

Analysis and simulation of The Lady in the Lake problem

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Motivation Problem

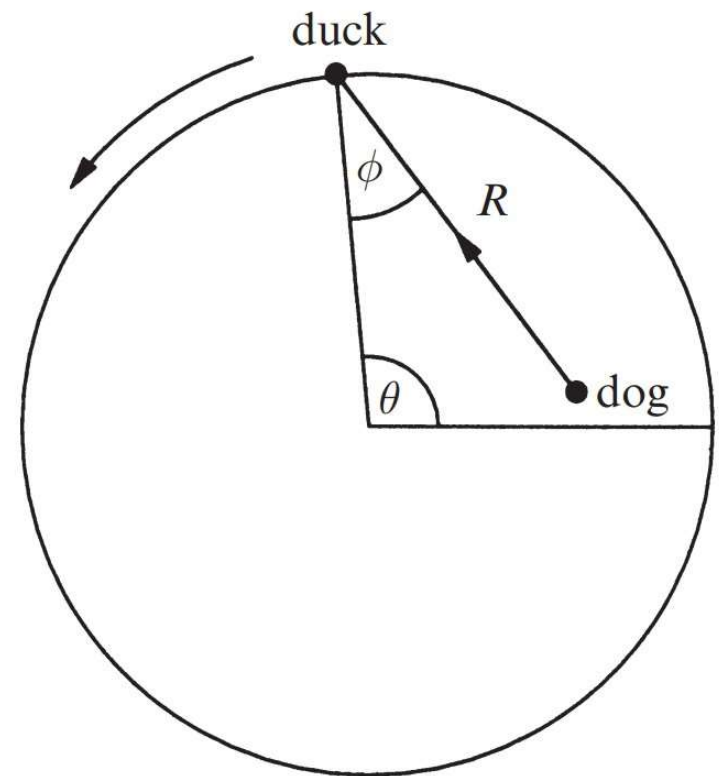
Problem 7.1.9 from (Strogatz), picture taken from the book directly.

Pond with radius $r = 1$

Duck: constant speed $v_{duck} = d\theta$ along the edge of the pond

Dog: constant speed $v_{dog} = kv_{duck}$ from center of the pond ($R = 1$ at the beginning) straight towards the duck

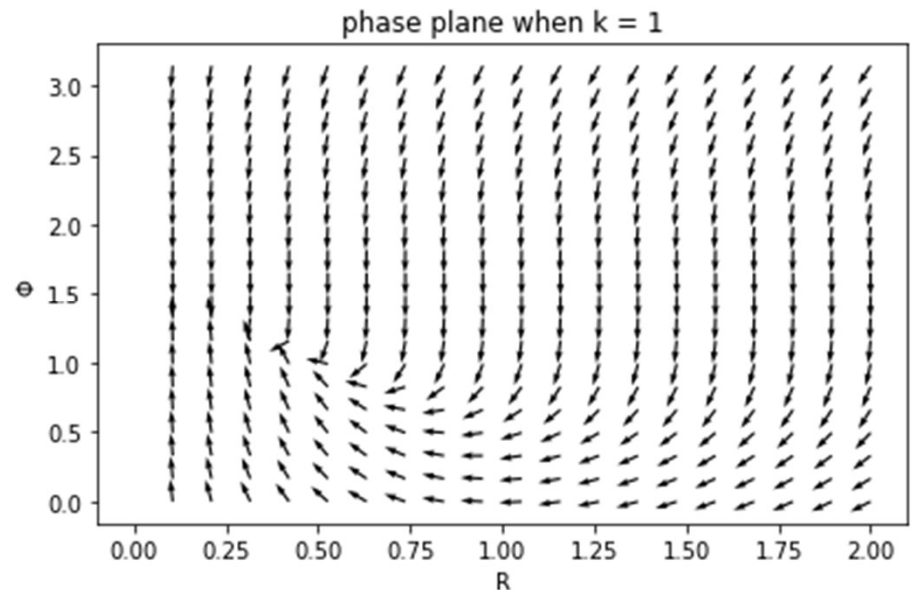
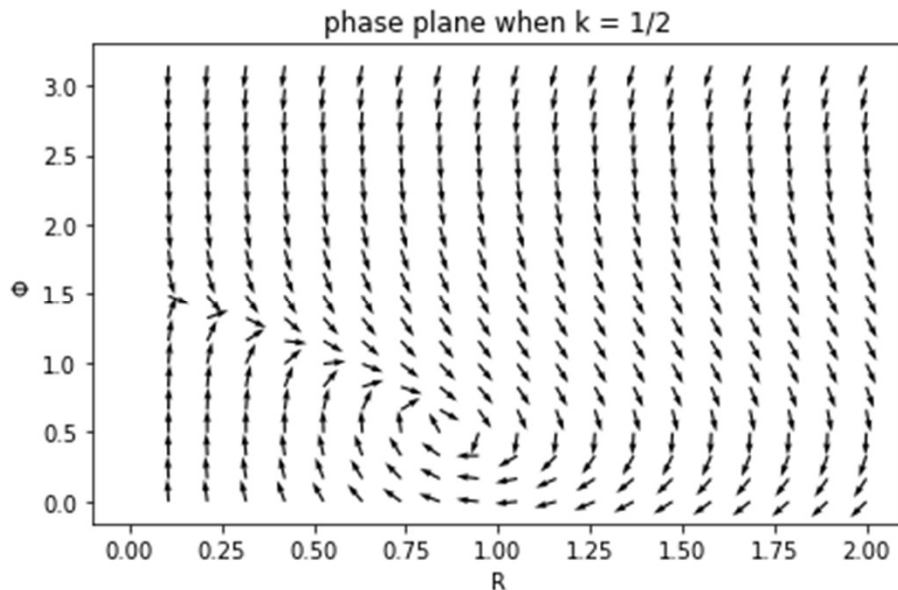
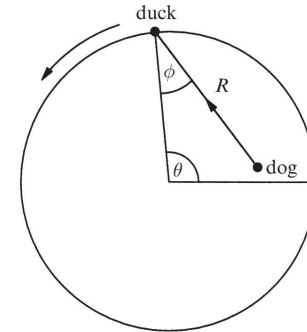
What does the dog do in the long run when $k = 1$ and $k = \frac{1}{2}$?



Motivation Problem: Solution

The system can be described by

$$\begin{cases} \frac{dR}{d\theta} = \sin \phi - k \\ \frac{d\phi}{d\theta} = \frac{\cos \phi}{R} - 1 \end{cases} \text{ with } R \in [0,2], \phi \in \left[0, \frac{\pi}{2}\right]$$



When $k = \frac{1}{2}$, and starts from $R = 1$, attracted to $\phi = \frac{\pi}{6}$, $R = \frac{\sqrt{3}}{2}$

When $k = 1$, and starts from $R = 1$, no clear attraction, but still cannot reach $R = 0$

Motivation

The analysis of phase plane can give us some information about the long term behavior. However, it may not be able to solve the following questions easily.

- For what k s can the dog reach the duck as close as possible? (R cannot be zero in the analysis, so there is no definite number to define “closest”.)
- What will happen if the speed of dog and duck are not constant?
- What will happen if the dog doesn't swim straight towards the duck?

Basic Definitions: Dynamical System

The motivation problem is a simple pursuit-evasion game, formulated as a nonlinear continuous-time autonomous dynamic system which can be described by:

$$\dot{\mathbf{x}} = f(\mathbf{x}, a), \mathbf{x}(0) = \mathbf{x}_0,$$

where $\mathbf{x} \in F^n$, F a field, \mathbf{x}_0 the initial condition and a a parameter.

