

# PONG

- With a dash of AI



# Overview

Problem Statement

Project Goals

The Implementation

Demo

# The Problem Statement

- To build a python-based app from scratch that plays pong
- Interfaces:

Player vs Player

Player vs Computer

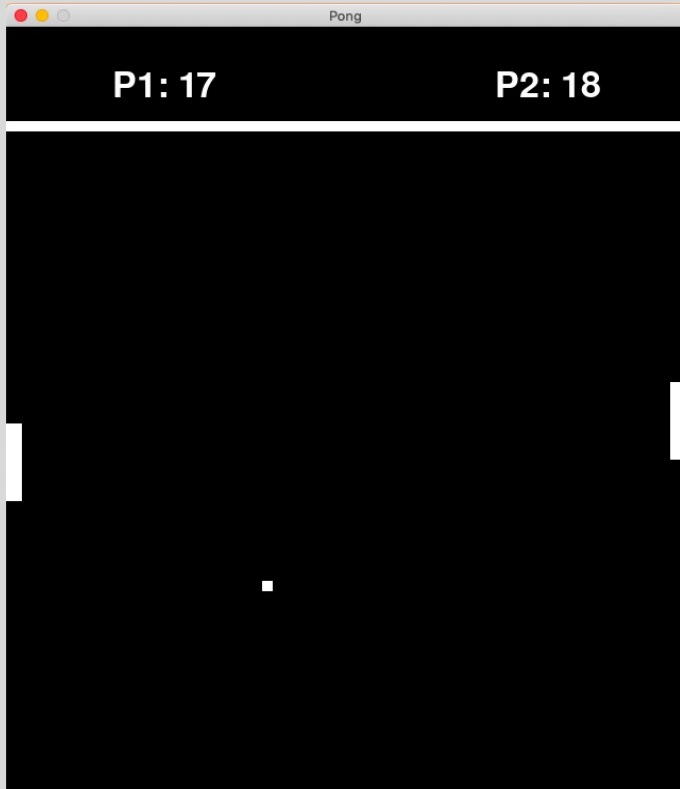
Something Just Like This



# Project Goals

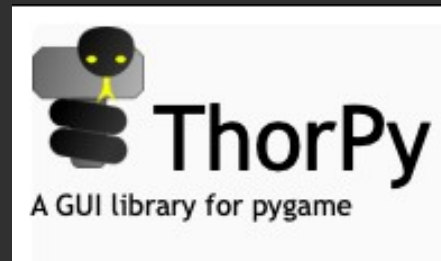
1. **Build a GUI**
2. **Create the game interface**
3. **Create AI models**
4. **Train the AI models**

# The Implementation





os



shutil

random



# LIBRARIES



sys



math

importlib

# Building the GUI

Reactions, text, buttons  
added using Thorpy

```
# create a reaction that calls the function when space is pressed
space = thorpy.Reaction(reacts_to=pygame.KEYDOWN,
                        reac_func=spacedown,
                        event_args={"key": pygame.K_SPACE})

# escape quits the game
esc = thorpy.Reaction(reacts_to=pygame.KEYDOWN,
                     reac_func=escape,
                     event_args={"key": pygame.K_ESCAPE})

# initialize the window
application = thorpy.Application((640, 740), "Pong")

# create all elements
title = thorpy.make_text(text="PONG", font_size=50, font_color=(255, 0, 0))
instr_text = "Press Space to Play"
instr = thorpy.make_text(
    text=instr_text, font_size=30, font_color=(50, 100, 50))
elements = [title, instr]

# make a box to add the elements
box = thorpy.Box.make(elements=elements)
box.fit_children(margins=(30, 30))
box.center()
box.set_main_color((220, 220, 220, 180))

# generate background and add element box to it
background = thorpy.Background.make(
    color=(0, 0, 0), elements=[box])

# add reaction to window
background.add_reaction(space)
menu = thorpy.Menu(background)

# activate the objects
menu.play()
```

