

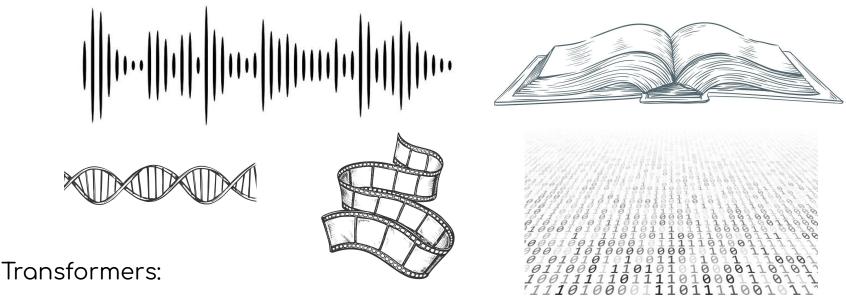


### DeciMamba

Exploring the Length Generalization Potential of Mamba

Assaf Ben-Kish, Itamar Zimerman, Shady Abu-Hussein, Nadav Cohen, Amir Globerson, Lior Wolf, Raja Giryes

### Long Sequences Are All Around Us



- Short Sequences SOTA
- Long Sequences Limited due to quadratic complexity w.r.t input length

### The Long Sequence Architecture Zoo

Sub-Quadratic Recurrent Models

State-Space Models

Hyena

You Are Here RWKV

Griffin

•••

Transformer-Based Solutions

Flash Attention

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Longformer

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**Activation Beacon** 

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### Mamba

A promising long sequence model, tailored for auto-regressive sequence prediction:

- ✓ Parallel training, sequential inference
- ✓ Scales linearly w.r.t input sequence length
- ✓ Performance on short-sequence tasks matches Transformers in important modalities like text.
- ✓ Intuitive sequence model
- ✓ Hardware-Aware implementation
- X Length Generalization

### Studying Length Generalization via Passkey Retrieval

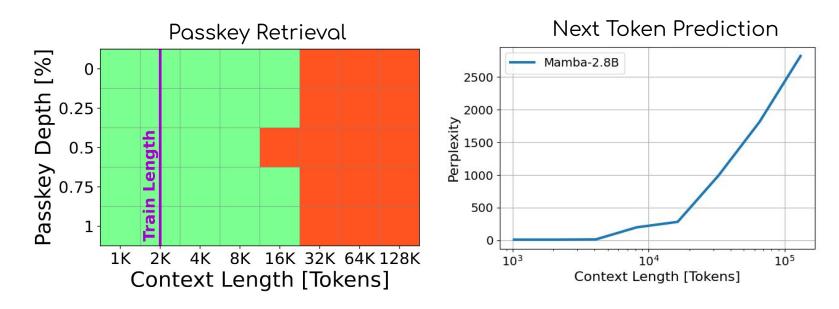
#### Sample from the Passkey Retrieval task:

There is an important pass key hidden inside a lot of irrelevant text. Find it and memorize it:

The Boston Celtics are an American professional basketball team based in Boston. The Celtics compete in the National Basketball Association (NBA) as a member of the Atlantic Division of the Eastern Conference. Founded in 1946 as one of the league's original eight teams, the Celtics play their home games at TD Garden.....The pass key is 51928. Remember it. 51928 is the pass key.....The Celtics are regarded as one of the most successful teams in NBA history and hold the records for most NBA championships won, with 18, and most recorded wins of any NBA franchise. Answer: 51928

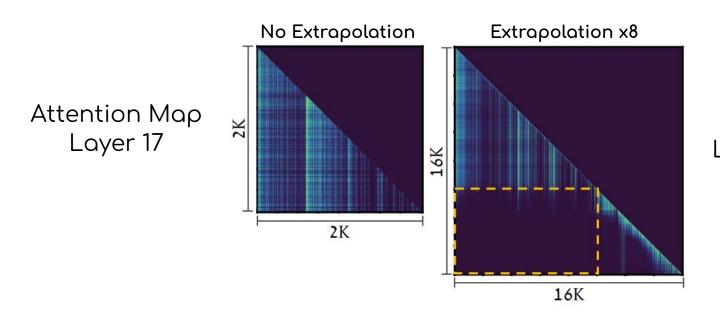
### Mamba Does Not Length Generalize Well

When trained on sequences of 2K tokens:



Identifying The Problem

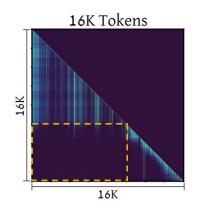
### Mamba's Hidden Attention During Extrapolation