It studies how trajectories and limiting behavior $\lim_{t \rightarrow \infty} \mathbf{x}(t)$ changes w.r.t. \mathbf{x}_0 and a .

Basic Definitions: Pursuit-Evasion Game

The Lady in the Lake problem is a pursuit-evasion game, formulated as a **two-person deterministic zero-sum differential game** with variable terminal time. The differential system is

$$\dot{\mathbf{x}} = f(t, \mathbf{x}, u^1(t), u^2(t)), \mathbf{x}(0) = \mathbf{x}_0.$$

This is nonautonomous. Parameters become player strategies dependent on time.

Terminal time: $T = \inf\{t \in \mathbb{R}^+, \mathbf{x}(t) \in \Lambda\}$, where Λ is the target set with boundary $\partial\Lambda$ defined by $l(t, x) = 0$

Objective function: $J(u^1, u^2) = \int_0^T g(t, \mathbf{x}, u^1(t), u^2(t))dt + q(T, x(T))$

The goal is to optimize the objective function.

Basic Definitions: Issacs Equation

Consider the minimax value of the objective function:

$$V(t, \mathbf{x}) = \min_{u^1} \max_{u^2} \left\{ \int_0^T g(t, \mathbf{x}, u^1(t), u^2(t)) dt + q(T, \mathbf{x}(T)) \right\}$$

If V is C^1 in t and \mathbf{x} , then it satisfies the Issacs equation:

$$-\frac{\partial V}{\partial t} = \min_{u^1} \max_{u^2} \left\{ \frac{\partial V}{\partial \mathbf{x}} f(t, \mathbf{x}, u^1(t), u^2(t)) + g(t, \mathbf{x}, u^1(t), u^2(t)) \right\}$$

Let $H = \frac{\partial V}{\partial \mathbf{x}} f(t, \mathbf{x}, u^1(t), u^2(t)) + g(t, \mathbf{x}, u^1(t), u^2(t))$, the following gives the solution to the costate function $p = \frac{\partial V}{\partial \mathbf{x}}$.

$$\frac{d}{dt} \left(\frac{\partial V}{\partial \mathbf{x}} \right) = - \frac{\partial H}{\partial \mathbf{x}}, \frac{\partial V}{\partial \mathbf{x}}(T) = \frac{\partial}{\partial \mathbf{x}} \left(q(T, \mathbf{x}^*(T)) \right) \text{ (solve backwards in time)}$$

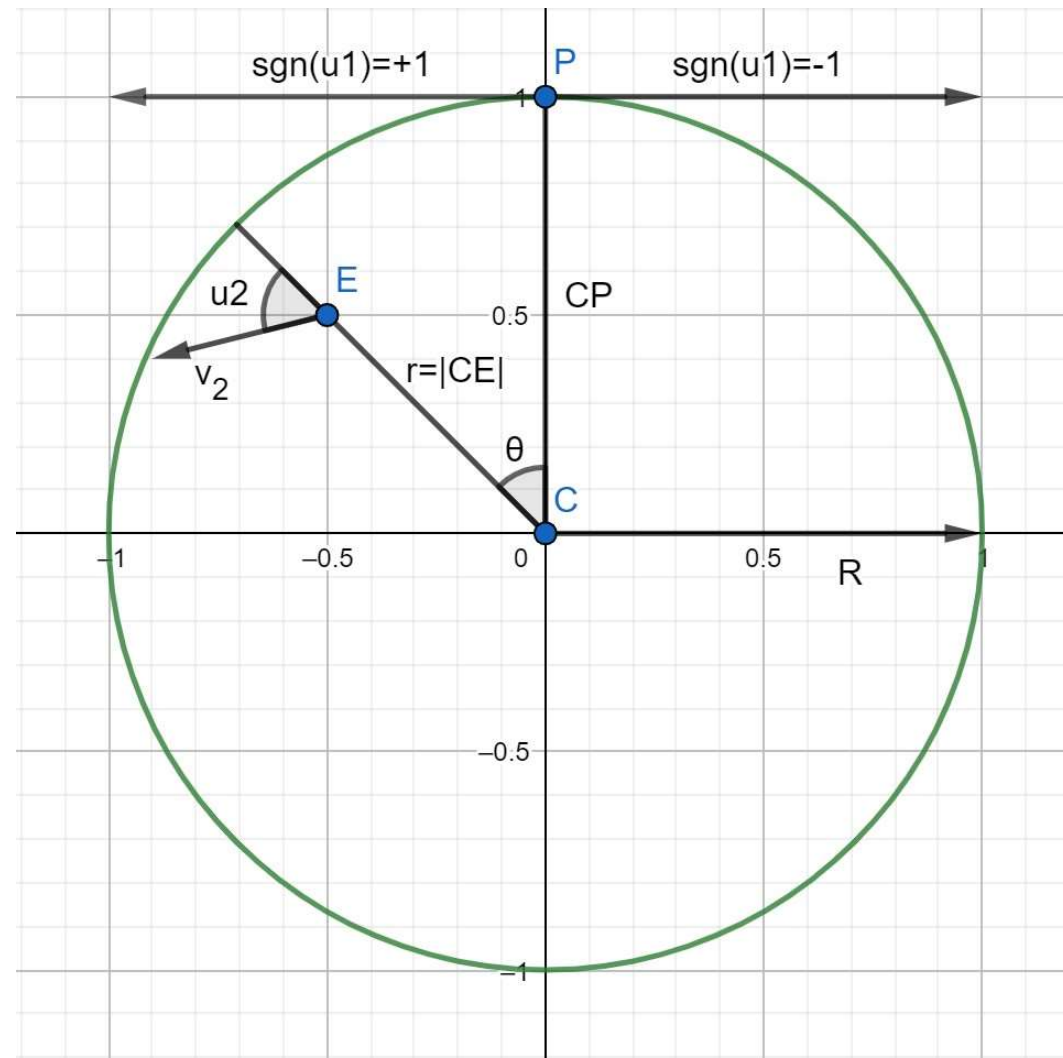
The Lady in the Lake problem

Circular pond with radius R

Pursuer P moves along the perimeter, with strategy $|u^1| \leq 1$, trying to catch the lady (evader) as she gets to the boundary.

Evader E starts from center C , with constant speed v_2 , strategy u^2 representing direction she goes w.r.t. \overrightarrow{CE} , trying to reach the edge of the pond without being caught.

Let $\theta = \angle PCE$, $r = |CE|$



The Lady in the Lake problem: Analysis

The differential equations for the system is:

$$\dot{\mathbf{x}} = \begin{pmatrix} \dot{\theta} \\ \dot{r} \end{pmatrix} = \begin{cases} \frac{v_2 \sin u^2}{r} - \frac{u^1}{R} \\ v_2 \cos u^2 \end{cases}$$

Objective function $J = |\theta(T)|$ with $T = \inf\{t : r(t) = R\}$, $|\theta(t)| \leq \pi$, $\forall t \in [0, T]$

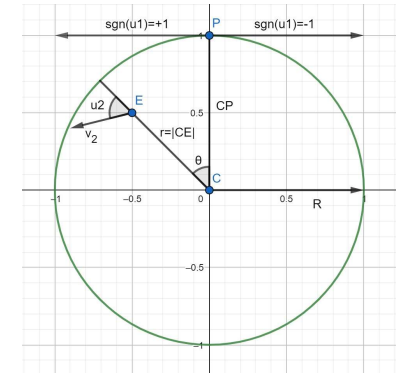
We can see that $g \equiv 0$, $q(T, \mathbf{x}(T)) = |\theta(T)|$.

