

Using Differential Equations and Artificial Intelligence in Modeling Fritz

Team Number: 1148, Problem C

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Overview

- Introduction of the Problem
- Initial Brainstorming Process
- Differential Modeling of Food Object
- General Assumptions
- Differential Modeling of Fritz
- Code
- Results
- Conclusion



The Problem

- Meet Fritz! He cannot catch food tossed to him
- Is Fritz clumsy?
- Is his owner bad at tossing food?



Initial Brainstorming

- 3D Probability Density Function for Modeling of Fritz's Likelihood of Catching Food
- Arc Length Parametrization of a Curve in \mathbb{R}^3

$$s(t) = \int_0^t \|\mathbf{r}'(u)\| du$$

where $\mathbf{r}(t) = \langle \mathbf{x}(t), \mathbf{y}(t), \mathbf{z}(t) \rangle$, a parametrized curve through \mathbb{R}^3 space

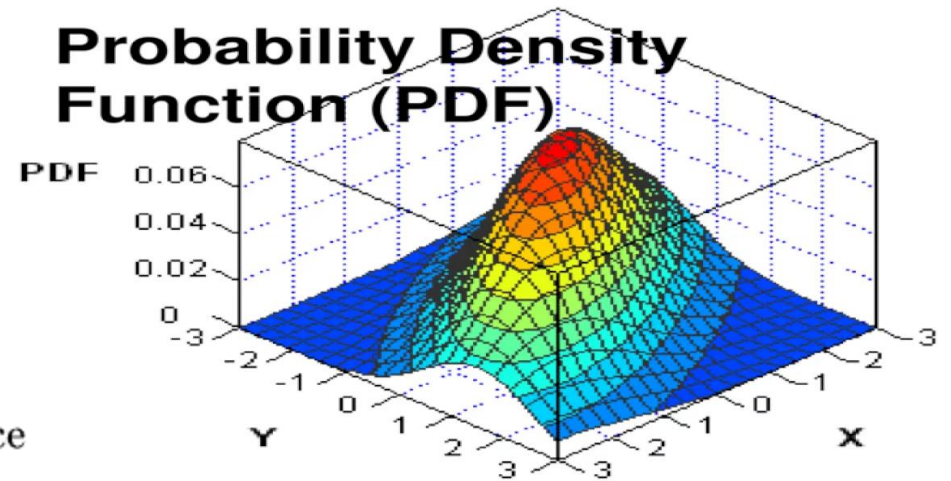


Photo credits: <https://vdocuments.mx/probability-density-function-pdf.html>

Differential Equations-Air Resistance

- Differential Equations (No air Resistance)

$$\ddot{y} = -9.8$$

$$\ddot{x} = 0$$

- Differential Equations (Air Resistance)

