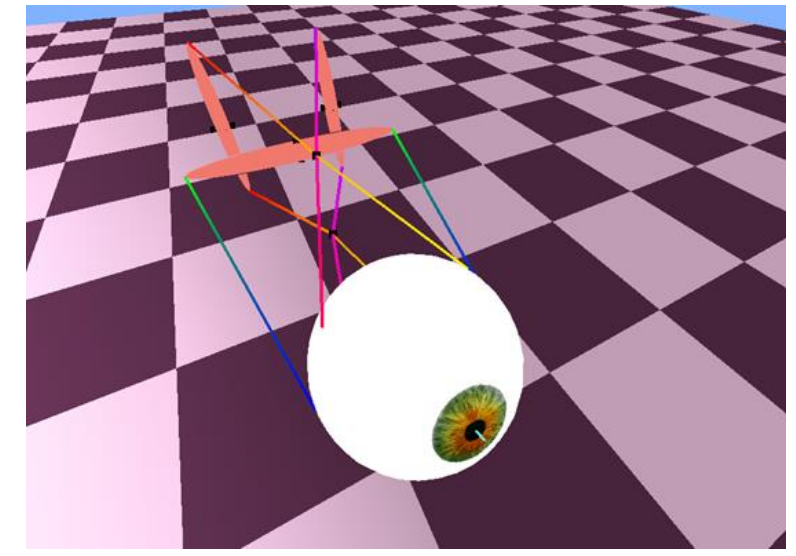


Learning open-loop saccadic control of a 3D biomimetic eye using the actor-critic algorithm



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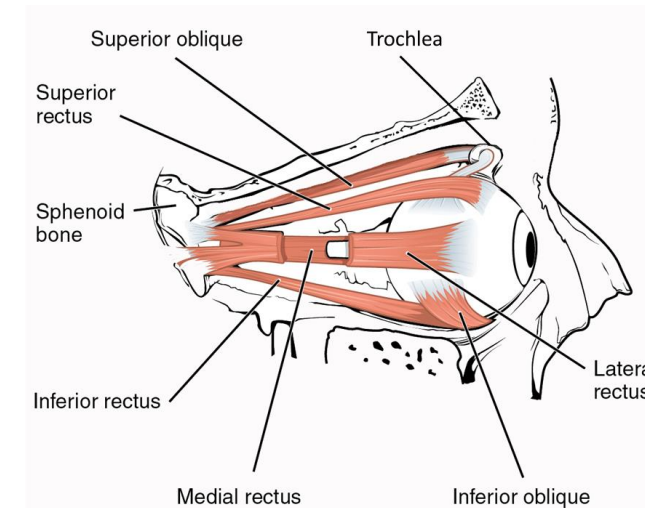
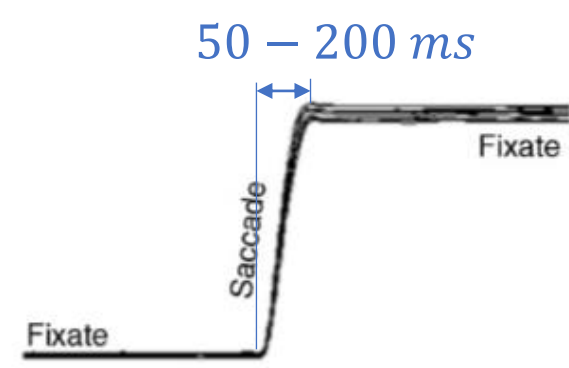
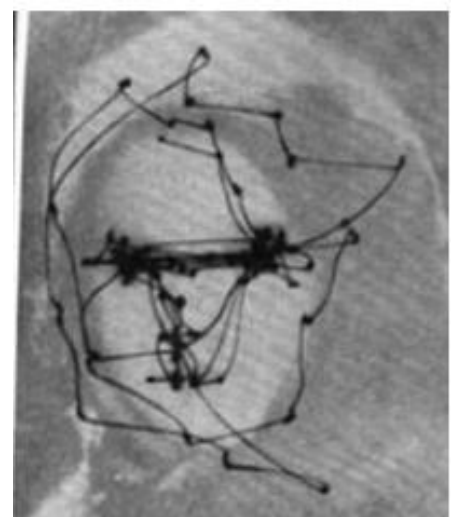
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Objectives