So the Issacs equation is:

$$0 = -\frac{\partial V}{\partial t} = \min_{u^1} \max_{u^2} \left\{ \frac{\partial V}{\partial \mathbf{x}} f(t, \mathbf{x}, u^1(t), u^2(t)) \right\} = \min_{u^1} \max_{u^2} \left\{ \frac{\partial V}{\partial \theta} \left(\frac{v_2 \sin u^2}{r} - \frac{u^1}{R} \right) + \frac{\partial V}{\partial r} (v_2 \cos u^2) \right\}$$

It is separable in u^1 and u^2 .

$$0 = \min_{u^1} \left\{ -\frac{u^1}{R} \frac{\partial V}{\partial \theta} \right\} + v_2 \max_{u^2} \left\{ \frac{1}{r} \frac{\partial V}{\partial \theta} \sin u^2 + \frac{\partial V}{\partial r} \cos u^2 \right\}$$



The Lady in the Lake problem: Analysis

$$\text{Let } H = \frac{\partial V}{\partial \theta} \left(\frac{v_2 \sin u^2}{r} - \frac{u^1}{R} \right) + \frac{\partial V}{\partial r} (v_2 \cos u^2)$$

The costate function is given by

$$\begin{aligned} \frac{d}{dt} \left(\frac{\partial V}{\partial \theta} \right) &= -\frac{\partial H}{\partial \theta} = 0, \frac{\partial V}{\partial \theta}(T) = \frac{\partial}{\partial \theta} |\theta(T)| = \text{sgn}(\theta(T)) \\ \frac{d}{dt} \left(\frac{\partial V}{\partial r} \right) &= -\frac{\partial H}{\partial r} = \frac{\partial V}{\partial \theta} \left(\frac{v_2 \sin u^2}{r^2} \right), \frac{\partial V}{\partial r}(T) = 0 \end{aligned}$$

The θ equation gives $\frac{\partial V}{\partial \theta}(t) = \text{sgn}(\theta(T)), \forall t \in [0, T]$.

This will help us find the optimal strategies.

The Lady in the Lake problem: Analysis

$$0 = \min_{u^1} \left\{ -\frac{u^1}{R} \frac{\partial V}{\partial \theta} \right\} + v_2 \max_{u^2} \left\{ \frac{1}{r} \frac{\partial V}{\partial \theta} \sin u^2 + \frac{\partial V}{\partial r} \cos u^2 \right\} (*)$$

$$\min_{u^1} \left\{ -\frac{u^1}{R} \frac{\partial V}{\partial \theta} \right\} = -\frac{1}{R} \left| \frac{\partial V}{\partial \theta} \right| = -\frac{1}{R} \text{ is achieved when } u^{1*} = \text{sgn} \left(\frac{\partial V}{\partial \theta} \right) = \text{sgn} \theta(T).$$

$\max_{u^2} \left\{ \frac{1}{r} \frac{\partial V}{\partial \theta} \sin u^2 + \frac{\partial V}{\partial r} \cos u^2 \right\} = \max_{u^2} \left\{ \left(\frac{1}{r} \frac{\partial V}{\partial \theta}, \frac{\partial V}{\partial r} \right) \cdot (\sin u^2, \cos u^2) \right\}$ is achieved when the two vectors $\left(\frac{1}{r} \frac{\partial V}{\partial \theta}, \frac{\partial V}{\partial r} \right)$ and $(\sin u^2, \cos u^2)$ are parallel.

$$\text{Let } \frac{1}{r} \frac{\partial V}{\partial \theta} = k \sin u^{2*}, \frac{\partial V}{\partial r} = k \cos u^2, \text{ then } \max_{u^2} \left\{ \left(\frac{1}{r} \frac{\partial V}{\partial \theta}, \frac{\partial V}{\partial r} \right) \cdot (\sin u^2, \cos u^2) \right\} = k$$

Substitute u^{1*} and k back into (*), $k = \frac{1}{Rv_2}$.

$$\text{Then } \sin u^{2*} = \frac{Rv_2}{r} \frac{\partial V}{\partial \theta} = \frac{Rv_2}{r} \text{sgn} \theta(T)$$

The Lady in the Lake problem: Analysis

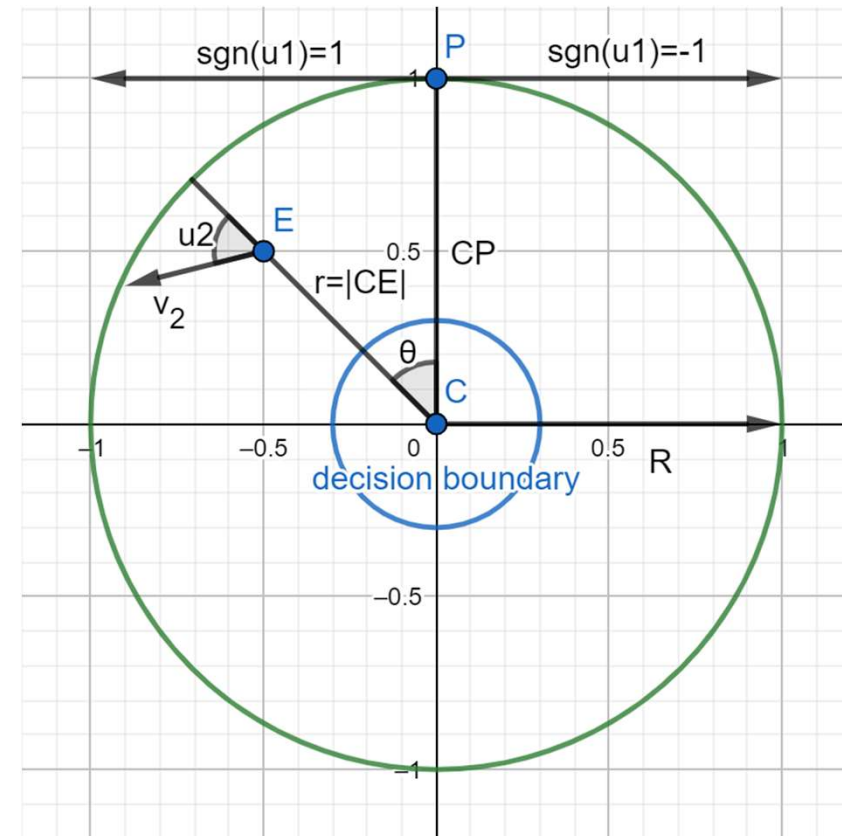
$$u^{1*} = \text{sgn}(\theta(T)), \sin u^{2*} = \frac{Rv_2}{r} \text{sgn} \theta(T)$$

What do these strategies mean?

$u^{1*} = \text{sgn}(\theta(T))$ means **P** should try to move in the direction to minimize θ

$\sin u^{2*} = \frac{Rv_2}{r} \text{sgn} \theta(T)$ is only valid when $r \geq Rv_2$
E should try to maximize θ when **E** is outside of the decision boundary $\{\mathbf{x} : \|\mathbf{x}\| = Rv_2\}$.