$$\ddot{y} - \frac{\rho C_d A}{2m} \dot{y}^2 = -9.8$$

$$\ddot{x} - \frac{\rho C_d A}{2m} \dot{x}^2 = 0$$

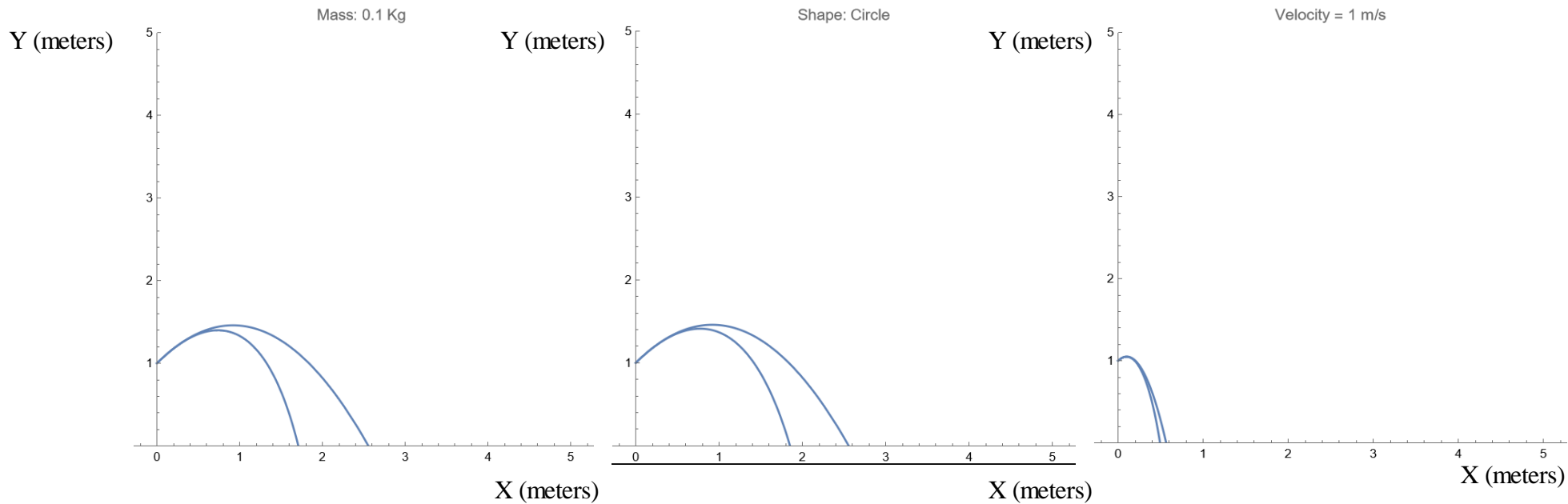
ρ - Air Density

C_d - Coefficient of Drag

A - Reference Area

m - Mass























Differential Equations-Solutions



Changing Mass

Changing Shape

Changing Velocity

	0.38		1.16
	0.42		1.17
	0.47		1.20
	0.50		1.55
	0.59		1.55
	0.80		1.60
	1.05		1.98
	1.17		2.00
	1.17		2.05
	1.38		2.20
	1.42		2.30

Air Density Remains Constant

Graphs Generated in Mathematica

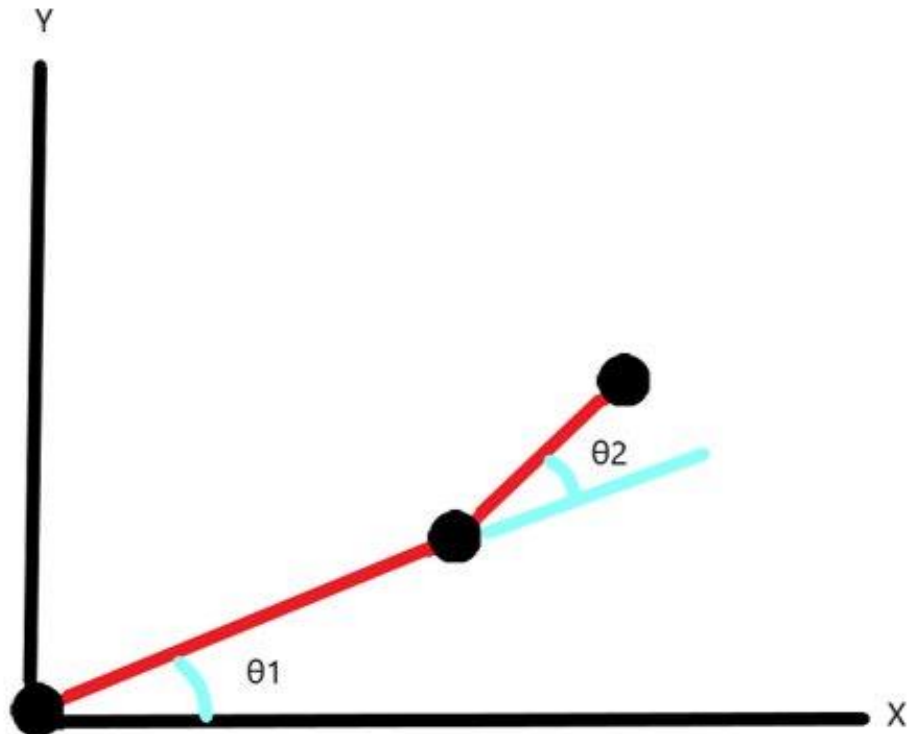
Assumptions Moving Forward

- Air resistance will be neglected, therefore the size of the object does not matter
- Launch angle will remain constant
- Motion and model will be in two dimensions
- No wind

