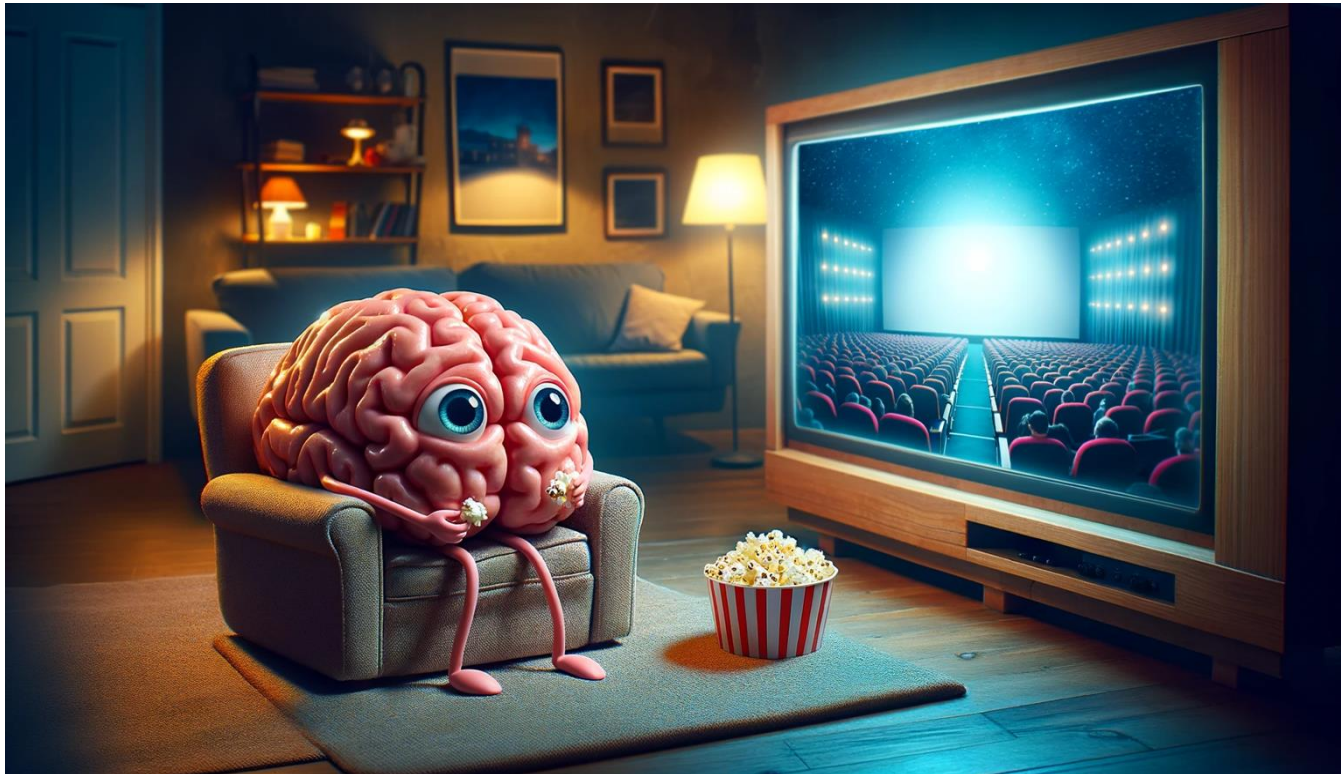


From movie frames to the brain...



This image was created with the assistance of DALL-E

... and back!

Florian David
Neuro-X student

*Supervised by
Michael Chan*

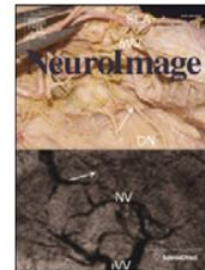
June 11th, 2024



Contents lists available at [ScienceDirect](http://www.sciencedirect.com)

NeuroImage

journal homepage: www.elsevier.com/locate/ynimg



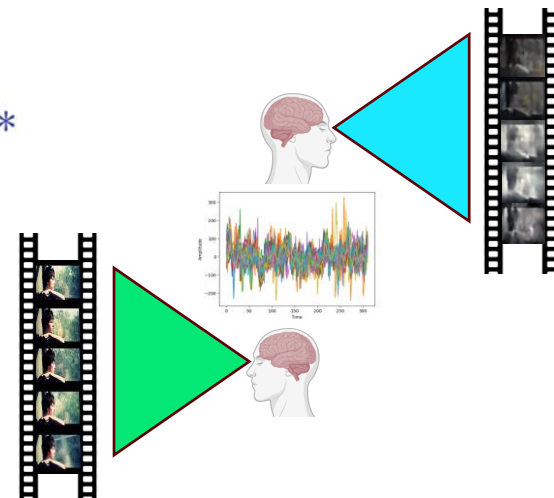
Review

Encoding and decoding in fMRI

Thomas Naselaris^a, Kendrick N. Kay^b, Shinji Nishimoto^a, Jack L. Gallant^{a,b,*}

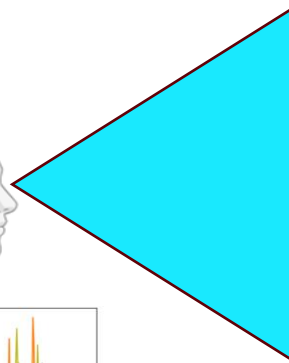
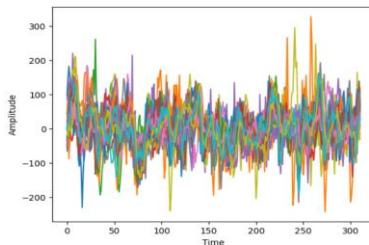
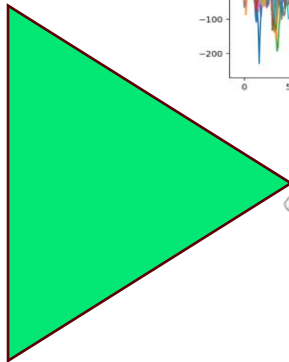
^a Helen Wills Neuroscience Institute, University of California, Berkeley, CA 94720, USA

^b Department of Psychology, University of California, Berkeley, CA 94720, USA



1.

Build an **encoder** to predict brain activity while someone is watching a movie



2.

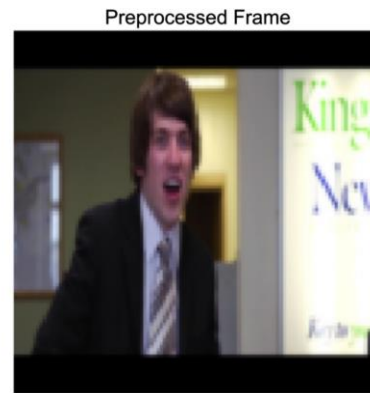
Build a **decoder** to validate the predicted brain activity

3.

Merge both models into an end-to-end **encoder-decoder** to generate fMRI data on unseen movies

Preprocessing on all 14 movies:

- 112x112 pixels
- 32 frames per TR



Splitting of 13/14 movies



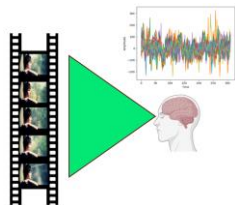
Train Set ~ 64%

Validation Set ~ 16%

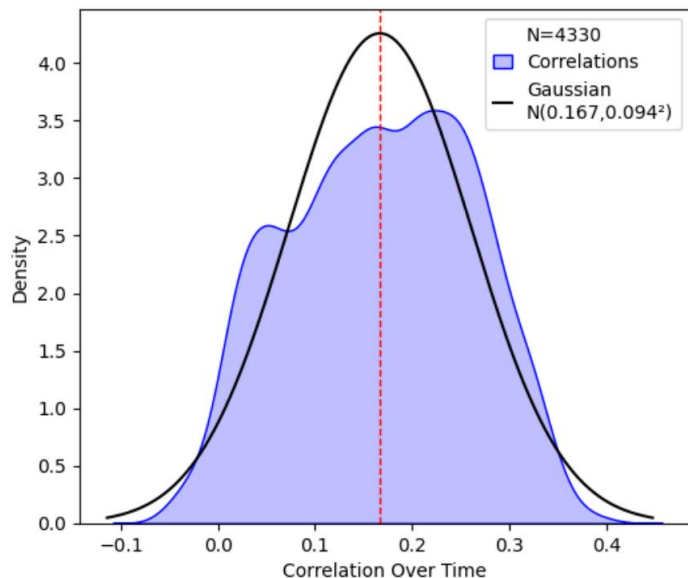
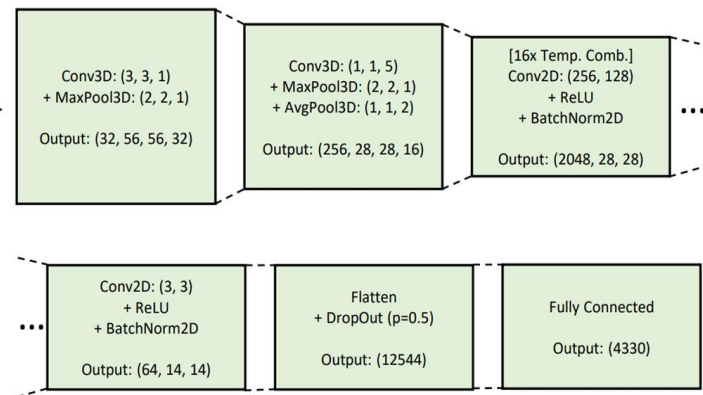
Test Set ~ 20%