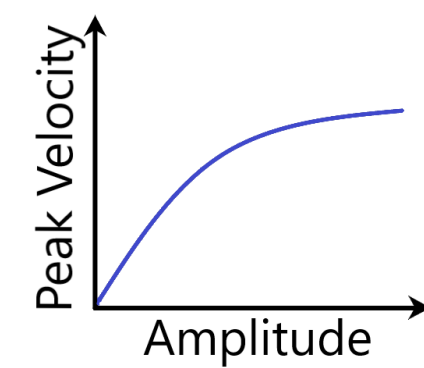
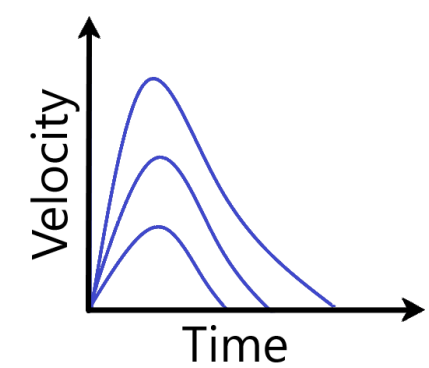
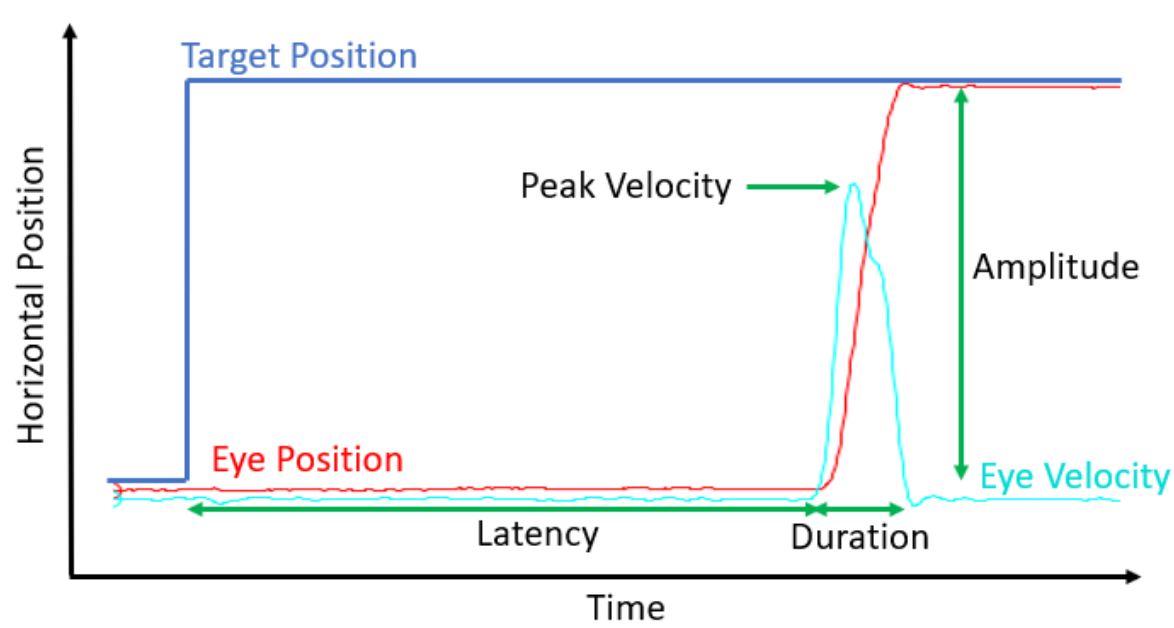
- Understand how brain controls eye saccades by using a bio-inspired robotic eye.
- We utilize reinforcement learning with biologically motivated action and rewards.
- We aim to see stereotypical kinematic characteristics of primate saccades emerge from our control.