When $r < Rv_2$, **E**'s maximum angular velocity $\omega_E = \frac{v_2}{r} \geq \frac{1}{R} \geq \omega_P$, so **E** can always make θ as large as possible before reaching the decision boundary.



The Lady in the Lake problem: Analysis

Under what conditions can **E** escape?

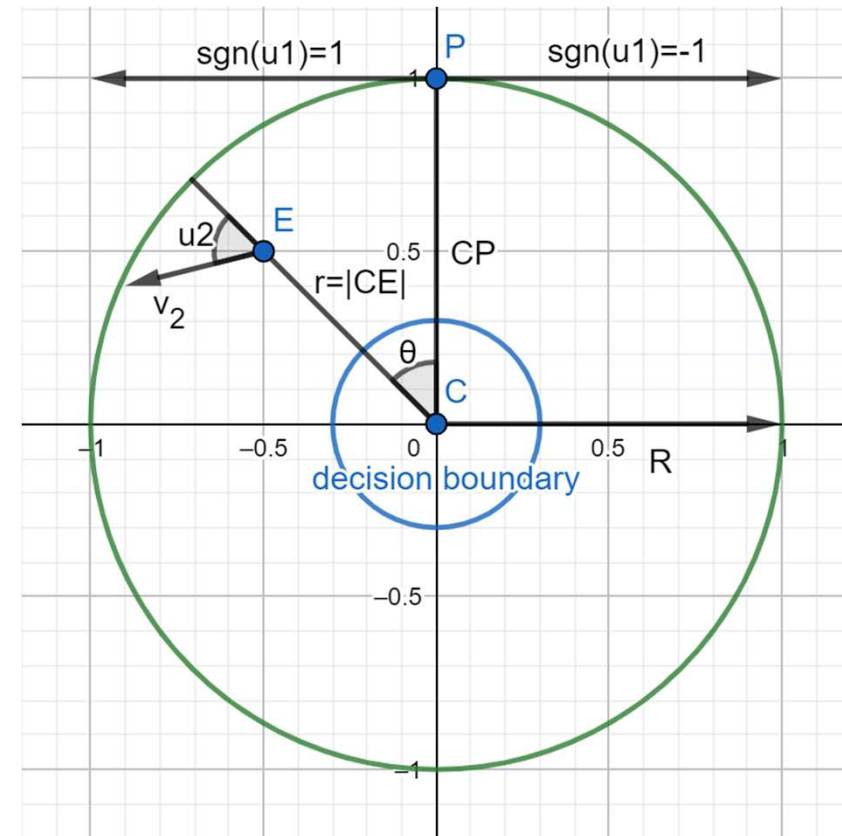
Ans: **E** can escape when $|\theta(T)| > 0$

T. Basar calculated $|\theta(T)| = \pi + \arccos v_2 - \frac{1}{v_2} \sqrt{1 - v_2^2}$.

$|\theta(T)| > 0$ if $v_2 > 0.21723$.

So, **E** should swim with speed $v_E \geq 0.217v_P$ to escape.

And from the previous slide, **E** should try to maximize θ when reaching the decision boundary, and keep $\sin u^{2*} = \frac{Rv_2}{r} \operatorname{sgn} \theta(T)$ at each r .



Inverse problem

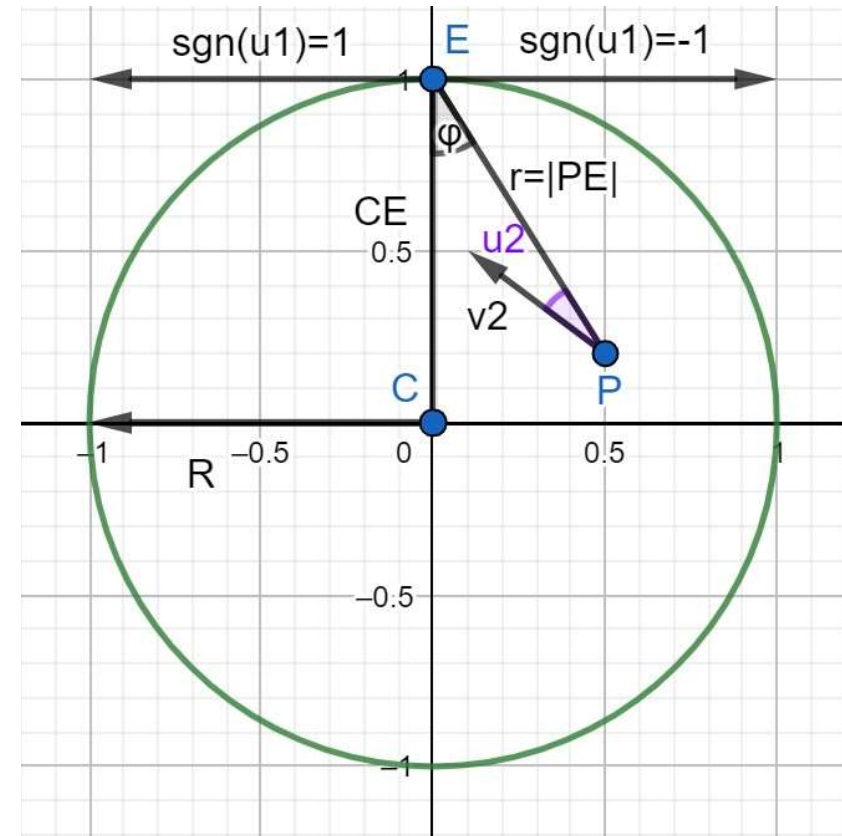
Motivation problem + strategies

Circular pond with radius R

Evader **E** moves along the perimeter, with strategy $|u^1| \leq 1$ trying to survive as long as possible without being caught by **P**.

Pursuer **P** starts from center C , with constant speed v_2 , strategy u^2 representing direction she goes w.r.t. \overrightarrow{PE} , trying to catch **E** as fast as possible.

Let $\phi = \angle CEP$, $r = |PE|$



Inverse problem: Analysis

The differential equations for the system is:

$$\dot{\mathbf{x}} = \begin{pmatrix} \dot{\phi} \\ \dot{r} \end{pmatrix} = \begin{cases} \left(\frac{R \cos \phi}{r} - 1 \right) u^1 - \frac{v_2 \sin u^2}{r} \\ u^1 \sin \phi - v_2 \cos u^2 \end{cases}$$

Objective function $J = r(T)$ with $T = \min\{\inf\{t : r(t) = 0\}, 15 \text{ sec}\}$

The Issacs equation is:

$$0 = \min_{u^2} \max_{u^1} \left\{ \frac{\partial V}{\partial \phi} \left(\left(\frac{R \cos \phi}{r} - 1 \right) u^1 - \frac{v_2 \sin u^2}{r} \right) + \frac{\partial V}{\partial r} (u^1 \sin \phi - v_2 \cos u^2) \right\}$$

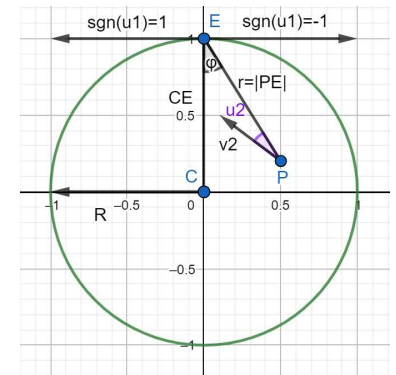
It is separable in u^1 and u^2 .