1 full movie in the Test Set:
You Again (613TR)





Input: (3, 112, 112, 32)

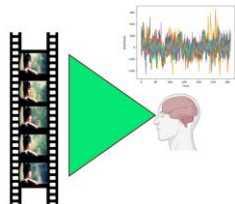


⇒ Best results on the **validation set**

ML4Science results

- Average correlations on the **validation set**: 0.167
- Average correlations on the **test set**: 0.001

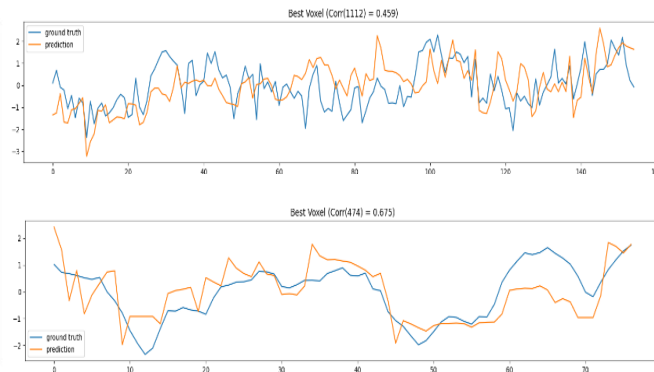
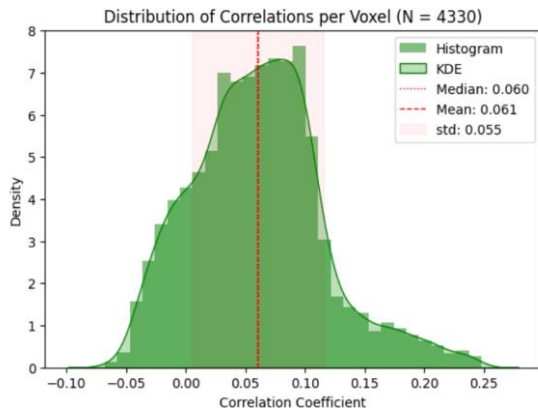
Can we improve it?



Improvement of the Encoder

- **1 subject** → **average activity across 30 subjects**
- batch size = 1 → batch size = 16
- other hyperparameters

Global

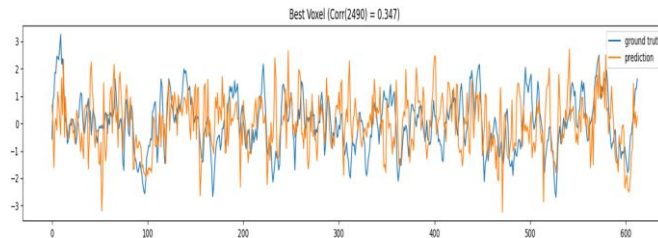
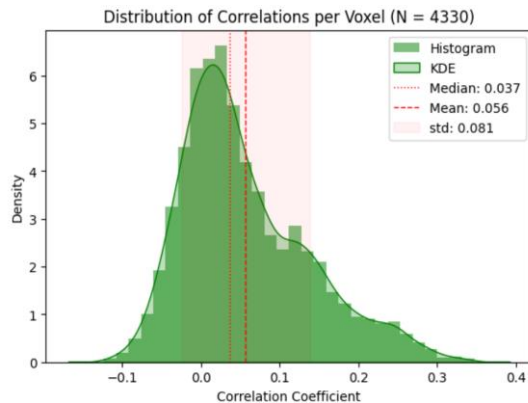


Correlations over time

How well each voxel has been predicted

- Mean correlations: **0.061**
- Median correlations: 0.060

You Again

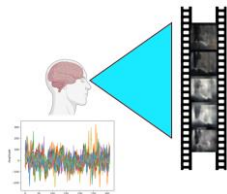


Correlations over time

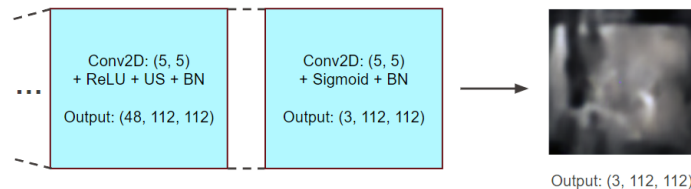
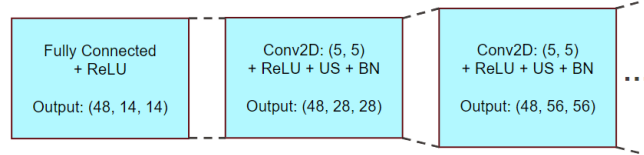
How well each voxel has been predicted

- Mean correlations: **0.056**
- Median correlations: 0.037

Decoder



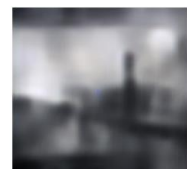
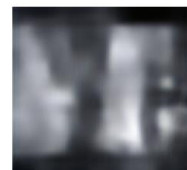
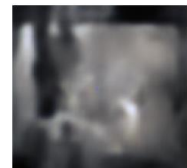
Input: (4330,)



Left: Simplified frame



Right: Predicted frame



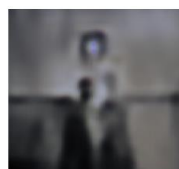
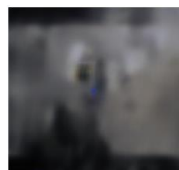
Payload



Payload



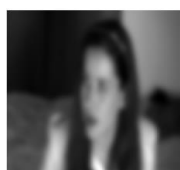
After The Rain



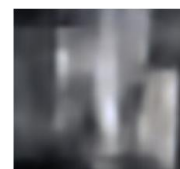
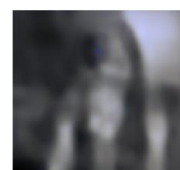
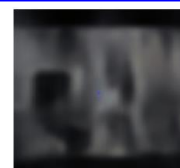
Between Viewings



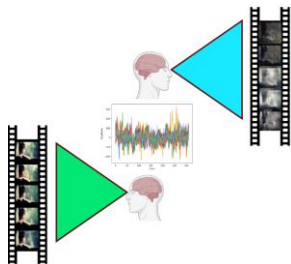
First Bite



Chatter



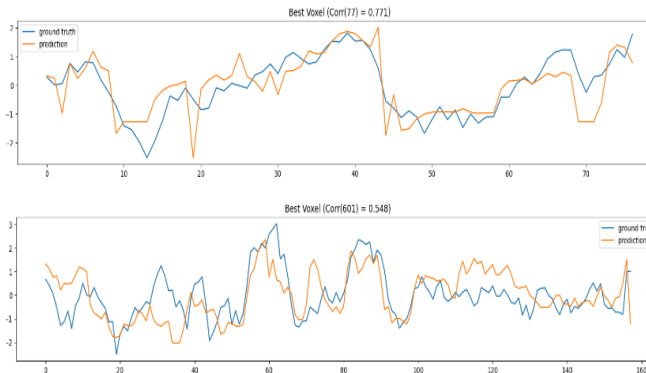
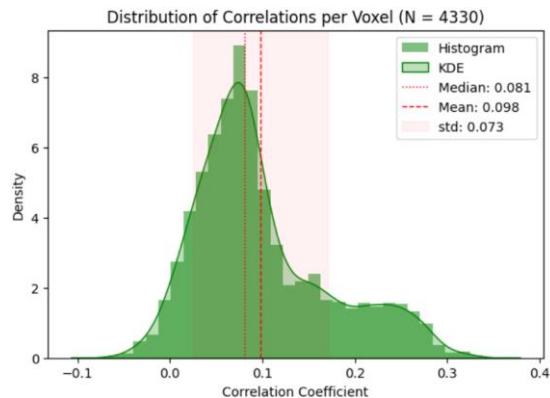
→ The decoder can be seen as a regularizer for the encoder



Model training

- End-to-end training: all the weights are updated at each iteration
- Objective:
 - minimize the encoder loss
 - minimize **MSE**
 - minimize **cosine distance**
 - minimize the decoder loss
 - maximize **perceptual similarity**
 - maximize **structural similarity**
 - minimize **total variation**

Global

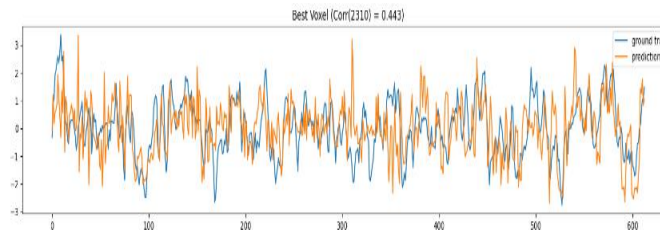
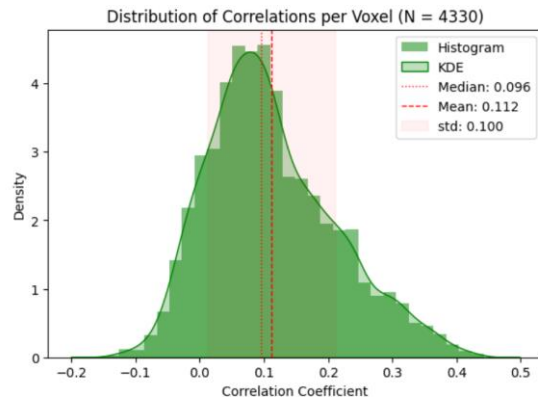


