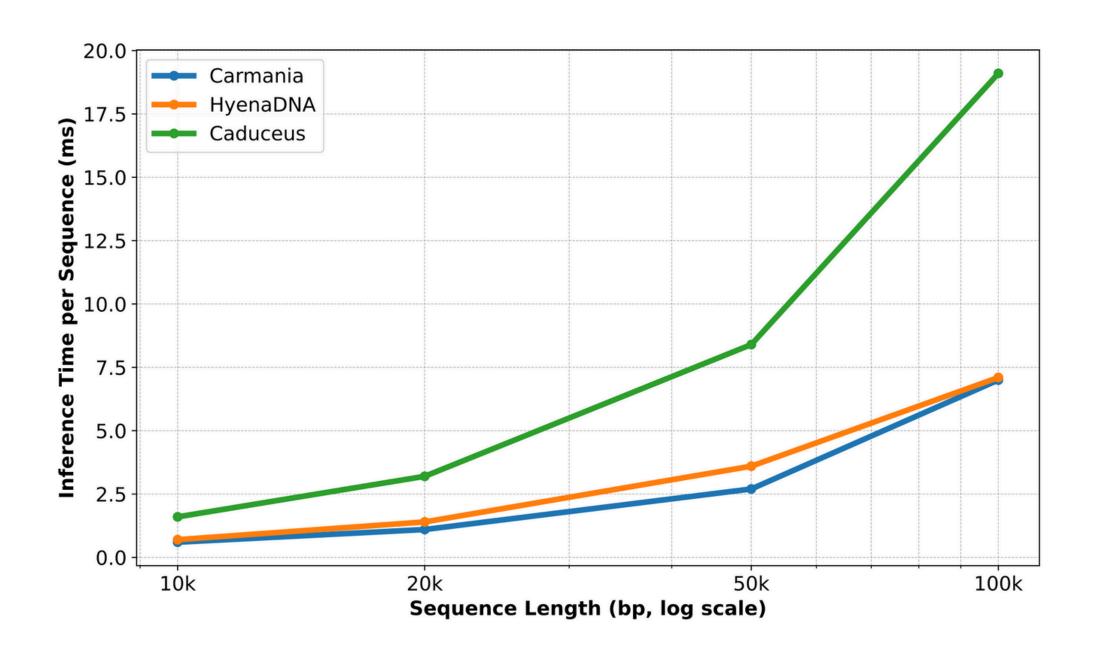
Surbhi Goel University of Pennsylvania surbhig@cis.upenn.edu

#### **CARMANIA: Our Long-Context Genomic LM**

- CARMANIA is a 160k-context genomic Transformer language model.
- This Markovian nature of Transformers aligns with **transition-matrix objectives (TM-Loss).**
- Enables use of **sliding-window attention** instead of full attention.
- Reduces complexity from  $O(N^2) \rightarrow O(kN)$ -