$$0 = \max_{u^1} \left\{ u^1 \left(\frac{\partial V}{\partial \phi} \left(\frac{R \cos \phi}{r} - 1 \right) + \frac{\partial V}{\partial r} \sin \phi \right) \right\} + v_2 \min_{u^2} \left\{ -\frac{1}{r} \frac{\partial V}{\partial \phi} \sin u^2 - \frac{\partial V}{\partial r} \cos u^2 \right\}$$

The optimal strategies are:

$$u^{1*} = \text{sgn} \left(\frac{\partial V}{\partial \phi} \left(\frac{R \cos \phi}{r} - 1 \right) + \frac{\partial V}{\partial r} \sin \phi \right), (\sin u^{2*}, \cos u^{2*}) \parallel \left(-\frac{1}{r} \frac{\partial V}{\partial \phi}, -\frac{\partial V}{\partial r} \right), \text{ with}$$

reverse direction



Inverse problem: Analysis

$$\text{Let } H = \frac{\partial V}{\partial \phi} \left(\left(\frac{R \cos}{r} - 1 \right) u^1 - \frac{v_2 \sin u^2}{r} \right) + \frac{\partial V}{\partial r} (u^1 \sin \phi - v_2 \cos u^2)$$

The costate function is given by

$$\begin{aligned} \frac{d}{dt} \left(\frac{\partial V}{\partial \phi} \right) &= u^1 \frac{\partial V}{\partial \phi} \frac{R \sin \phi}{r} - u^1 \frac{\partial V}{\partial r} \cos \phi, \frac{\partial V}{\partial \phi} (T) = \frac{\partial}{\partial \phi} r(T) = 0 \\ \frac{d}{dt} \left(\frac{\partial V}{\partial r} \right) &= \frac{\partial V}{\partial \phi} \left(\frac{R \cos \phi}{r^2} u^1 - \frac{v_2 \sin u^2}{r^2} \right), \frac{\partial V}{\partial r} (T) = 1 \end{aligned}$$

This time, it doesn't really give any helpful information for solving the system.

Inverse Problem: Analysis

$$u^{1*} = \text{sgn} \left(\frac{\partial V}{\partial \phi} \left(\frac{R \cos}{r} - 1 \right) + \frac{\partial V}{\partial r} \sin \phi \right)$$

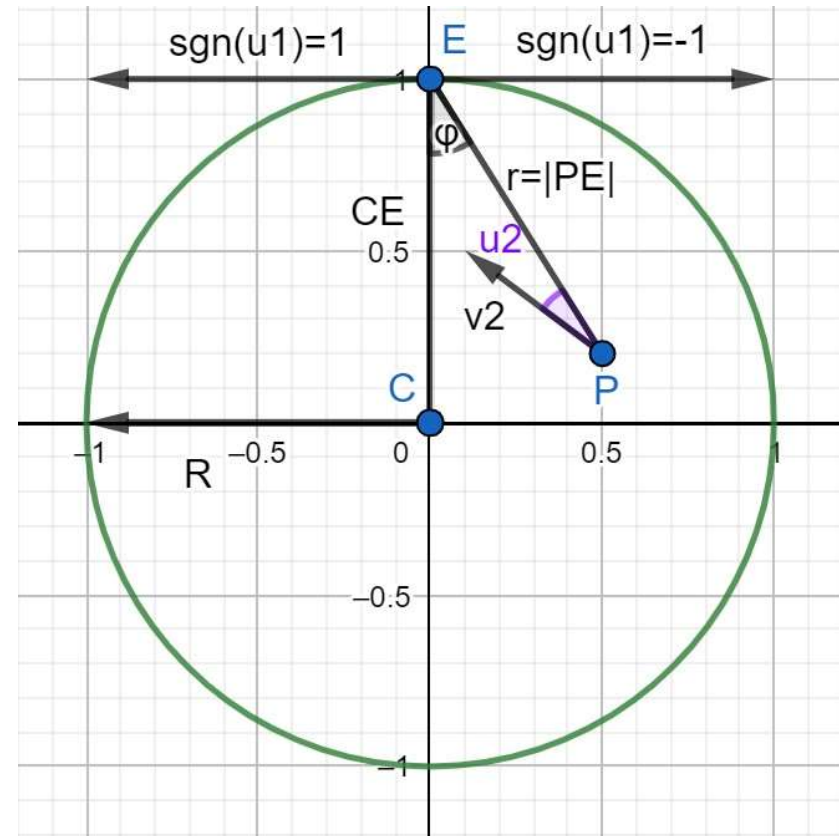
$$(\sin u^{2*}, \cos u^{2*}) \parallel \left(-\frac{1}{r} \frac{\partial V}{\partial \phi}, -\frac{\partial V}{\partial r} \right)$$

What do these strategies mean?

u^{1*} is the sign of a dot product between $\left(\frac{\partial V}{\partial \phi}, \frac{\partial V}{\partial r} \right)$ and $\mathbf{a} = \left(\left(\frac{R \cos}{r} - 1 \right), \sin \phi \right)$.

\mathbf{a} is the $(\dot{\phi}, \dot{r})$ for the simplified problem with speed $v_P = 0$ along \overrightarrow{PE} . So **E** is trying to make ϕ larger so that $v_2 \cos u^2$ can be as small as possible and it will take longer for **P** to minimize r .

However, u^{2*} does not tell much.