Correlations over time

How well each voxel has been predicted

- Mean correlations: **0.098**
- Median correlations: 0.081

You Again



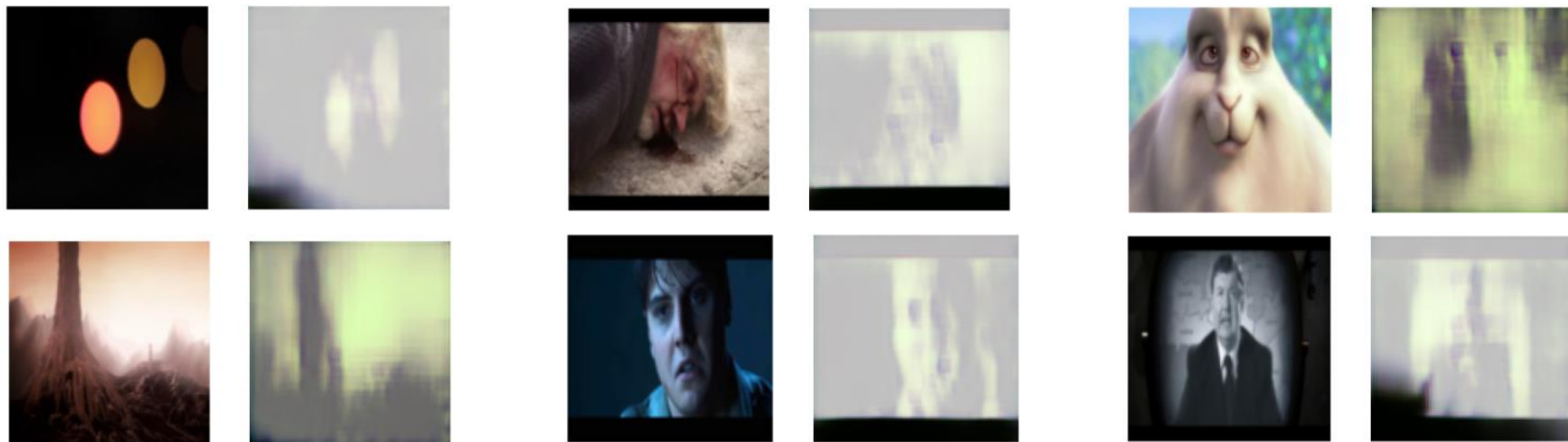
Correlations over time

How well each voxel has been predicted

- Mean correlations: **0.112**
- Median correlations: 0.96

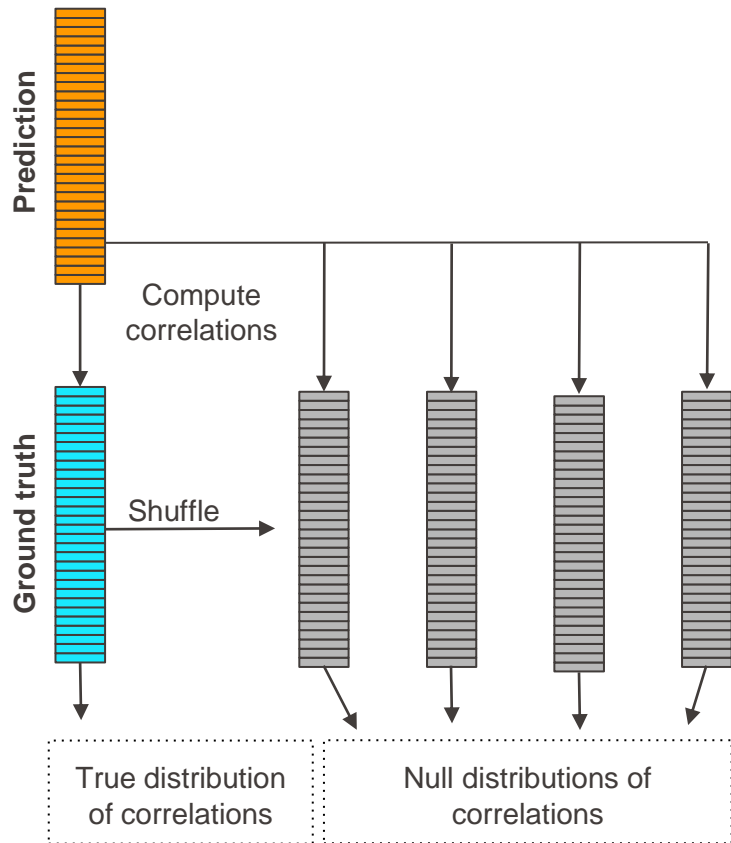
Encoder-Decoder

Global



You Again



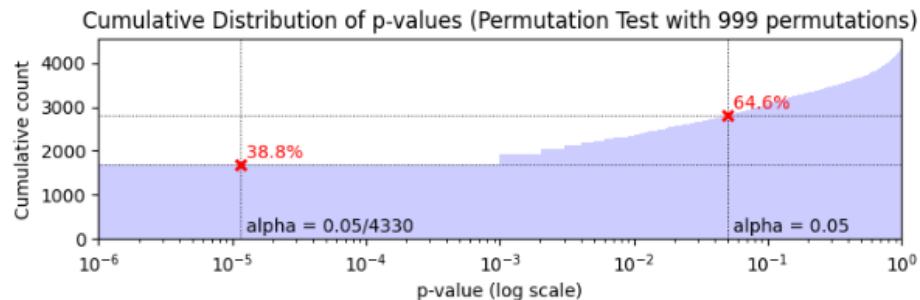
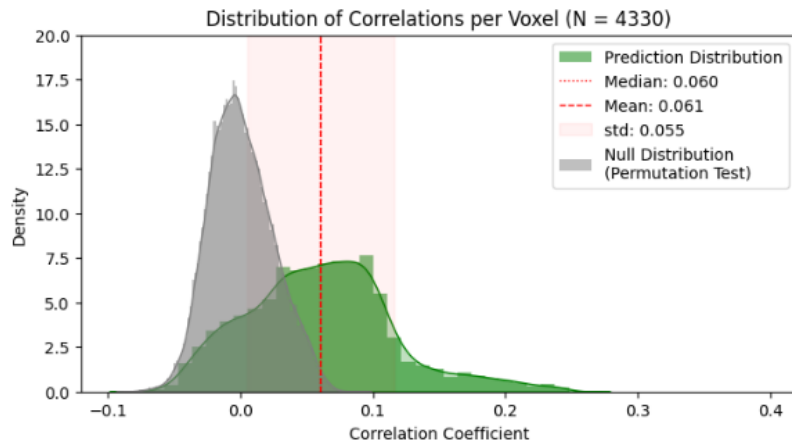


For each voxel:

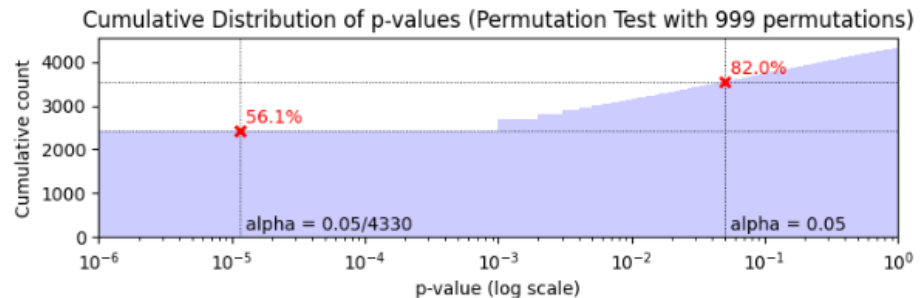
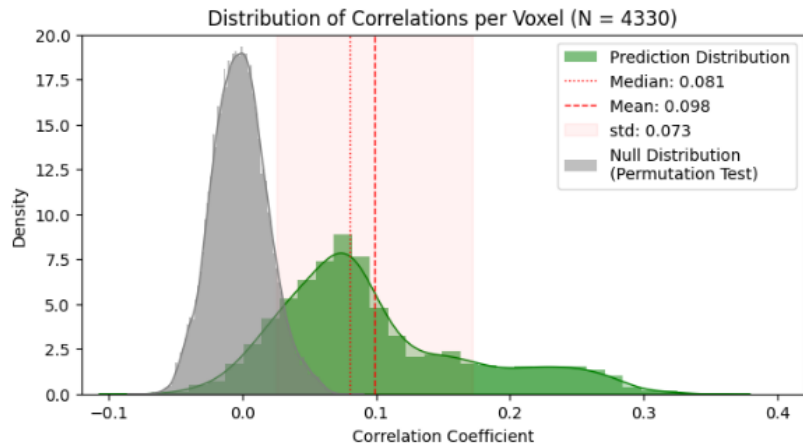
- compute correlations between prediction and ground truth
- compute correlations between prediction and shuffled ground truth multiple times
- p-value: calculate the probability that the true correlation value is lower than the correlation with a shuffled label