Limited Effective Receptive Field

### Why Do Limited ERFs Happen?

$$\begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_L \end{bmatrix} = \begin{bmatrix} C_1 \bar{B}_1 & 0 & \cdots & 0 \\ C_2 \bar{A}_2 \bar{B}_1 & C_2 \bar{B}_2 & \cdots & 0 \\ \vdots \\ C_L \Pi_{k=2}^L \bar{A}_k \bar{B}_1 & C_L \Pi_{k=3}^L \bar{A}_k \bar{B}_2 & \cdots & C_L \bar{B}_L \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_L \end{bmatrix}$$



$$\prod_{k=j+1}^{L} \bar{A}_k = \prod_{k=j+1}^{L} \exp(A\Delta_k) = \exp(A\sum_{k=j+1}^{L} \Delta_k)$$

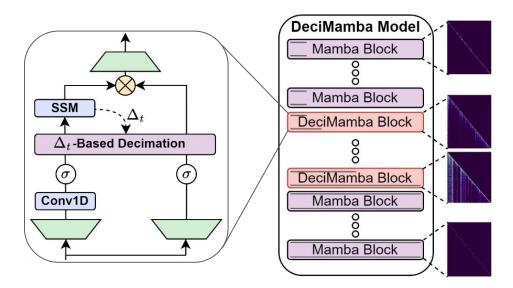
The delta sum is bounded by the training sequence length L → The learnt decay rate A is not constrained to support longer sums.

### DeciMamba

(Decimating-Mamba)

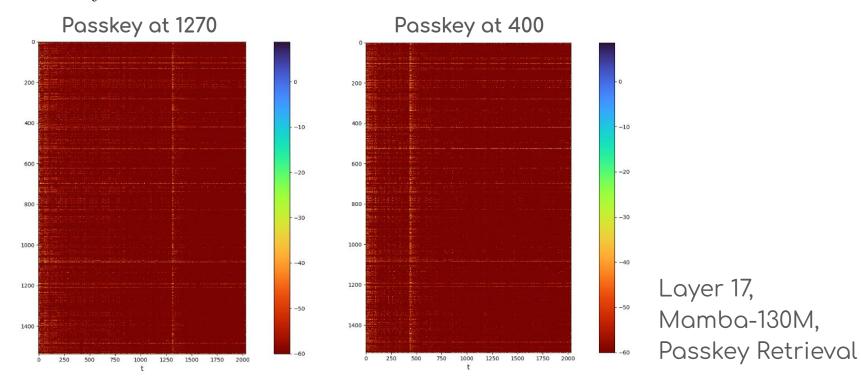
### DeciMamba - Algorithm

- (i) Assign token importance score
- (ii) Perform global pooling (Decimation)
- (iii) Feed the State-Space Model (SSM) with the pooled sequence



### $\Delta_t$ = Token Importance Score

#### Values of $\Delta_t$ per channel:



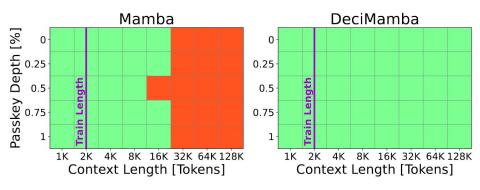
# Results

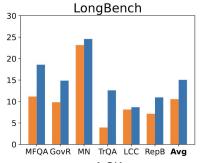
#### Computation Speed [Sec]

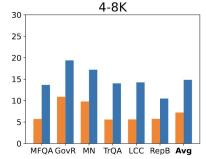
### Results

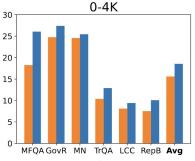
Model \ Context Length	8k	16k	32k	64k	128k	256k	512k
Mamba-130m	0.17	0.3	0.69	1.27	2.12	4.03	8.17
DeciMamba-130m	0.14	0.19	0.31	0.59	1.08	2.01	4.22

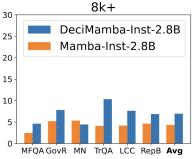
- ✓ Extension by more than a magnitude
- ✓ Real-world long-range NLP tasks
- ✓ Works in zero-shot
- ✓ Less computational resources
- ✓ No re-training







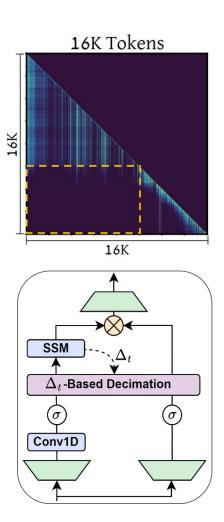




## Recap

### Recap

- Mamba is a promising long context model
- Mamba does not length generalize well due to chronic overfitting to the training length
- DeciMamba overcomes this limitation by pooling the sequence in global layers
- DeciMamba extends the context in both synthetic and real-world long context tasks



### Thank You!



DeciMamba repo on Github