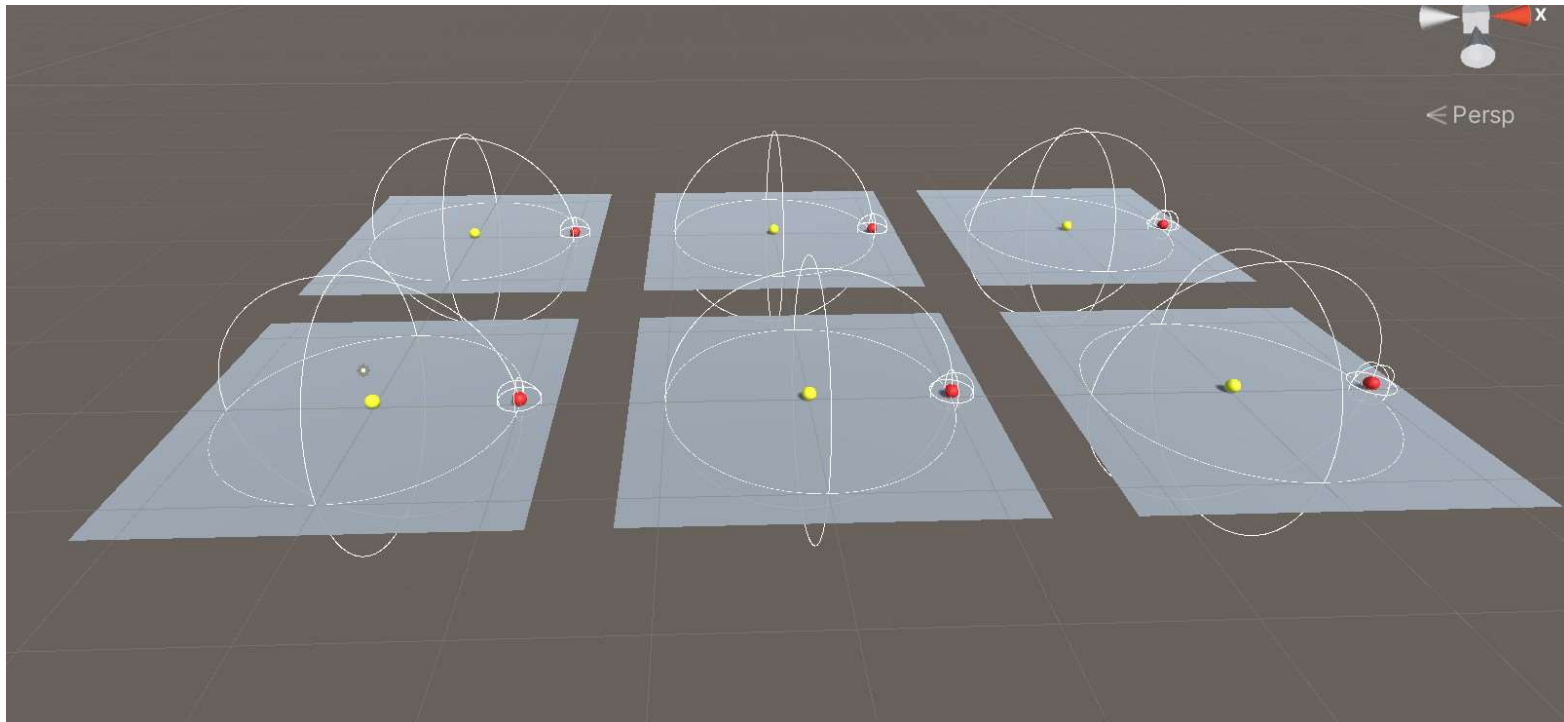
Tools for simulation

Unity with ml-agents is used to simulate the game.

- 6 training regions concurrently.
- Red sphere for P , yellow sphere for E .
- Large gizmos: boundary of the game $\|\mathbf{x}\| = R = 10$
- Small gizmos around P : criteria for catching $\|\mathbf{x}_P - \mathbf{x}_E\| \leq 1.5$ (account for non-negligible size).



<https://unity.com/>



Tools for simulation

Proximal Policy Optimization is used to train the agents (players)

- Actor-critic (online policy) Reinforcement Learning: critic estimates the value function, actor updates the policy distribution.
- Better than A2C and TRPO in most continuous control environments.

Hyperparameters

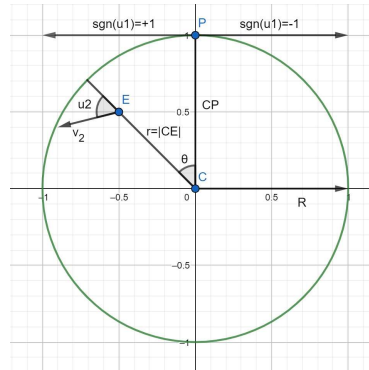
- Batch size: 512
- Buffer size: 5120
- Learning rate: 3.0×10^{-4} , decreasing linearly
- Maximum steps for each team: 10^6
- Neural network: 3-layer MLP with 512 neurons in each layer
- Discount factor: $\gamma = 0.99$

Self-play (for non-cooperative agents with inverse reward)

- Snapshots: taken every 20,000 steps
- Training team switches: every 100,000 steps
- Opponent's policy changes: every 2,000 steps, 50% chance using latest policy.

Original Problem: Setup

$$\text{Let } l = \|PE\|, m = \frac{\overrightarrow{CE} \cdot \overrightarrow{CP}}{\|CE\| \|CP\|}.$$



	Pursuer	Evader
Instantiation	Randomly on the circle	At origin
Observation	$\mathbf{s_P} = (x_P, z_P, x_E, z_E, l, m)$	$\mathbf{s_P} = (x_E, z_E, x_P, z_P, l, m)$
Action	$a_P \in [-1,1]$, speed + direction	$a_E \in [-1,1]$, direction
Reward	+ m at each step +1000 if successfully catches E −1000 if E escapes	− m at each step +1000 if successfully escapes −1000 if caught

Given the Actions, position of **P** and velocity of **E**:

$$\mathbf{x_P} = (10 \cos(\theta_P + a_P dt), 10 \sin(\theta_P + a_P dt))$$

$$\mathbf{v_E} = (v_2 \cos(\theta_E + a_E \pi), v_2 \sin(\theta_E + a_E \pi))$$

Original Problem: Results

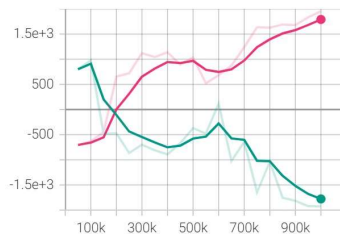
When $v_E < 0.217v_P$:

Note that a_P is the angular velocity, the actual speed is $v_P = a_P R = 10a_P$.

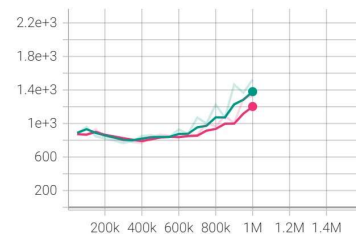
So, we can take $v_E = 1 < 0.217 \cdot 10 = 2.17$

Environment Red is pursuer. Blue is evader.

Cumulative Reward
tag: Environment/Cumulative Reward

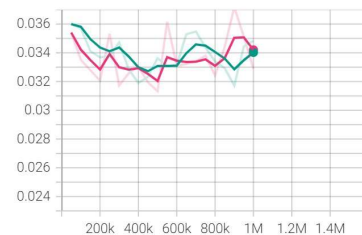


Episode Length
tag: Environment/Episode Length

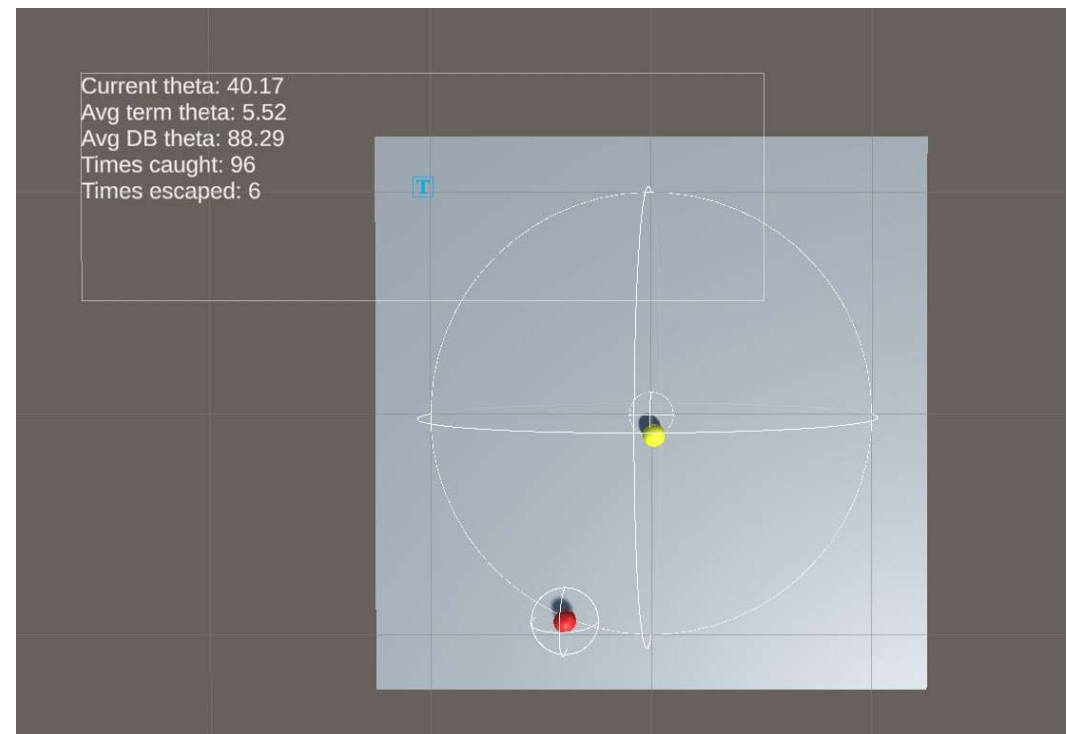
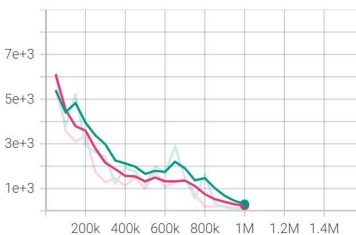