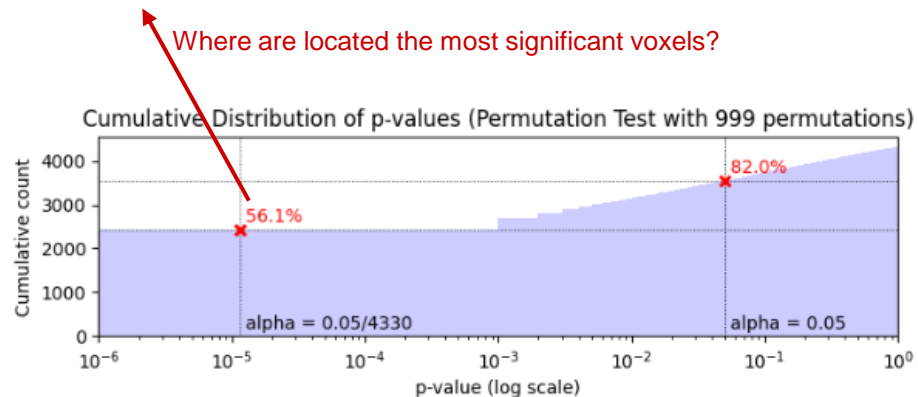
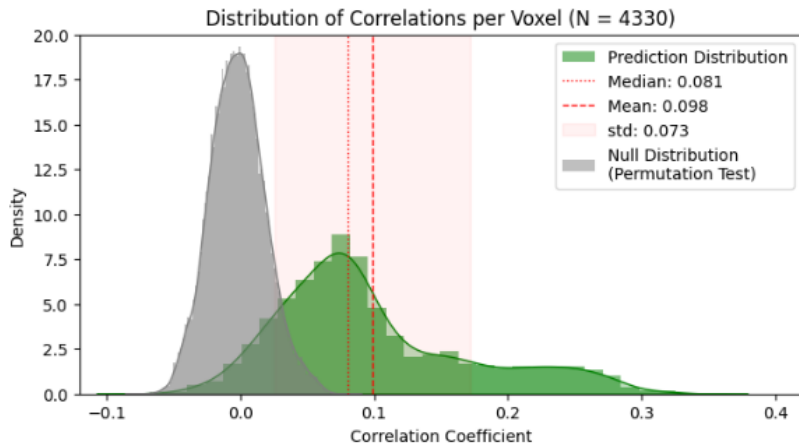
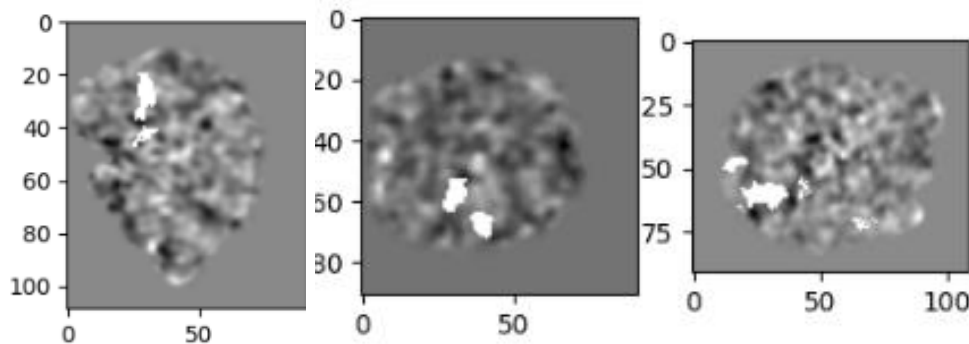
→ **get 1 p-value per voxel**

Encoder



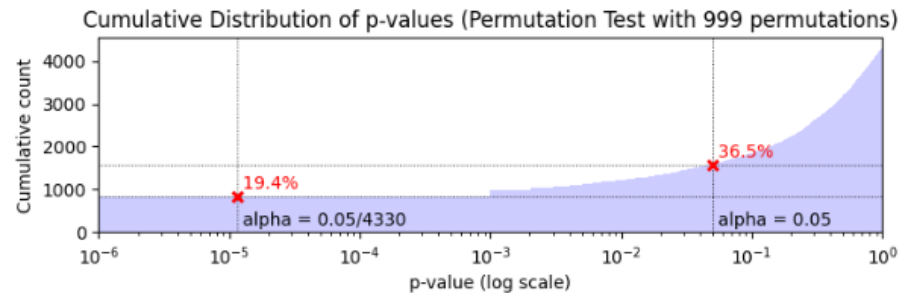
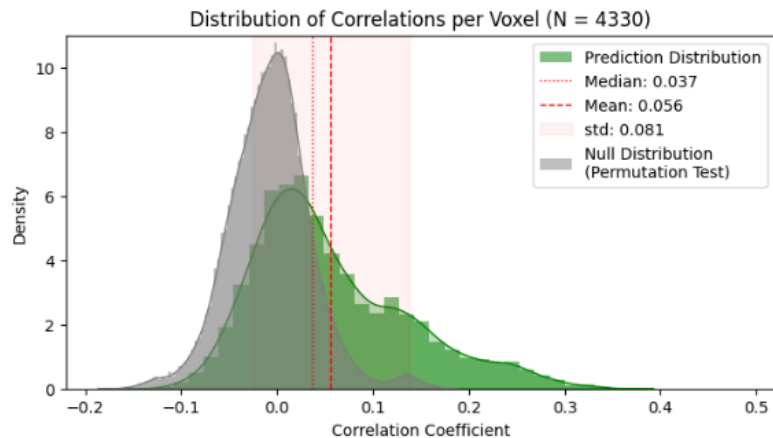
Encoder-Decoder



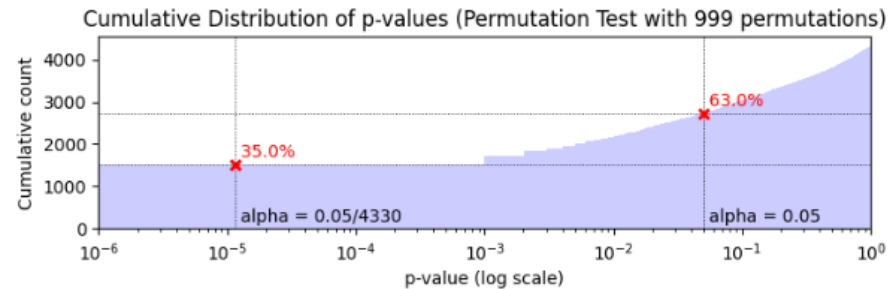
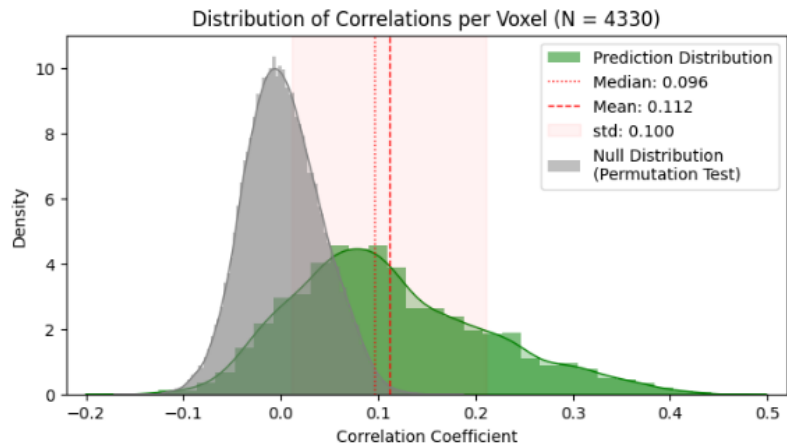


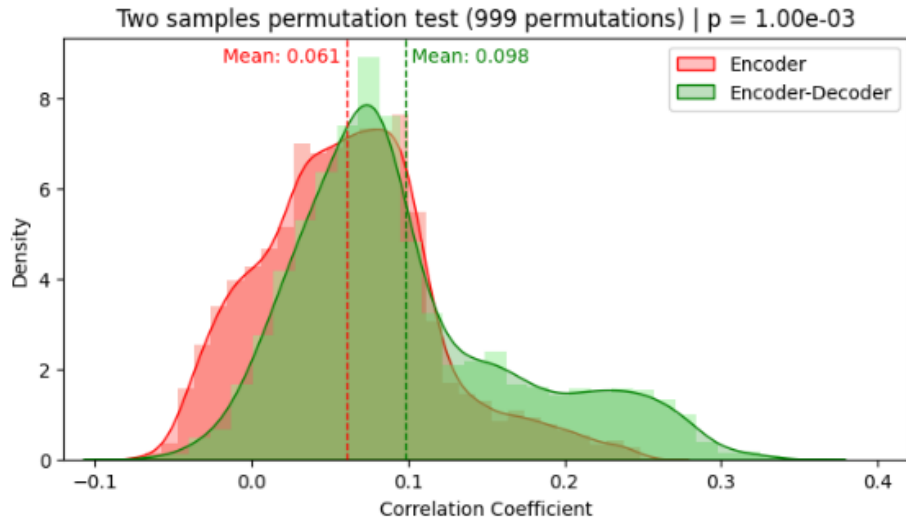
Significance – You Again

Encoder



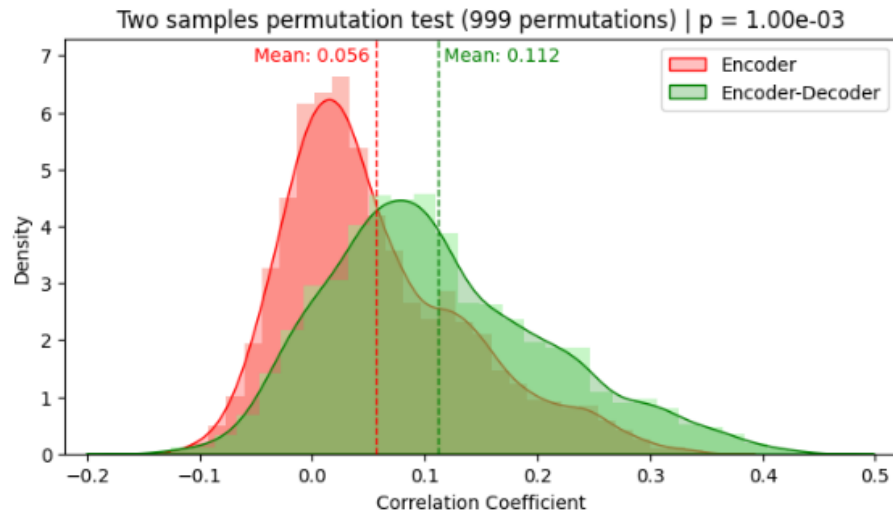
Encoder-Decoder





Global

- Means difference: 0.037
- $p < 0.05$



You Again

- Means difference: 0.056
- $p < 0.05$

Scientific aim(s)

ML4Science results

→ Average correlations on the **test set: 0.001**

Can we improve it?