Eye Saccades

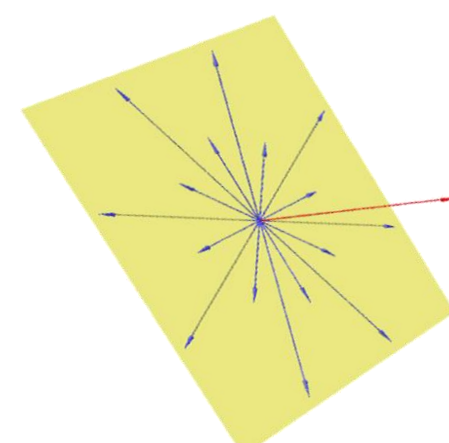
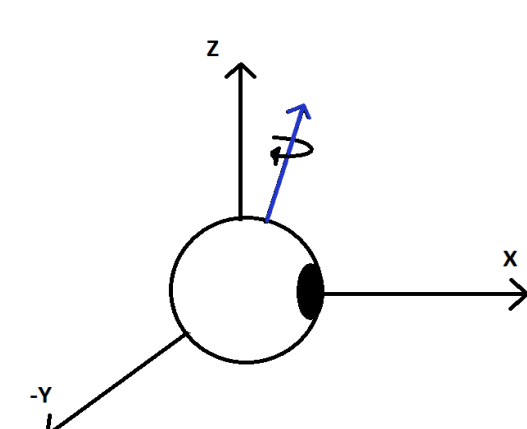
- Rapid eye movements (~700°/sec) to change eye fixation.



- Duration too short for visual feedback.
- 6 muscles control 3 DOF rotations
- High viscous friction due to surrounding fat tissues.
- Stereotypical velocity characteristics.



- Stereotypical orientation characteristic (Listing's Law)

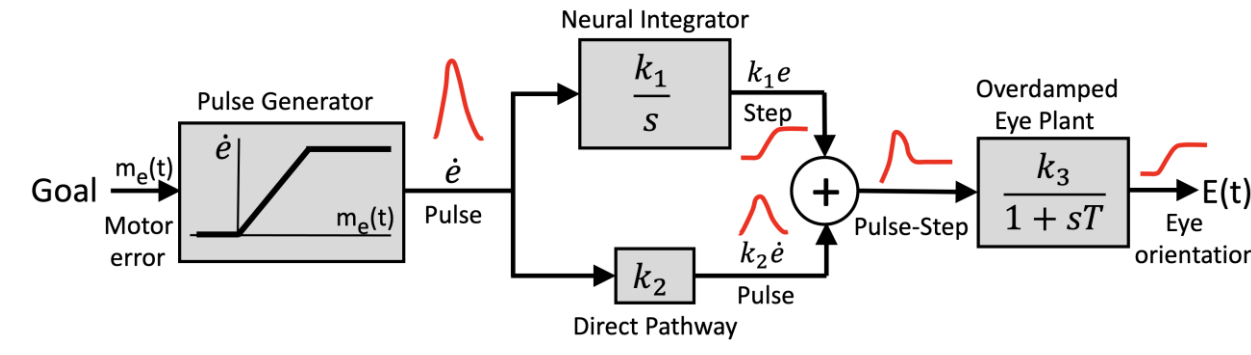


The rotation vectors for orientations from primary position all lie in a plane.

Existing Works

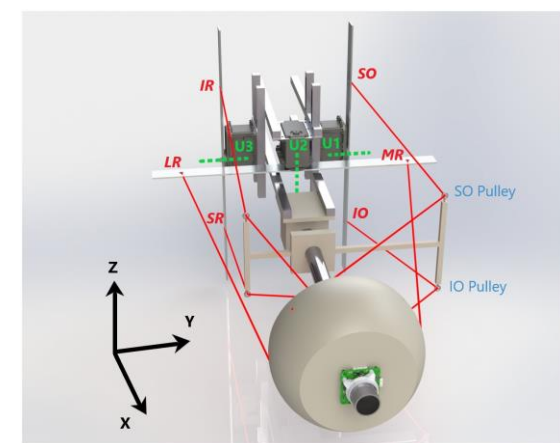
Works on understanding these characteristics:

1. (1973) Robinson 1D model control theory



2. (2010) Shadmehr et al. – Optimal control 1D model

3. (2021) John et al. – Optimal control 3D model

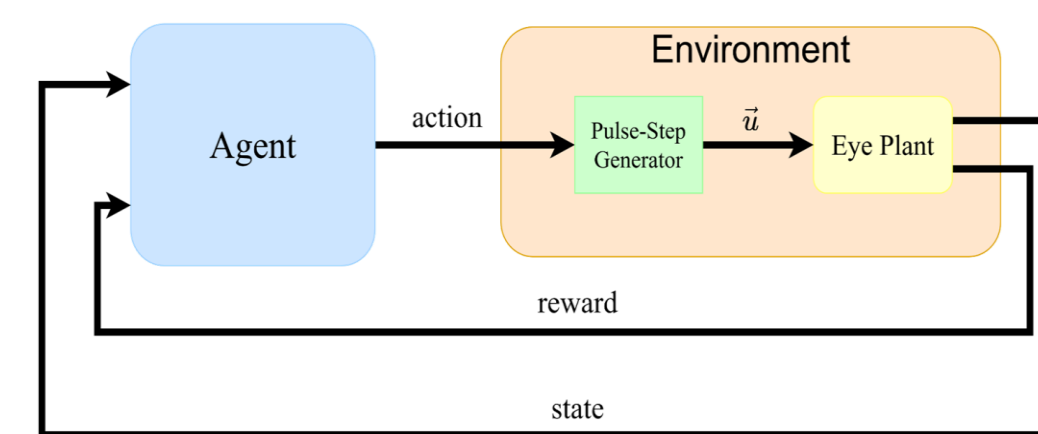


- 1.D. A. Robinson, "Models of the saccadic eye movement control system," Biologic Cybernetics, 1973.
- 2.R. Shadmehr, J. Orban, M. Xu-Wilson, and T.-Y. Shih, "Temporal discounting of reward and the cost of time in motor control," J. Neuroscience, vol. 30, pp. 10 507–16, 08 2010.
- 3.A. John, C. Aleluia, A. J. Van Opstal, and A. Bernardino, "Modelling 3d saccade generation by feedforward optimal control," PLOS Computational Biology, vol. 17, no. 5, pp. 1–35, 05 2021.

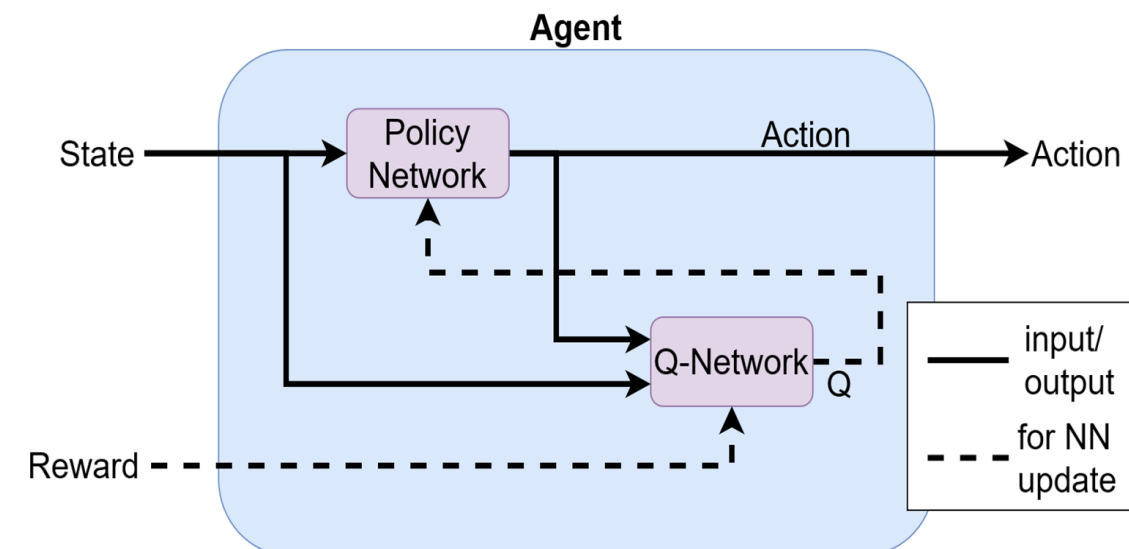
Our Approach

Agent – Actor-Critic Algorithm

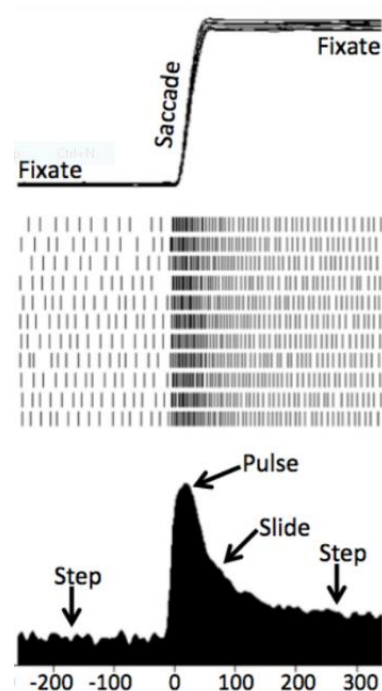
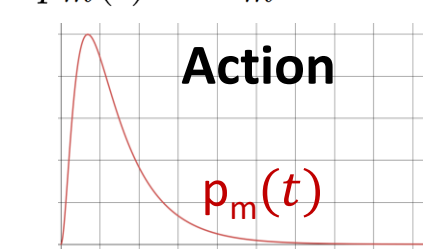
- Policy Network : Defines Action
- Q Network : Estimates reward for action.