Losses

Policy Loss
tag: Losses/Policy Loss



Value Loss
tag: Losses/Value Loss

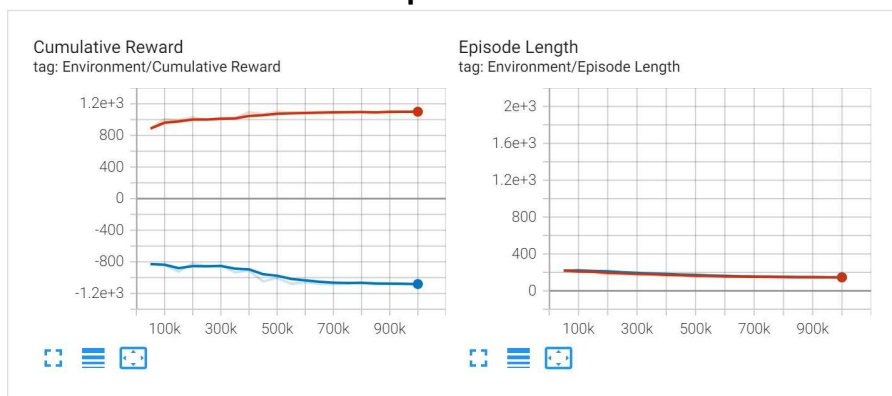


Original Problem: Results

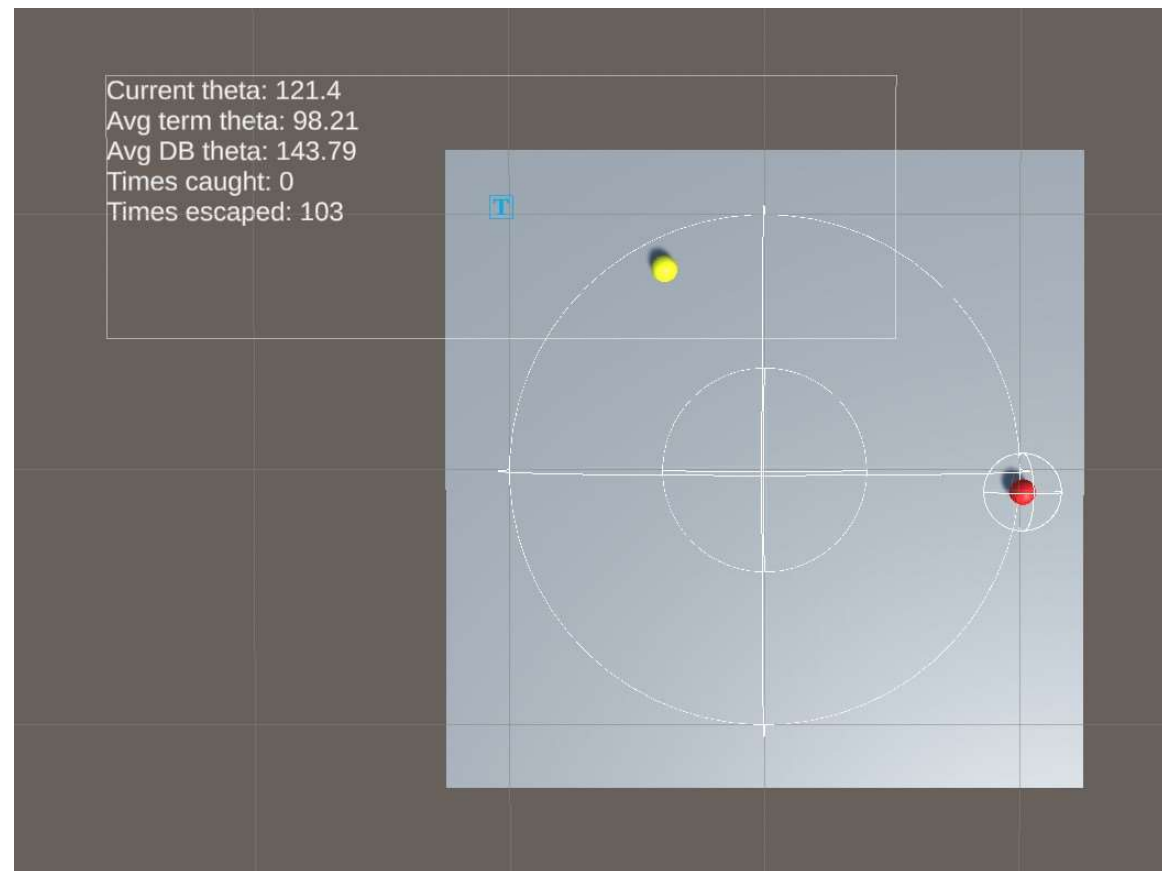
When $v_E > 0.217v_P$:

Take $v_E = 4$

Environment Blue is pursuer. Red is evader.

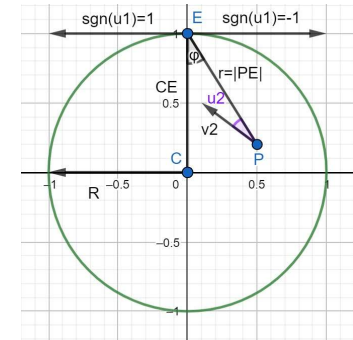


Losses



Inverse Problem: Setup

Let $r = \|PE\|$, $m = \frac{\overrightarrow{CE} \cdot \overrightarrow{CP}}{\|CE\| \|CP\|}$, $\theta = \text{atan2}(\overrightarrow{PE})$.
P will be hard reset to origin if $\|\mathbf{x}_P\| > 10$.



	Pursuer	Evader
Instantiation	At origin	Randomly on the circle
Observation	$\mathbf{s}_P = (x_P, z_P, x_E, z_E, r, m)$	$\mathbf{s}_P = (x_E, z_E, x_P, z_P, r, m)$
Action	$a_P \in [0.5, 1.0] \times [-1, 1]$, speed + direction	$a_E \in [-1, 1]$, speed + direction
Reward	$+0.5 - \frac{r}{20}$ at each step $+1000$ if successfully catches E -1000 if E survives for 15 sec	$+\frac{r}{20} - 0.5$ at each step $+1000$ if survives for 15 sec -1000 if caught

Given the Actions, velocity of **P** and position of **E**:

$$\mathbf{v}_P = (v_2 a_{P1} \cos(\theta + a_{P2}\pi), v_2 a_{P1} \sin(\theta + a_{P2}\pi))$$

$$\mathbf{x}_E = (10 \cos(\theta_E + a_E dt), 10 \sin(\theta_E + a_E dt))$$

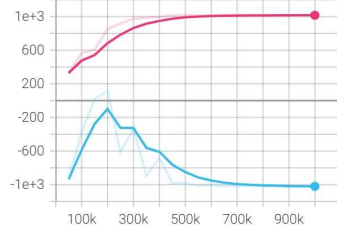
Inverse Problem: Results

Now, we don't know the optimal strategies for the agents.
So, for v_P , instead of a constant speed, it is used as the maximum speed.