17

1.

Build an **encoder** to predict brain activity while someone is watching a movie

→ The encoder alone can predict brain activity on an **unseen movie** with an average correlation of **0.056**

2.

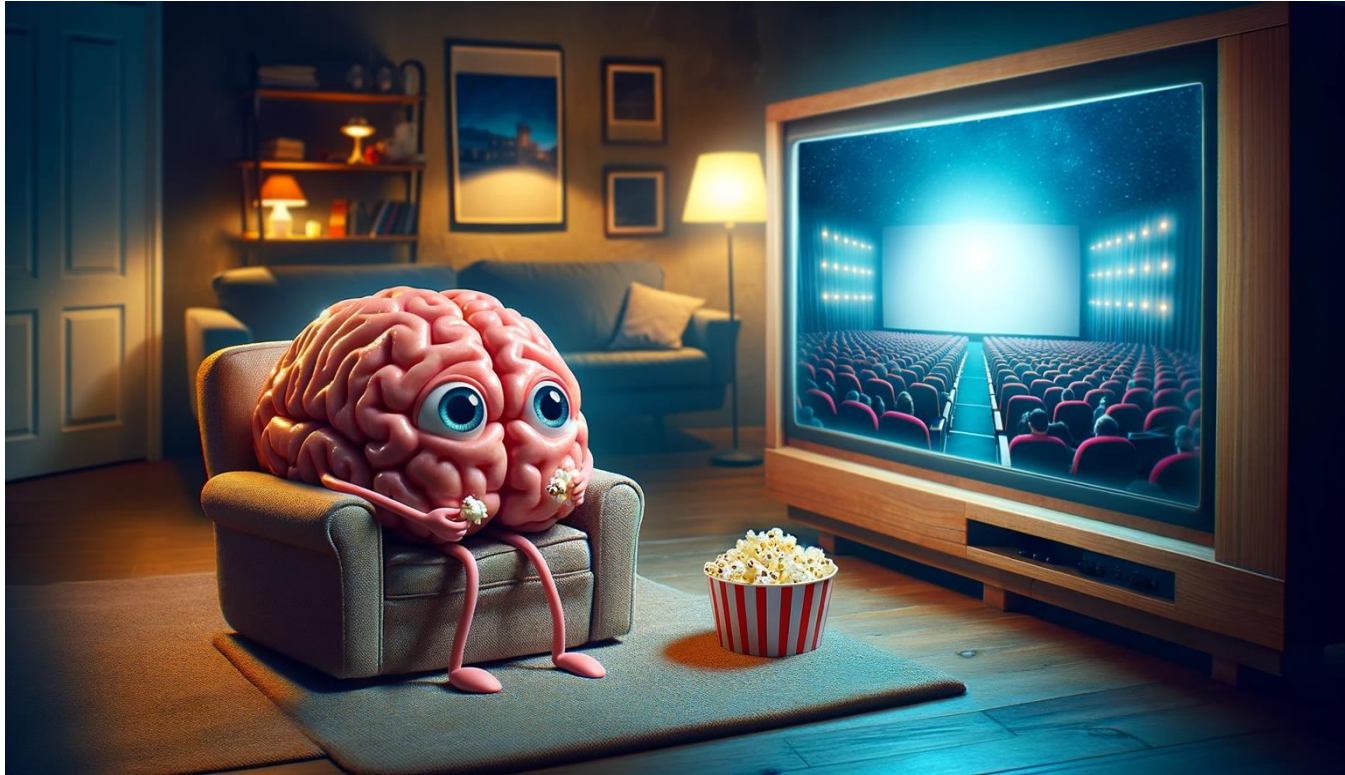
Build an **decoder** to validate the predicted brain activity

→ The decoder alone can learn to reconstruct movie frames from brain activity

3.

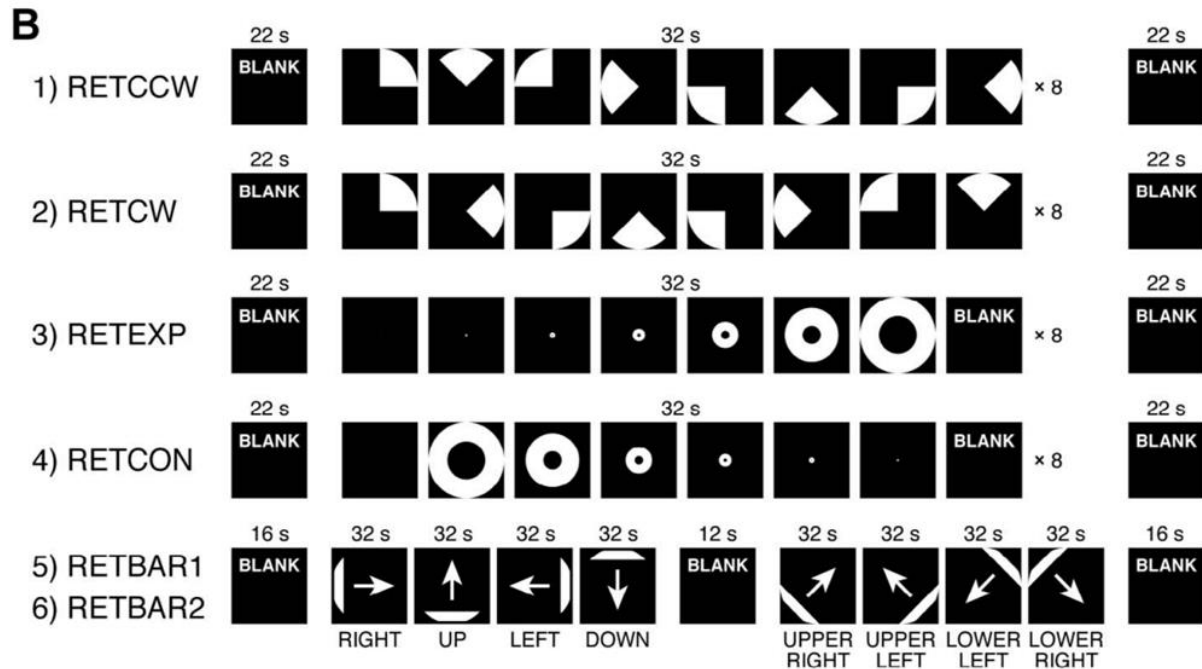
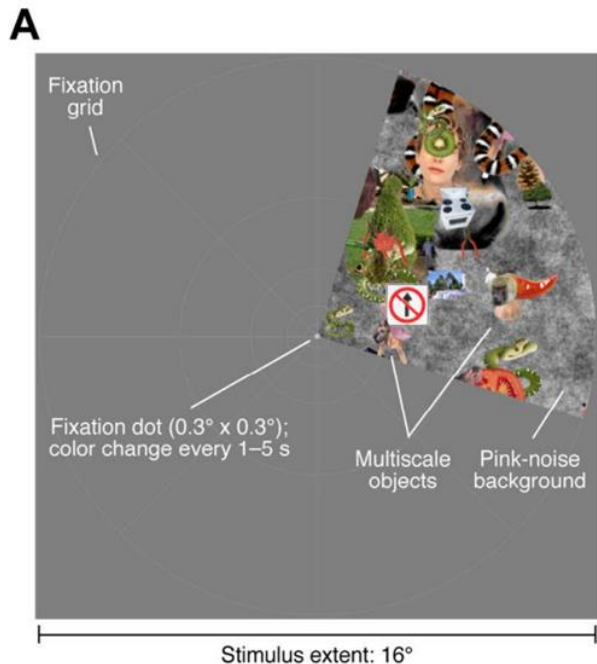
Merge both models into an end-to-end **encoder-decoder** to generate fMRI data on unseen movies

→ The end-to-end encoder-decoder can predict brain activity on an **unseen movie** with an average correlation of **0.112**