General Model - Using AI

- Reflects Learning
- Variables can be changed
- More complexity than Basic Ordinary Differential Equations
- Applicable to Modern Work Environment
- Follows Differential Motion

General Layout - Fritz



- Two Link Robot
 - Simplification
 - Ease of Programming

- Kinematic Equations

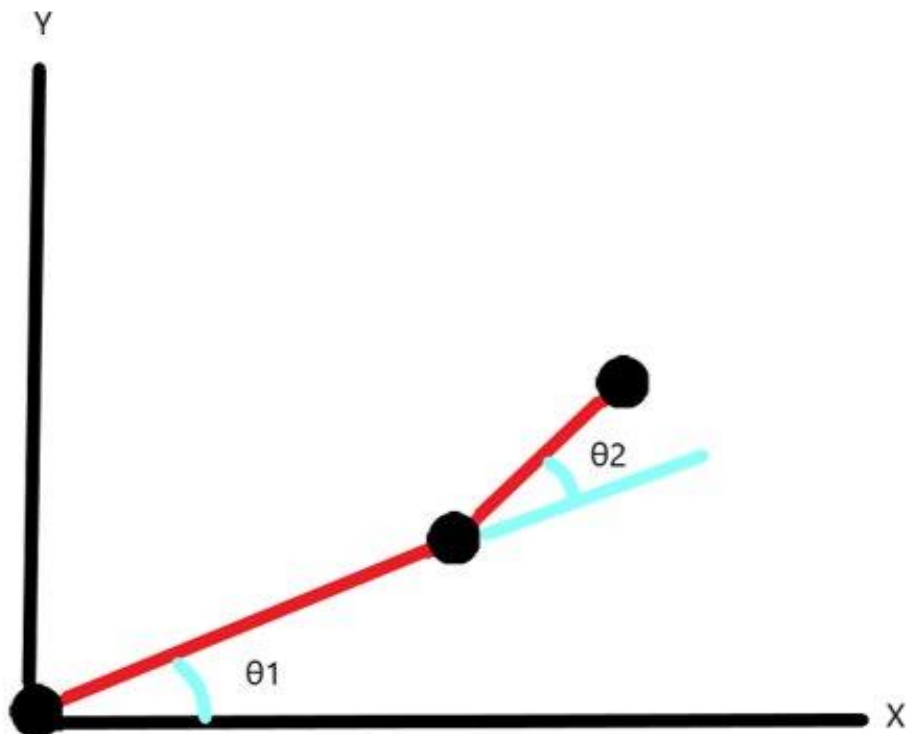
$$\begin{pmatrix} x_1 \\ y_1 \end{pmatrix} = \begin{pmatrix} l_1 \cos(\theta_1) + l_2 \cos(\theta_1 + \theta_2) \\ l_1 \sin(\theta_1) + l_2 \sin(\theta_1 + \theta_2) \end{pmatrix} \quad \begin{array}{l} 0 \leq \theta_1 \leq \pi \\ -\frac{\pi}{2} \leq \theta_2 \leq \frac{\pi}{2} \end{array}$$

- Inverse Kinematics

$$\theta_2 = \cos^{-1} \left(\frac{x^2 + y^2 - l_1^2 - l_2^2}{2l_1 l_2} \right)$$

$$\theta_1 = \tan^{-1} \left(\frac{y}{x} \right) - \tan^{-1} \left(\frac{l_1 \sin(\theta_2)}{l_1 + l_2 \cos(\theta_2)} \right)$$

Differential Motion - Fritz



Starting with Kinematic Equations:

$$\begin{pmatrix} x_1 \\ y_1 \end{pmatrix} = \begin{pmatrix} l_1 \cos(\theta_1) + l_2 \cos(\theta_1 + \theta_2) \\ l_1 \sin(\theta_1) + l_2 \sin(\theta_1 + \theta_2) \end{pmatrix} \quad \begin{array}{l} 0 \leq \theta_1 \leq \pi \\ -\frac{\pi}{2} \leq \theta_2 \leq \frac{\pi}{2} \end{array}$$

Remember Angles are a Function of Time:

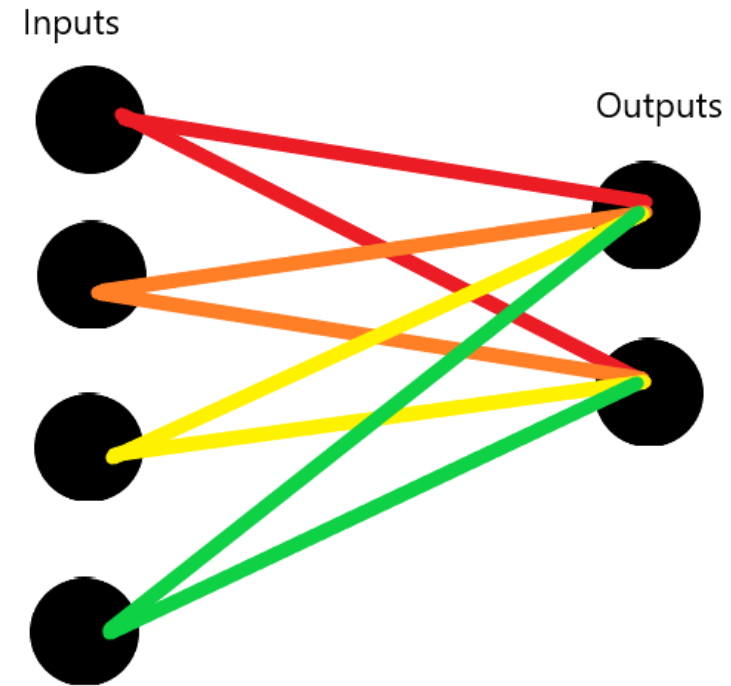
$$\theta_1 = \theta_1(t) \quad \theta_2 = \theta_2(t)$$

Computing Derivative W.R.T. Time:

$$\begin{pmatrix} \dot{x} \\ \dot{y} \end{pmatrix} = \begin{pmatrix} -l_1 \dot{\theta}_1 \sin(\theta_1) - l_2 (\dot{\theta}_1 + \dot{\theta}_2) \sin(\theta_1 + \theta_2) \\ l_1 \dot{\theta}_1 \cos(\theta_1) + l_2 (\dot{\theta}_1 + \dot{\theta}_2) \cos(\theta_1 + \theta_2) \end{pmatrix}$$

AI Implementation

- **NeuroEvolution of Augmented Topologies (NEAT)**
- Inputs
- Outputs
- Activation Function
- Population Size
- Fitness Function
- Max Generations



Values for AI

- **Inputs** – Theta 1 & 2, Fritz to Food Distance, Food X&Y Position and Time
- **Outputs**- Increase or Decrease to Theta 1 &2
- **Activation Function** – Hyperbolic Tangent
- **Population Size** - 100 Per Generation
- **Fitness Function** – "Reward" for Catches, Scaled "Punishment" for Misses
- **Max Generations** – Approx. 500 Generations