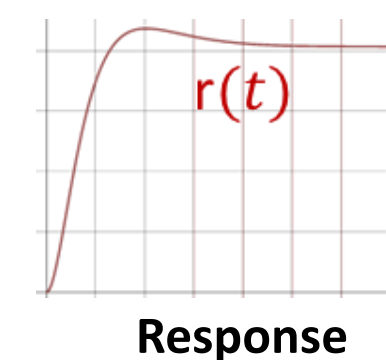
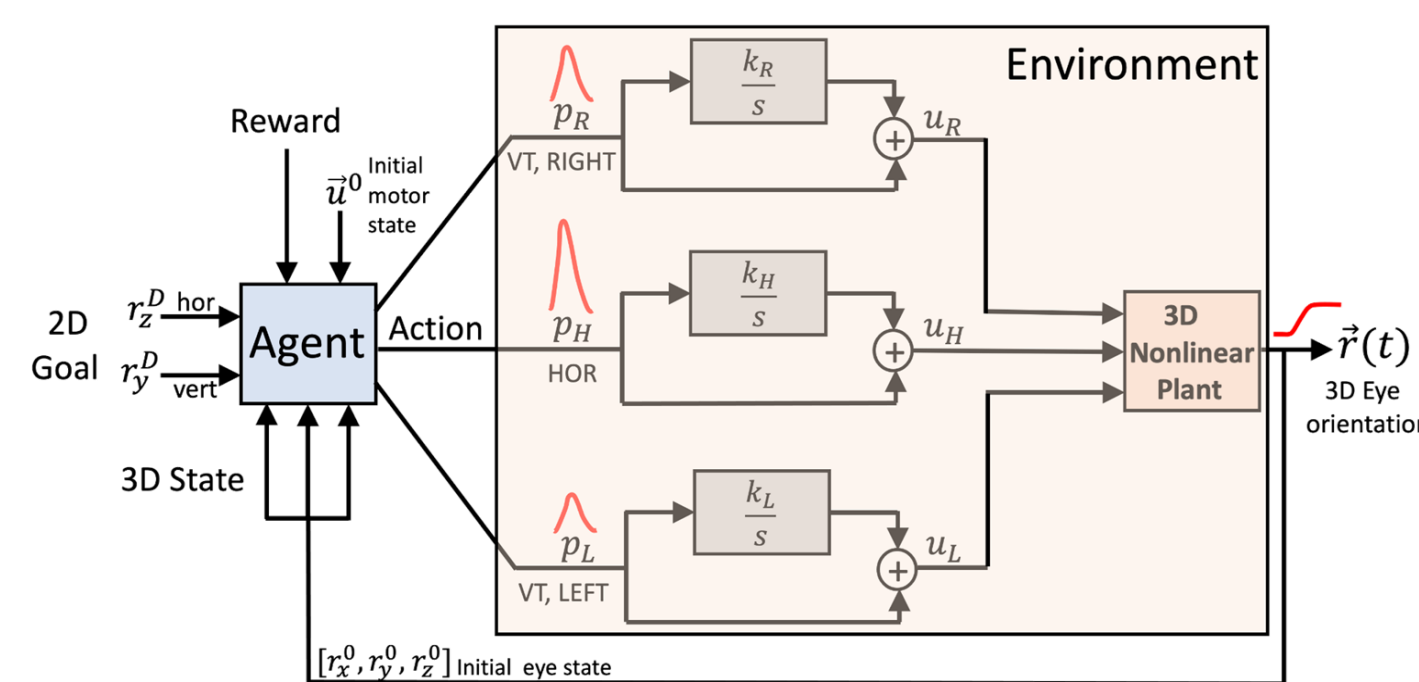
State = (orientation, motor position, 2D goal)