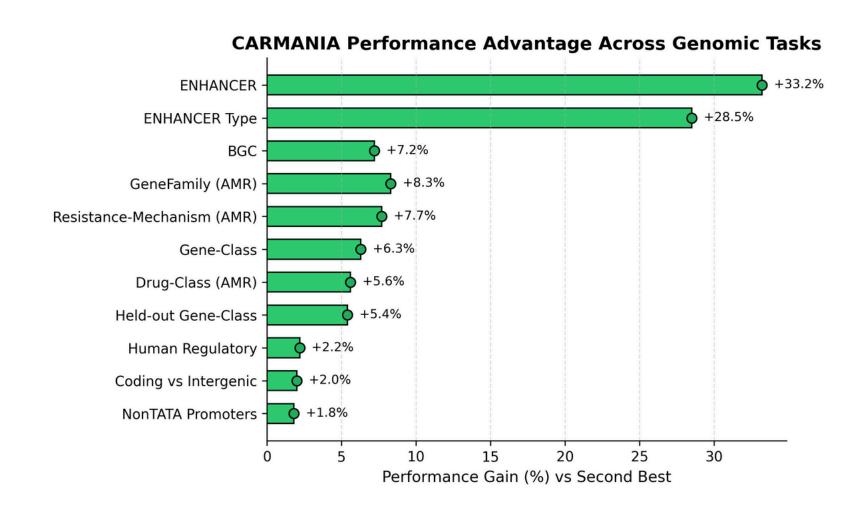


#### **Fast Inference Time!**



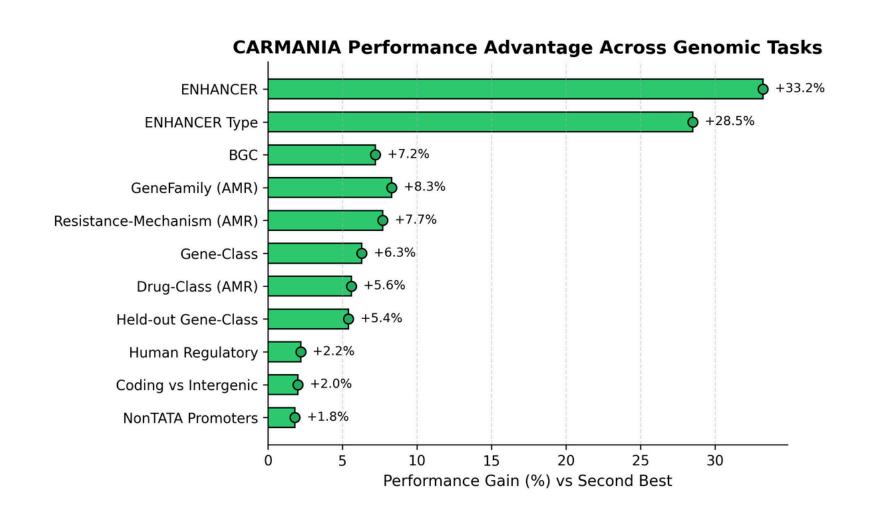
#### **Genomic Task Performance Gains**

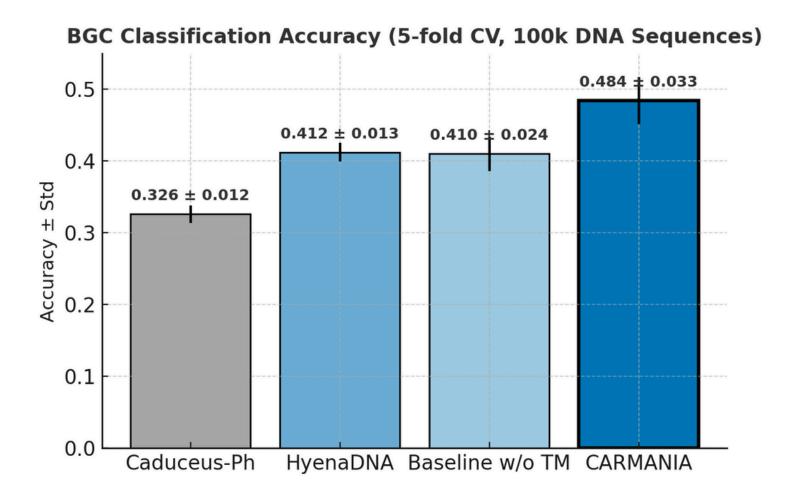
- **Scope:** Compared CARMANIA with top models on 40 genomic tasks.
- Wins: Outperformed competitors on 20 out of 40 tasks.
- Gains: Achieved up to 33% improvement, setting a new state-of-the-art benchmark across all evaluations.



#### **Genomic Task Performance Gains**

- **Scope:** Compared CARMANIA with top models on 40 genomic tasks.
- Wins: Outperformed competitors on 20 out of 40 tasks.
- Gains: Achieved up to 33% improvement, setting a new state-of-the-art benchmark across all evaluations.
- CARMANIA outperforms prior models on **BGC classification** (100,000 DNA sequences).





#### Summary



- 160k context: Longest-context genomic Transformer to date
- TM-loss: Markovian regularization improves global consistency
- Fast inference: 2.5× speedup on long sequences

# Thank you!



Power your pipeline with CARMANIA's 160k-context understanding