# Creating the game interface

Key events, moving  
objects using Pygame

```
# control loop
while 1:

    # set frame rate
    clock.tick(60)
    # black screen
    screen.fill((0, 0, 0))
    # set repeat rate for keys, press event registers in intervals of 100 ms only
    pygame.key.set_repeat(100, 100)

    # event loop
    for event in pygame.event.get():

        # exit on quit event
        if event.type == pygame.QUIT:
            # do something?
            return

        # key is pressed
        elif event.type == pygame.KEYDOWN:

            # move the left pad if w or s is pressed
            if event.key == pygame.K_w:
                player1.moveup()
            if event.key == pygame.K_s:
                player1.movedown()

            # move the right pad if down or up arrow key is pressed
            if event.key == pygame.K_UP:
                player2.moveup()
            if event.key == pygame.K_DOWN:
                player2.movedown()

        # pressed key is released
        elif event.type == pygame.KEYUP:

            # stop moving left pad when w or s is released
            if event.key == pygame.K_w or event.key == pygame.K_s:
                player1.movepos = [0, 0]
                player1.state = "still"
```



# Creating the AI models

## 5 Models:

- Random movement
- 2 Ball following
- 2 RL based

```
import os.path as path
from app import data
```

```
def make_prediction(ball_x, ball_y, pad_x, pad_y):
```

```
    if(ball_y < pad_y + 40):
        # move up if ball is 40 pixels below pad's centre
        pred = 1
        return pred

    elif(ball_y > pad_y + 40):
        # move down if ball is 40 pixels above pad's centre
        pred = -1
        return pred

    else:
        # don't move at all
        pred = 0
        return pred
```

```
def get_prediction(side):
```

```
    # get ball and pad positions
    ball_x, ball_y, angle, v = data.get_ball_pos()
    if(side == "left"):
        pad_x, pad_y = data.get_pad_pos("left")
    elif(side == "right"):
        pad_x, pad_y = data.get_pad_pos("right")

    # make and return action
    prediction = make_prediction(ball_x, ball_y, pad_x, pad_y)
    return prediction
```

# Reinforcement Learning

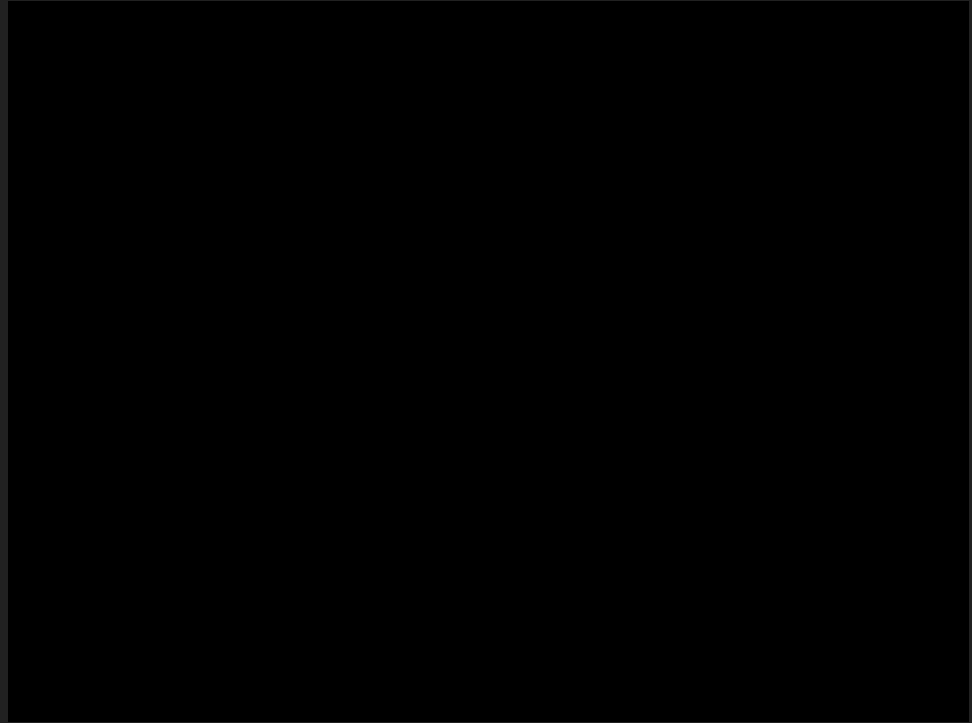
- Teaching the AI how to play by rewarding it for the correct action
- Can be done using different algorithms: Policy Gradients, Q-Learning

## Policy Gradients:

1. Start with random actions
2. Play a few games
3. Collect rewards
4. Modify actions to maximize reward
5. Repeat

## Training the AI models

Train the RL based AI models using the other AI models or with the player



DEMO

# Next Steps

1. **More training**
2. **More models**
3. **Convert to executables for smooth distribution across all systems**

Questions?