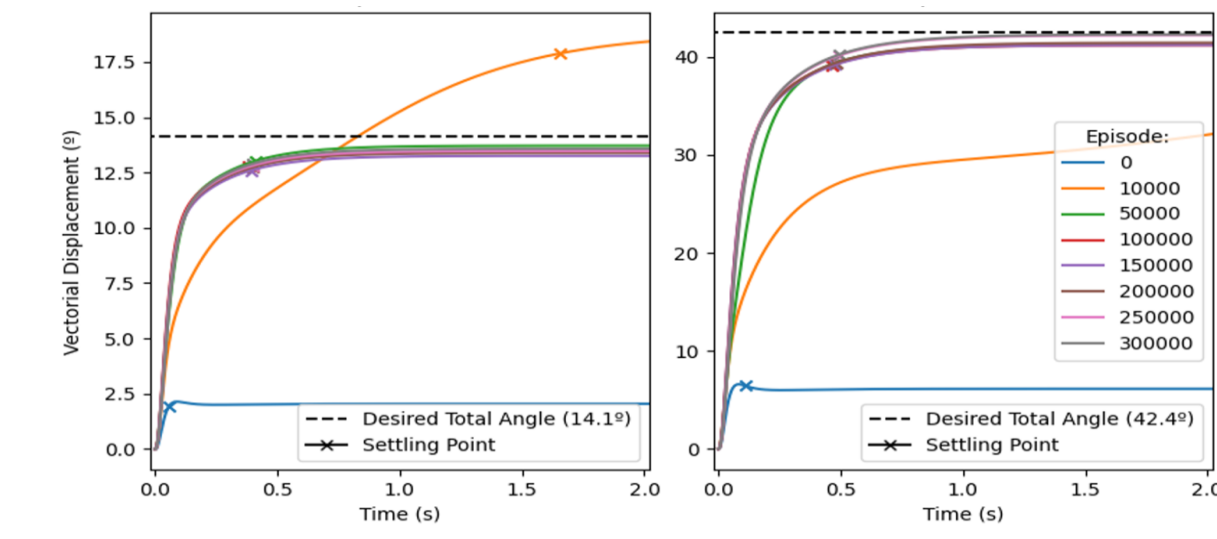
$$p_m(t) = A_m t^2 e^{-B_m t}$$



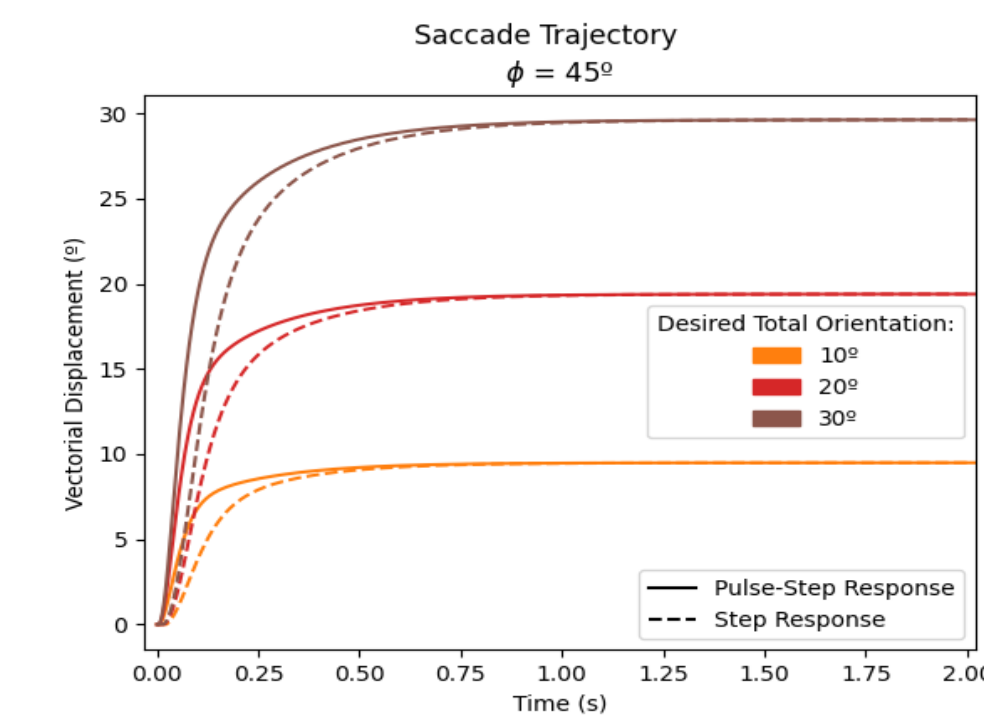
$$\text{Reward} = \lambda_a R_{acc} + \lambda_e R_{en} + \lambda_d R_{dur} + \lambda_f R_{for} + \lambda_o R_{over}$$



