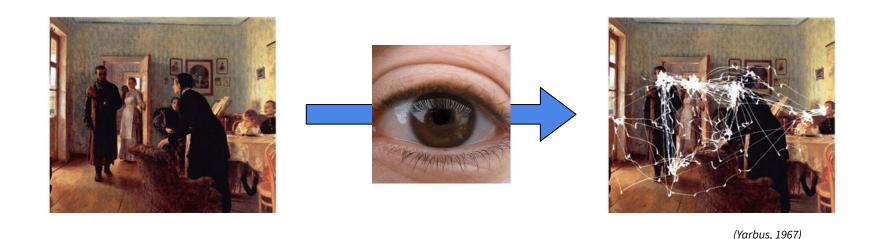
# Modeling Human Eye Movements with Neural Networks in a Maze-Solving Task

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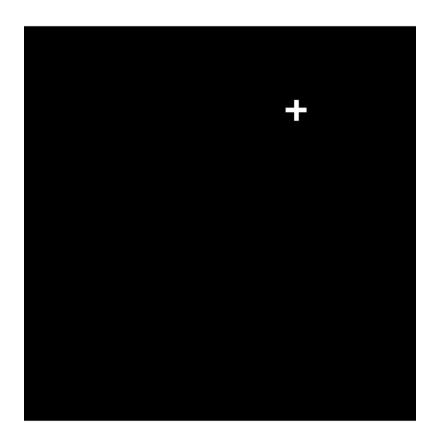
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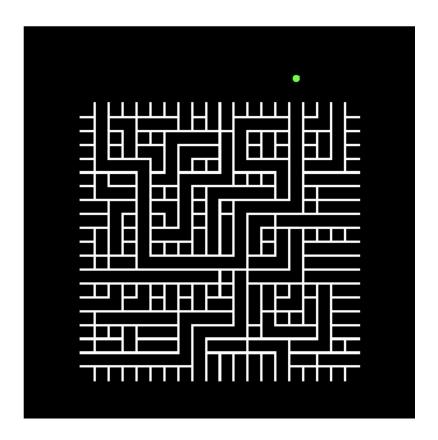


Goal: Build a **generative model of eye movements** that incorporates features of human vision like a fovea and discrete saccades.

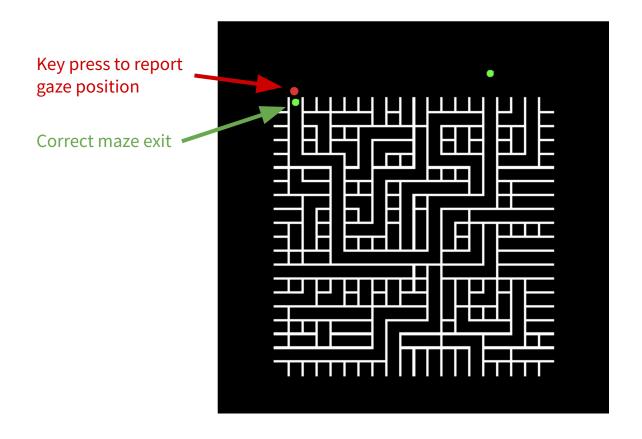
## Maze task: Fixation at Entrance



### Maze task: Maze Presentation

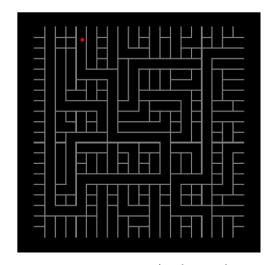


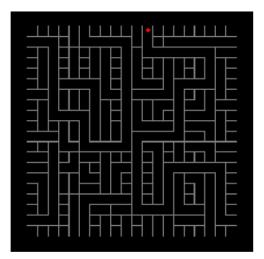
#### Maze task: Saccade to the Exit



#### Maze task

- Rich variety of spatial plans allows study of task-driven saccade sequences
- Eye movements are largely consistent across humans
- Allows us to test mental simulation as a strategy guiding eye movements





(videos shown 30% full speed)

#### Maze task

- Rich variety of spatial plans allows study of task-driven saccade sequences
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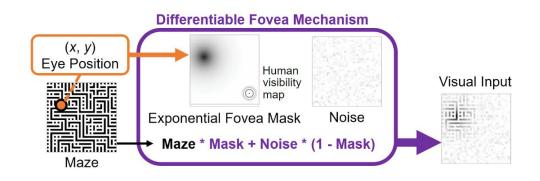
#### **Human data collection**