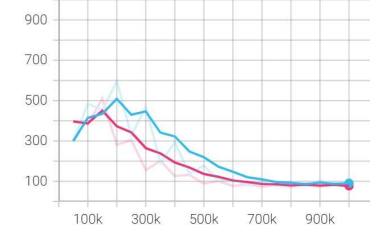
Take $v_P = 8 = 0.8v_E$, this means that **P** can have speed 4~8 based on my setup.

Environment Red is pursuer. Blue is evader.

Cumulative Reward
tag: Environment/Cumulative Reward

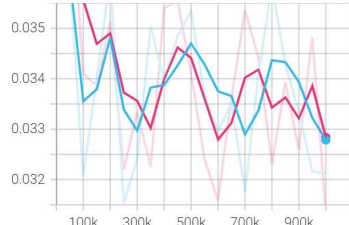


Episode Length
tag: Environment/Episode Length

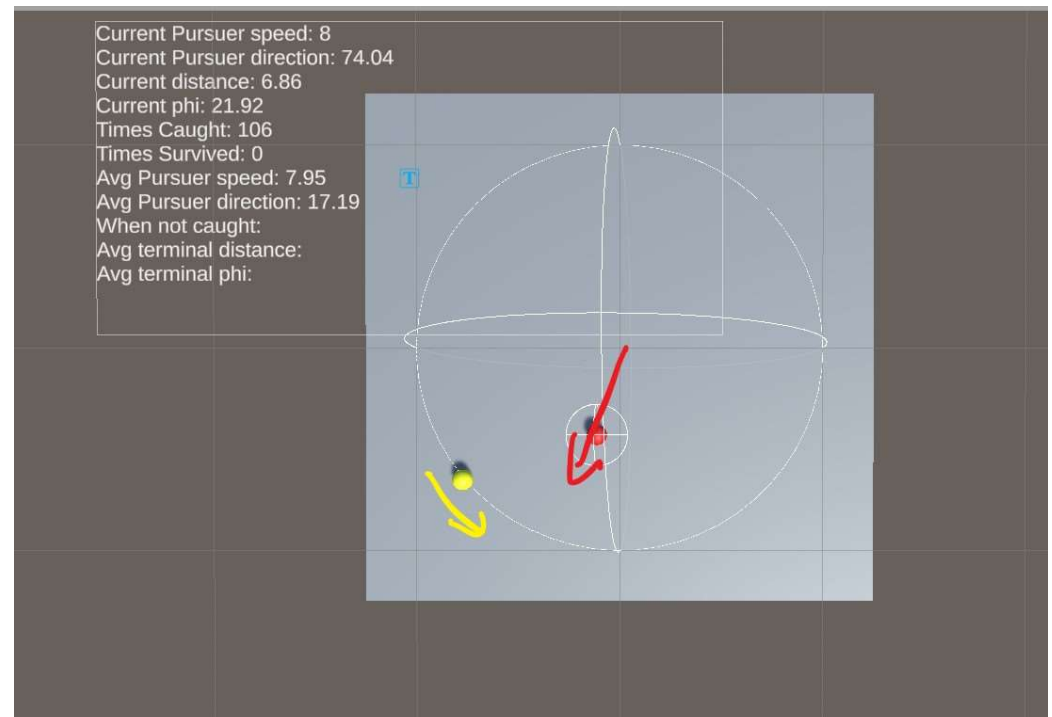
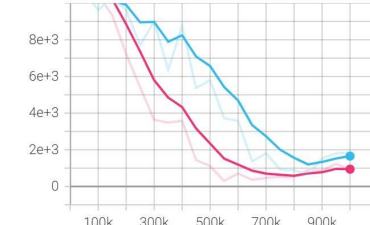


Losses

Policy Loss
tag: Losses/Policy Loss



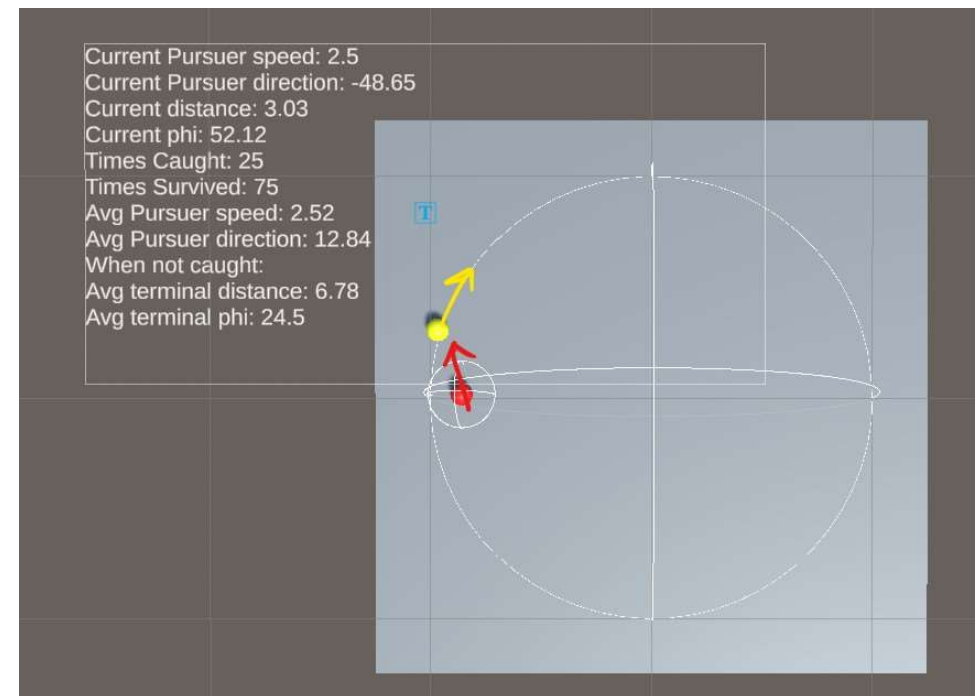
Value Loss
tag: Losses/Value Loss



Inverse Problem: Results

Take $v_P = 5 = 0.5v_E$, this means that **P** can have speed 2.5~5 based on my setup.

Red is pursuer. Green is evader.



Conclusion

What I have achieved:

- Analysis of “The Lady in the Lake” problem and its inverse problem.
- Reinforcement Learning simulation to verify the results of the original problem and analyze some behaviors of the inverse problem under different velocity conditions and different winners.

Limitations in simulation:

- Non-negligible size of the agents
- Relaxed terminal condition, can be improved by using collision detection.
- Potentially asynchronous calculation of angles, causing the game simulation to be not exactly zero-sum.
- Agents might learn better if the actions are speeds (v_x, v_z) directly instead of velocity + angle.

Limitations of the project:

- Unable to analytically solve the inverse problem. May need a different setup.
- Not tested with information advantages in either party.
- v_E can also be non-constant in the original problem for analysis.

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Q&A

Questions?

