# Semantic information shapes gaze patterns during naturalistic viewing of movies

**WashU** 

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#### Background

- Contrastive Language-Image Pretraining (CLIP)<sup>1</sup>: aligns image and text embeddings in a shared space.
- CLIP embeddings capture semantic information that can be predictive of human gaze pattern.<sup>2</sup>
- · Scene inversion disrupts processing of semantic information in a scene, but preserves low-level visual salience inversion disrupts viewing patterns.
- However, no studies have examined the role of semantic information on gaze patterns in dynamic movies.
- · This study investigates how semantic information relates to gaze patterns during passive movie viewing by mapping CLIP-extracted features to eye movements.

#### Hypotheses

Upright images have more accurate semantic action information, making their embeddings more likely to accurately predict human gaze patterns than those derived from flipped images.

### Eye-tracking Study

#### Participants:

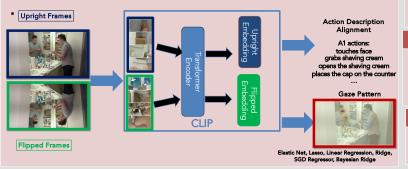
• N = 100, Female = 67, Male = 33, Mean Age: 20, SD Age: 2.39

#### Task:

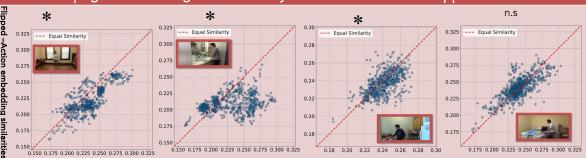
- · Participants watched four movies (exercise, grooming, cleaning, breakfast) of daily activities. The order of the movie was counterbalanced.
- · While watching movies, participants' gaze locations were recorded using EyeLink 1000.



# Mapping CLIP Embeddings to Gaze Patterns and Action Descriptions: Upright vs. Flipped Frames



## Upright Frames Align More Closely with Action Text than Flipped Frames



**Upright-Action Embedding Similarities** 

# Upright CLIP Embeddings Better Predict Passive Gaze Patterns than Flipped Embeddings

Train of the 80% frames in the beginning, test on 20% frames in the end for each movie.



We fit a linear mixed-effects model predicting Jensen-Shannon divergence from embedding type (upright vs. flip), with a random intercept for movie and an AR(1) structure to account for frame-toframe autocorrelation within each movie and embedding type.

Model shows a significant effect of embedding type, b = -0.016, SE = 0.004, t(995) = -3.49, p < .001

#### Conclusions & Future Directions

- Upright embeddings better predicts actual gaze patterns than flipped embeddings, likely because they preserve action-related semantic information that guides visual attention
- Other transformations that preserve low-level alignment information—such as diffeomorphic—could be applied to test whether our findings generalize beyond simple flip
- Comparing model variants to naturalistic gaze patterns provides a testbed for theory evaluation.

#### References

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#### **Acknowledgements**

We thank the Dynamic Cognition Lab for their support and helpful discussion.

We thank Dr. Tan Nguyen for his help with poster production stledu | Tweet at me: (