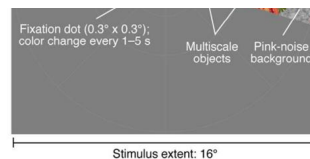
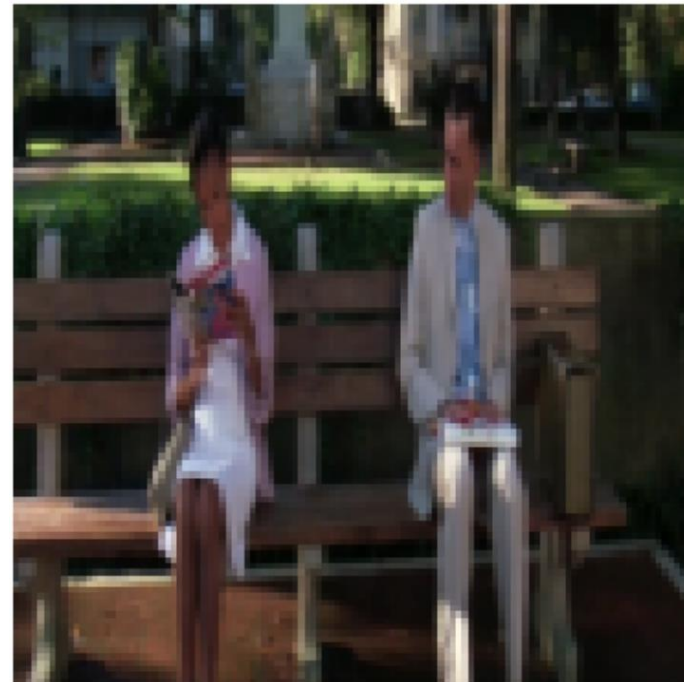
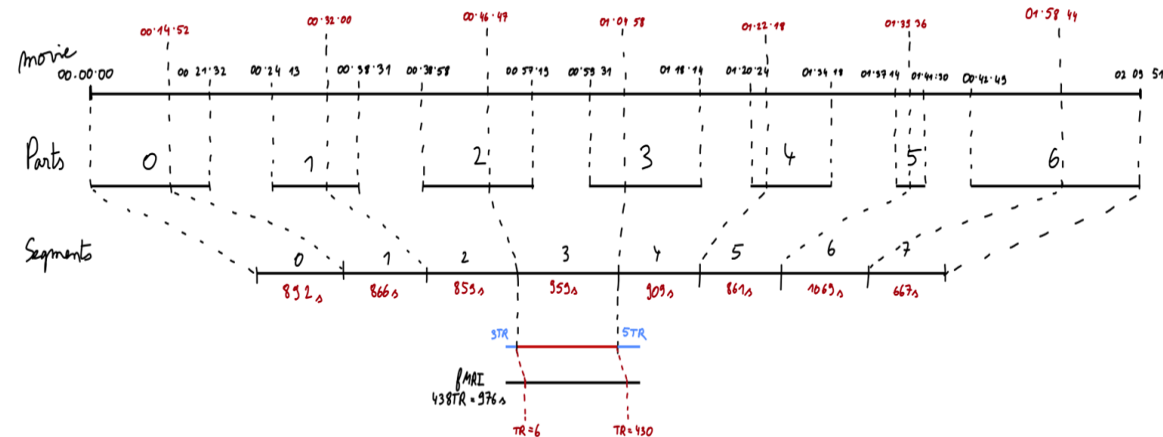


Thank you for your attention!

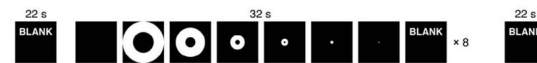
Bonus: what went wrong...



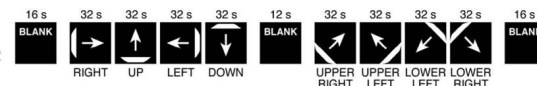
Bonus: what went wrong...



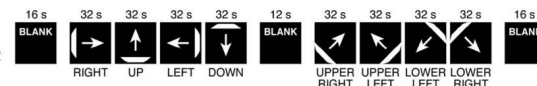
4) RETCON



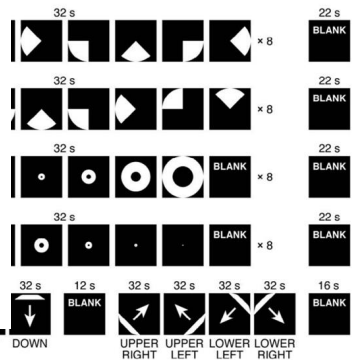
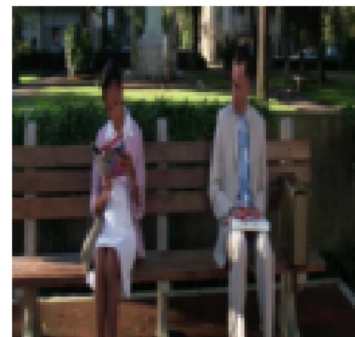
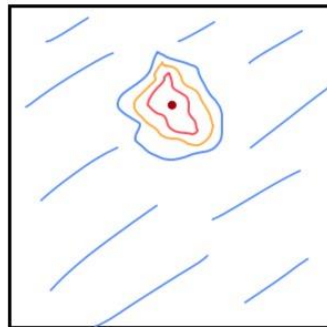
5) RETBAR1



6) RETBAR2

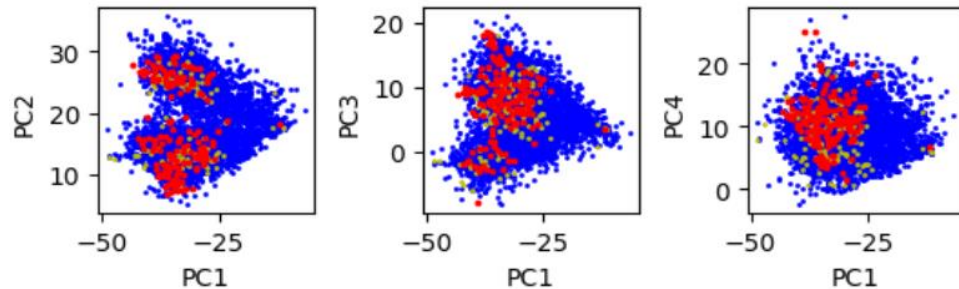


Bonus: what went wrong...

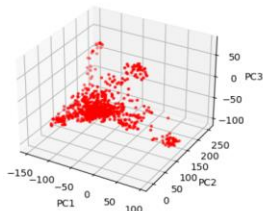


Stimulus extent: 16°

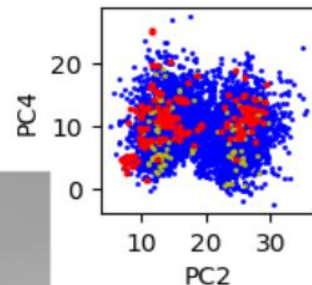
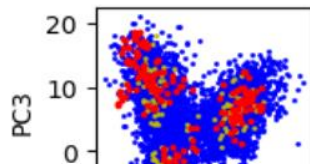
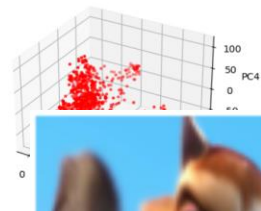
Bonus: what went wrong...



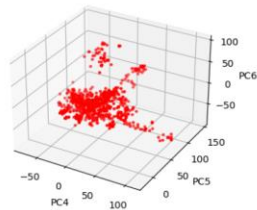
3D Scatter Plot of PC1, PC2 vs PC3



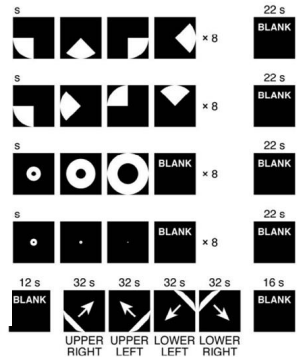
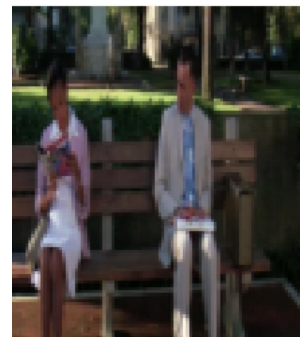
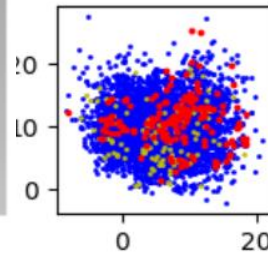
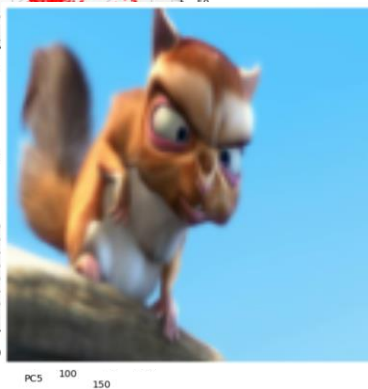
3D Scatter Plot of PC2, PC3 vs PC4



3D Scatter Plot of PC4, PC5 vs PC6



3D



Bonus: what went wrong...