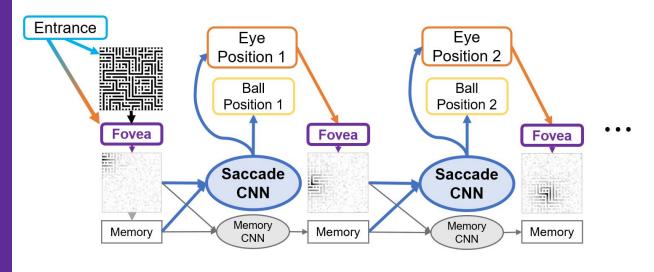
- 14 human volunteers
- Each subject completed ~400 trials
- Saccades extracted by filtering and thresholding eye velocity

#### Gaze RNN Models

 Novel differentiable fovea mechanism allows end-to-end training

 ConvRNN architecture unrolls through time for any number of saccades





$$t=1$$
  $t=2$ 

#### **3 Gaze RNN Variations**

• EXIT: Reach exit in as few saccades as possible

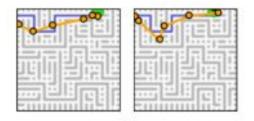
Minimize 
$$L_{\text{EXIT}} = \frac{1}{n} \sum_{i=1}^{n} (\hat{p}_i^{\text{eye}} - p^{\text{exit}})^2$$

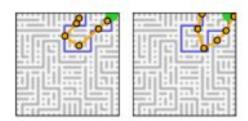
• SIM: Track an imaginary "ball" moving through the maze

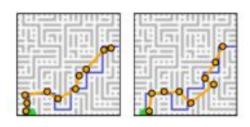
Minimize 
$$L_{ ext{SIM}} = rac{1}{n} \sum_{i=1}^n (\hat{p}_i^{ ext{ball}} - p_i^{ ext{ball}})^2$$

 $\bullet$  HYBRID: Weighted sum of the two loss terms  $\text{Minimize} \quad L_{\text{HYBRID}} = \beta \cdot L_{\text{EXIT}} + (1-\beta) \cdot L_{\text{SIM}}$ 

#### **Results:** Qualitative Behavior







**Human 1** 

**Human 2** 

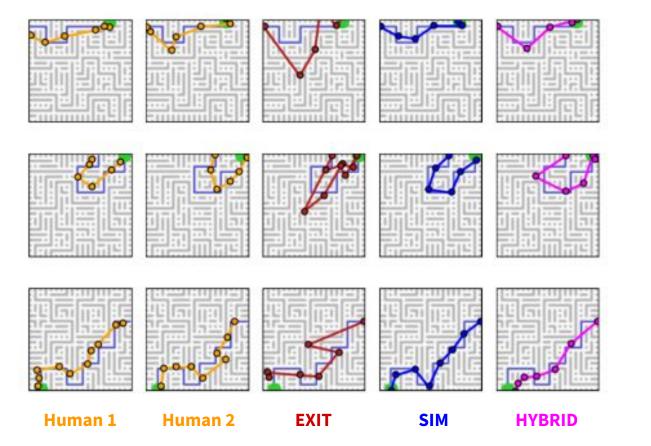
Exit

--- Path

Fixation points



#### **Results:** Qualitative Behavior



Exit

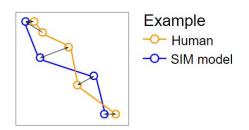
--- Path

Fixation points
Human
EXIT
SIM
HYBRID

#### **Results:** Quantitative Metrics

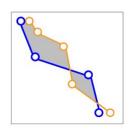
#### **Nearest Neighbors Distance**

Mean of the nearest point in path A to every point in path B and vice versa

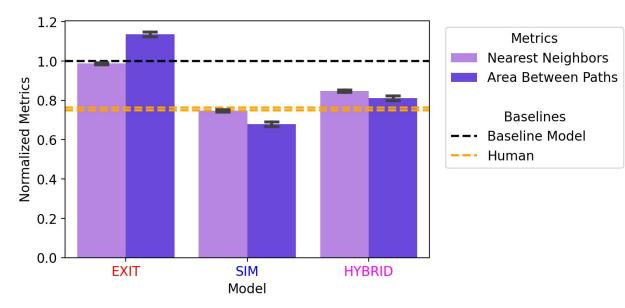


#### **Area Between Paths**

Total plane area of the polygon(s) formed between paths A and B



#### Metric scores between model and human eye paths



#### **Conclusions & Future Work**

- In the maze-solving task, a gaze RNN trained to run an internal simulation of a ball moving through the maze generates more human-like eye movements than a model trained to solve the task optimally
- Humans may employ a similar mental simulation when solving this task
- Future work may address:
  - Biological accuracy of model fovea hyperparameters
  - Introducing non-constant simulation speed
  - Allow gaze RNN to control fixation durations
  - Apply gaze RNNs to other task domains

# Acknowledgements







# Thank you!