Results



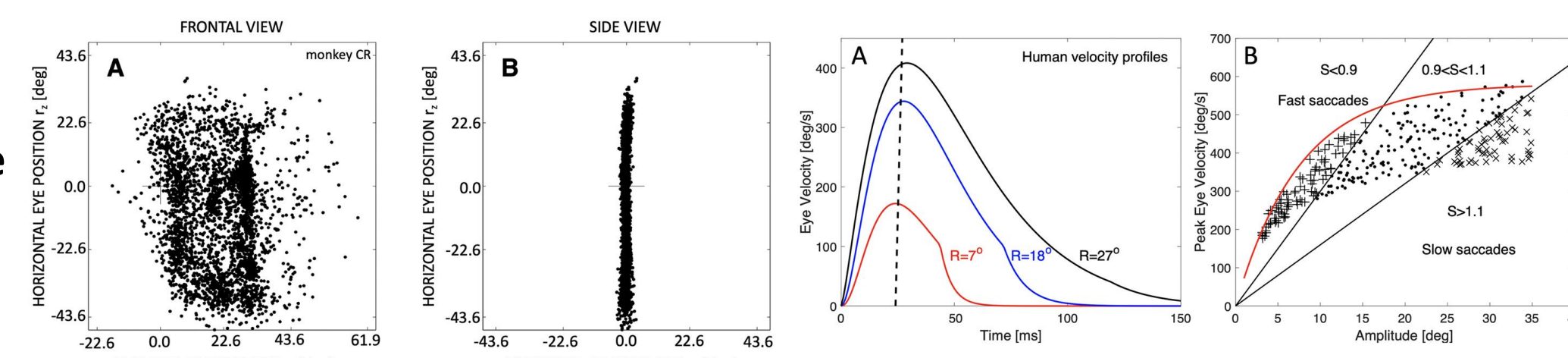
Amplitude responses for two different goals, [10°, 10°] and [30°, 30°], respectively, during different stages of the training (legend).



Showing difference between responses to the trained pulse-step signal (solid traces) vs a simple step signal (dashed), for 10°, 20° and 30° oblique saccades after training (300k episodes).

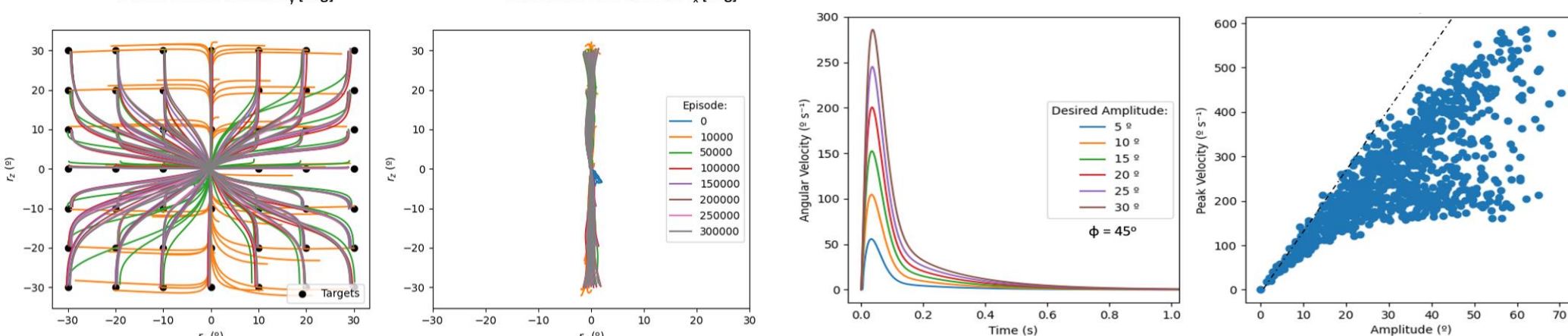
Listing's Plane

Primate Data



Velocity Characteristics

Our Results



Conclusions

- Generated 3D saccades with human-like characteristics from optimizing accuracy, duration, energy, fixation force and overshoot.
- Our results emerged without imposing any constraints.
- In future, we wish to apply the method to a more realistic robotic eye with 6 independently actuated muscles.