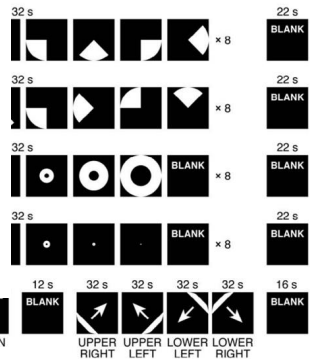
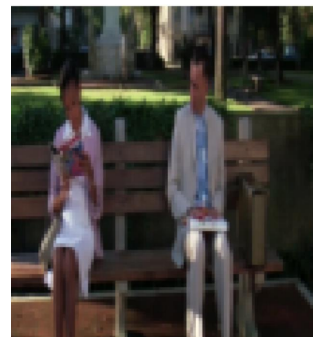
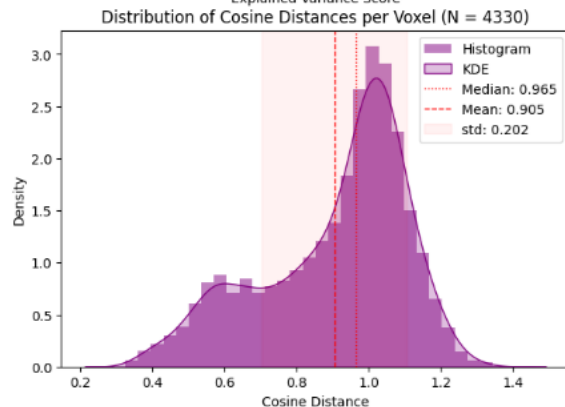
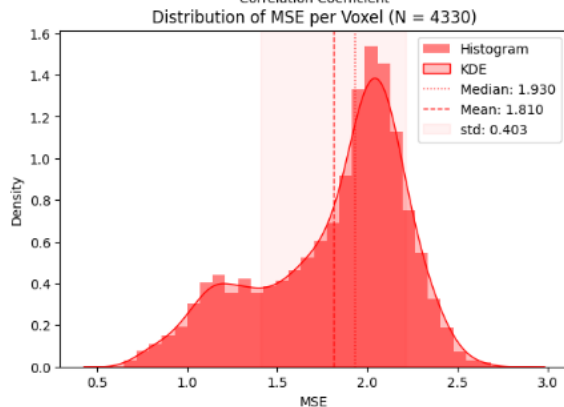
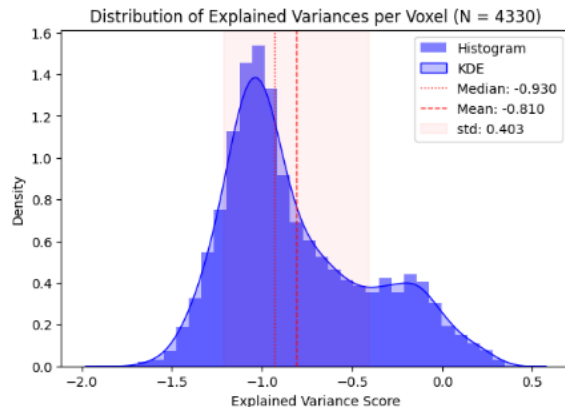
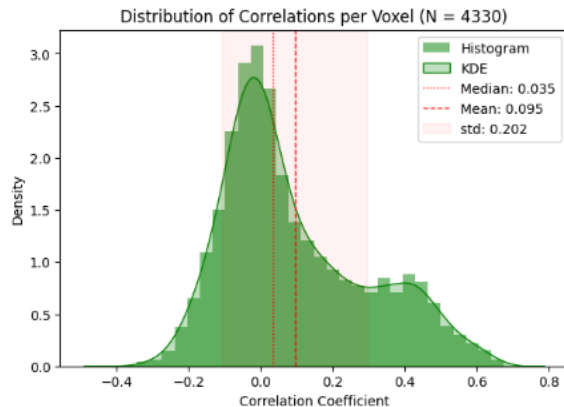
EPFL



- 1. Simplifications**
- Grey scale
 - Gaussian blur

- 2. Feature extraction**
- Pretrained VGG1
 - 4096 features/frame

- 5. Splitting each cluster**
- 80% in trainset
 - 20% in testset

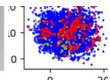
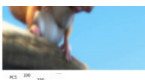
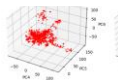
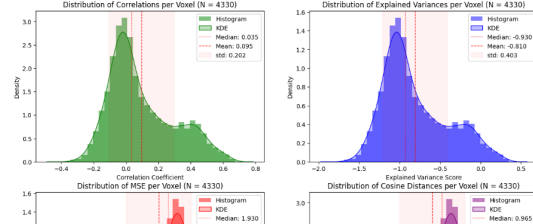


Stimulus extent: 16°

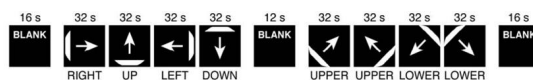
5) RETBAR1
6) RETBAR2



Bonus: what went wrong...



- 5) RETBAR1
6) RETBAR2



went wrong...



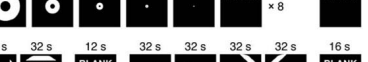
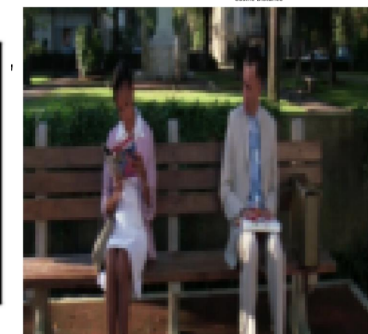
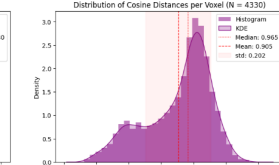
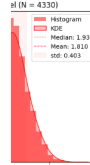
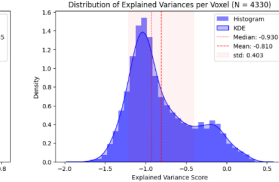
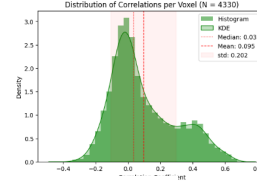
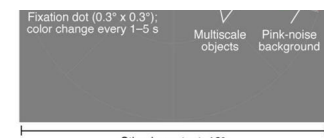
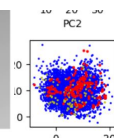
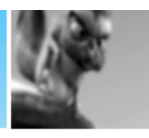
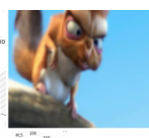
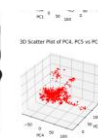
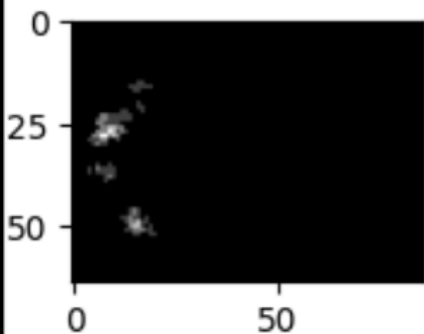
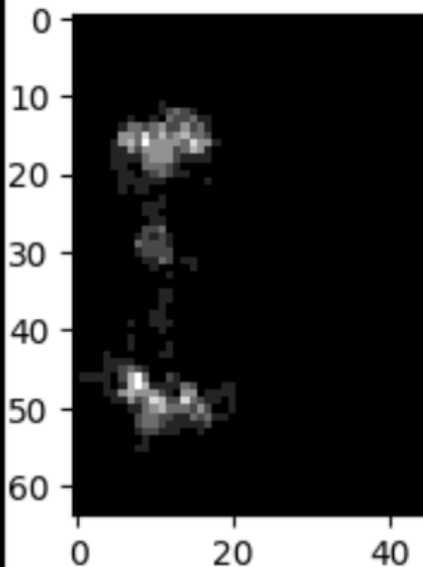
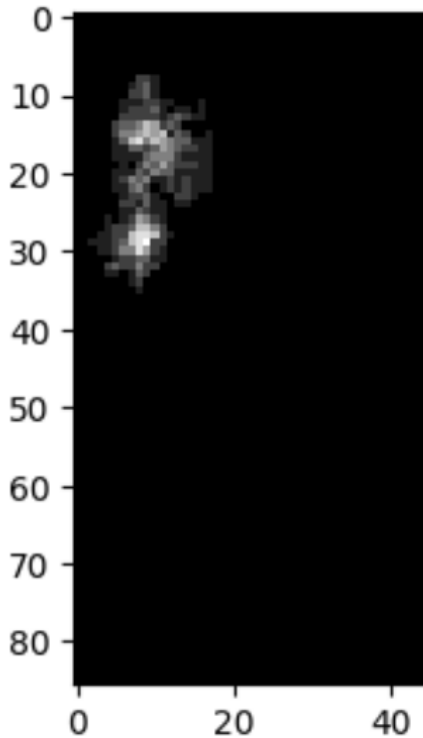
EPFL



1. Simplifications
 - Grey scale
 - Gaussian blur

2. Feature extraction
 - Pretrained
 - 4096 feat

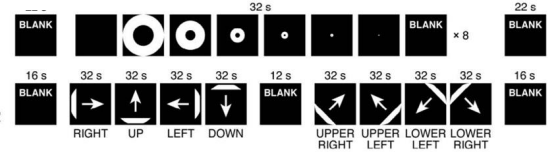
5. Splitting each clip
 - 80% in trainset
 - 20% in testset



4) RETCON

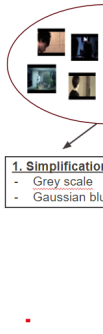
5) RETBAR1

6) RETBAR2





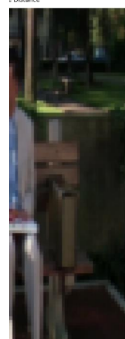
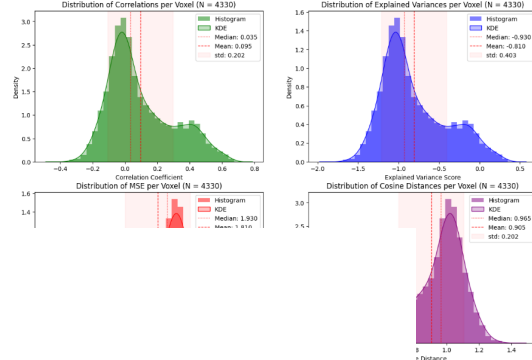
EPFL



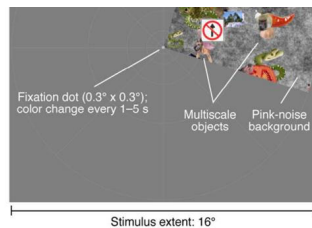
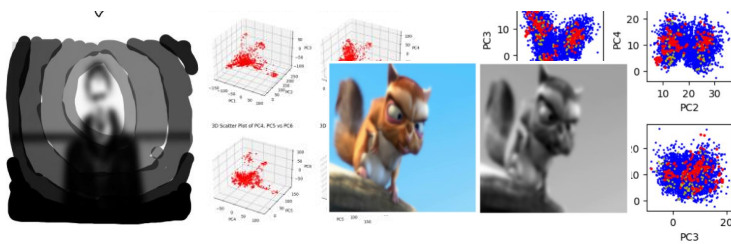
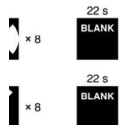
Delete Failed

send2trash failed: [Errno 28] No space left on device

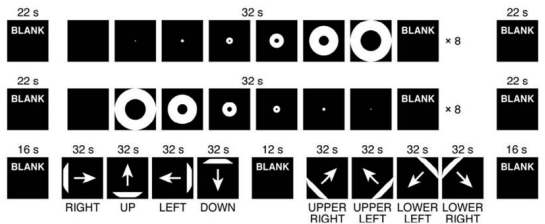
went wrong...



Dismiss



3) RETEXP



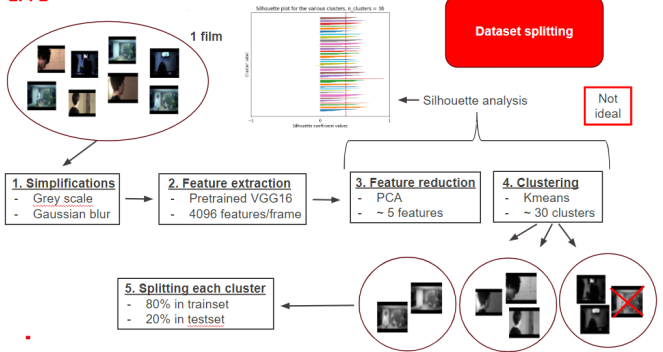
4) RETCON

5) RETBAR1

6) RETBAR2



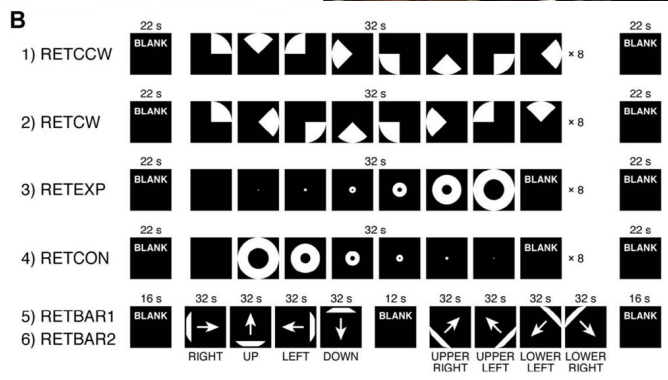
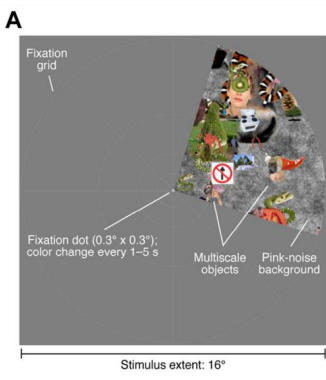
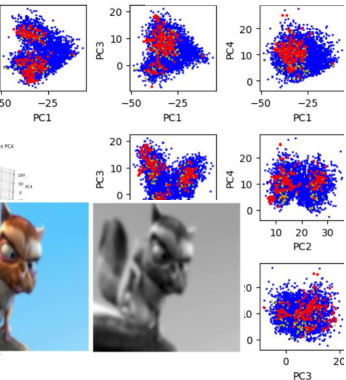
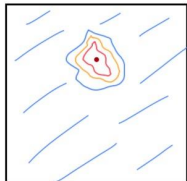
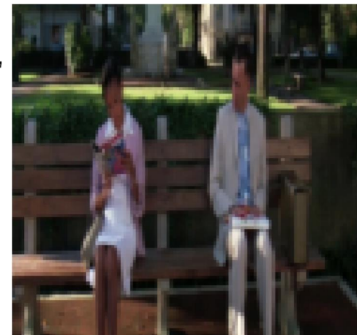
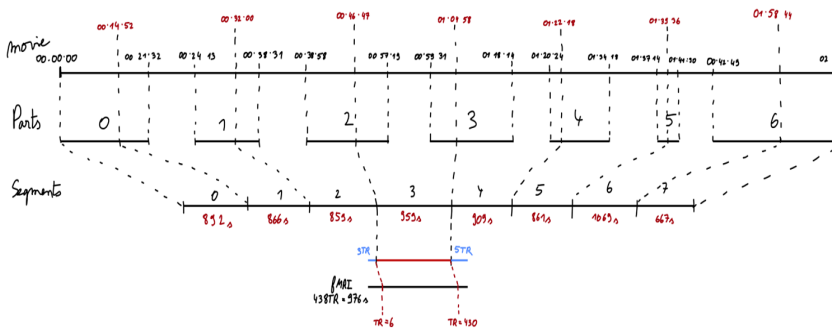
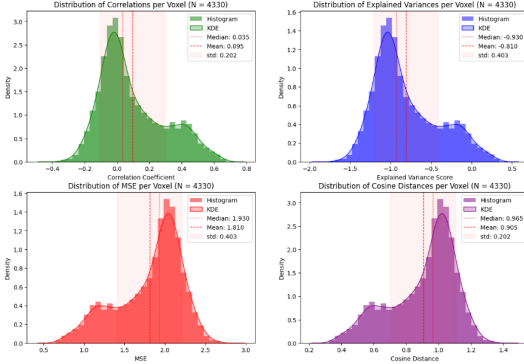
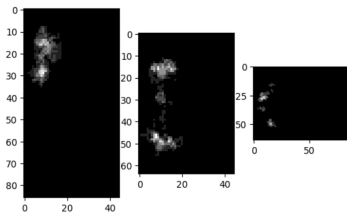
EPFL

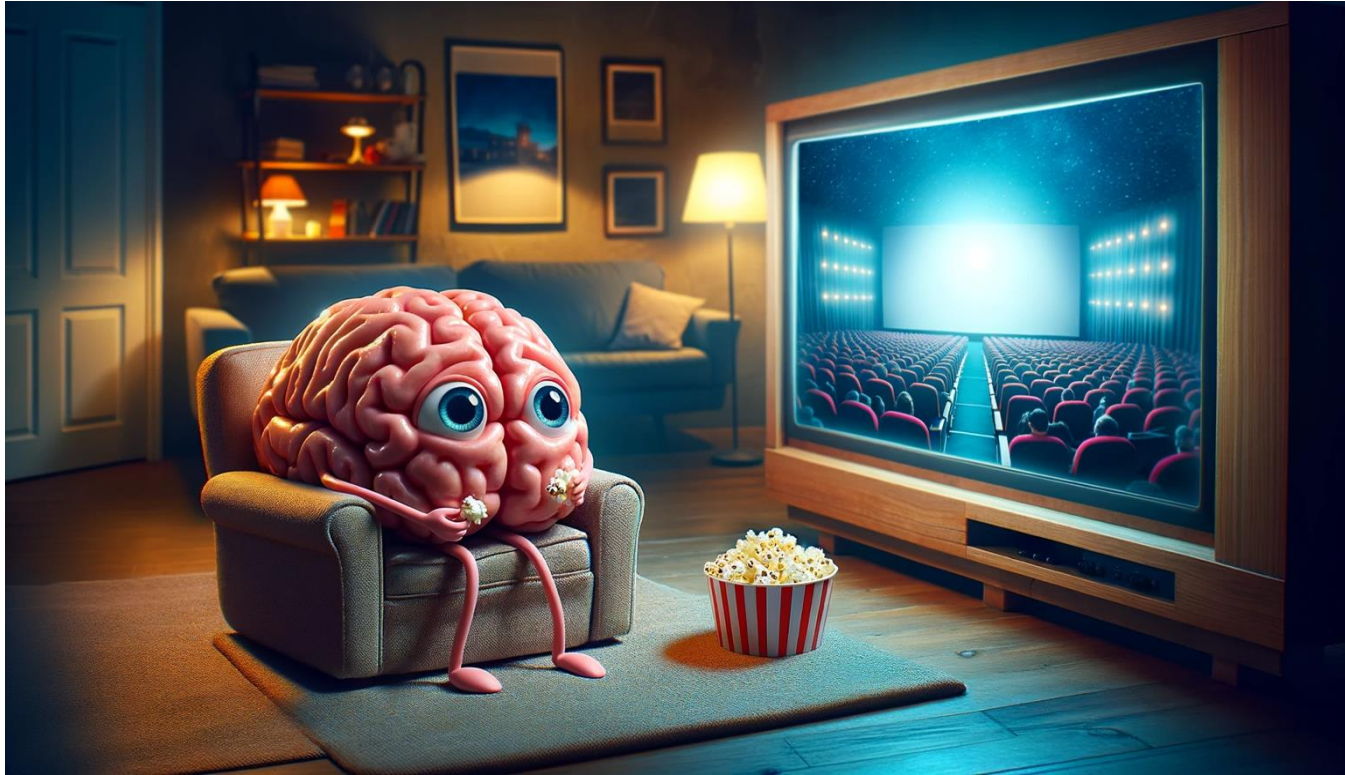


Delete Failed

send2trash f1ailed: [Errno 28] No space left on device

Dismiss





Thank you for your attention!