Code Part 1

```

1 import pygame
2 import math
3 from random import randint
4 import neat
5 import os
6
7 SCREEN_WIDTH = 1366
8 SCREEN_HEIGHT = 768
9 screen = pygame.display.set_mode((SCREEN_WIDTH, SCREEN_HEIGHT))
10 pygame.display.set_caption("SCUDEM")
11 clock = pygame.time.Clock()
12
13 # Constant Parameters
14 # dog
15 dog_angle1 = 1
16 dog_angle2 = 0
17 dog_length1 = 100
18 dog_length2 = 50
19 dog_base_pos_x = 200 # Relative to Cord System
20 dog_base_pos_y = 0 # Relative to Cord System
21 dog_delta_angle = 0.05
22 dog_mouth_rect = 0
23 dog_lines = 0
24 # ball
25 ball_x_pos = 1000
26 ball_y_pos = 100
27 ball_angle = .5
28 ball_velocity = 9
29 ball_rect_ball = 0
30 catches = 0
31
32
33 def to_pygame(x, y):
34     return x, -y + SCREEN_HEIGHT - 30
35
36
37 def distance_calc(dog, ball):
38     dog_pos = dog.base_pos_x + (
39         dog.length1 * math.cos(dog.angle1) + dog.length2 * math.cos(dog.angle1 + dog.angle2)), -1 * (
40         dog.length1 * math.sin(dog.angle1) + dog.length2 * math.sin(
41             dog.angle1 + dog.angle2)) + SCREEN_HEIGHT
42     ball_pos = ball.ball_x_pos - 5, -ball.ball_y_pos + SCREEN_HEIGHT
43     distance = ((dog_pos[0] - ball_pos[0]) ** 2 + (dog_pos[1] - ball_pos[1]) ** 2) ** 1 / 2
44     return distance
45
46
47 usage
48 class Dog:
49     def __init__(self, color):
50         self.angle1 = dog_angle1
51         self.angle2 = dog_angle2
52         self.length1 = dog_length1
53         self.length2 = dog_length2
54         self.base_pos_x = dog_base_pos_x # Relative to Cord System
55         self.base_pos_y = dog_base_pos_y # Relative to Cord System
56         self.delta_angle = dog_delta_angle
57         self.mouth_rect = dog_mouth_rect
58         self.lines = dog_lines
59         self.color = color
60         self.distance = 100000000
61         self.min_distance = 1000000000000
62         self.dog_x_pos = 0
63         self.dog_y_pos = 0
64         self.caught = False
65
66     def draw(self):
67         def to_robot(x, y):
68             return x + self.base_pos_x, -y + SCREEN_HEIGHT - 30
69
70         pygame.draw.line(screen, (0, 0, 0), to_pygame(x:0, -10), to_pygame(SCREEN_WIDTH, -10), width=10)
71         pygame.draw.circle(screen, self.color, to_robot(x:0, y:0), radius=10)
72         line_body = pygame.draw.line(screen, self.color, start_pos=to_robot(x:0, y:0),
73             end_pos=to_robot(self.length1 * math.cos(self.angle1),
74                 self.length1 * math.sin(self.angle1)), width=5)
75
76         pygame.draw.circle(screen, self.color,
77             to_robot(self.length1 * math.cos(self.angle1),
78                 self.base_pos_y + self.length1 * math.sin(self.angle1)), radius=10)
79         line_head = pygame.draw.line(screen, self.color,
80             to_robot(self.length1 * math.cos(self.angle1),
81                 self.length1 * math.sin(self.angle1)),
82             to_robot(self.length1 * math.cos(self.angle1) + self.length2 * math.cos(
83                 self.angle1 + self.angle2),
84                 self.length1 * math.sin(self.angle1) + self.length2 * math.sin(
85                     self.angle1 + self.angle2))),
86             width=5)
87         pygame.draw.circle(screen, self.color,
88             to_robot(self.length1 * math.cos(self.angle1) + self.length2 * math.cos(
89                 self.angle1 + self.angle2),
90                 self.length1 * math.sin(self.angle1) + self.length2 * math.sin(
91                     self.angle1 + self.angle2)), radius=10)
92         self.dog_x_pos = self.length1 * math.cos(self.angle1) + self.length2 * math.cos(self.angle1 + self.angle2)
93         self.dog_y_pos = self.length1 * math.sin(self.angle1) + self.length2 * math.sin(self.angle1 + self.angle2)
94         self.mouth_rect = pygame.Rect(
95             self.base_pos_x + (
96                 self.length1 * math.cos(self.angle1) + self.length2 * math.cos(self.angle1 + self.angle2)) - 5,
97             -1 * (self.length1 * math.sin(self.angle1) + self.length2 * math.sin(
98                 self.angle1 + self.angle2)) + SCREEN_HEIGHT - 35, 10, 10)
99         pygame.draw.rect(screen, (0, 0, 0), self.mouth_rect, width=1)
100
101         self.lines = [
102             line_head,
103             line_body
104         ]
105
106     def reset(self):
107         self.angle1 = dog_angle1
108         self.angle2 = dog_angle2
109         self.length1 = dog_length1
110         self.length2 = dog_length2
111         self.base_pos_x = dog_base_pos_x # Relative to Cord System
112         self.base_pos_y = dog_base_pos_y # Relative to Cord System
113         self.delta_angle = dog_delta_angle

```

Code Part 2

```
112 self.mouth_rect = dog_mouth_rect
113 self.lines = dog_lines
114
115 self.distance = 100000000
116 self.min_distance = 1000000000000
117 self.dog_x_pos = 0
118 self.dog_y_pos = 0
119 self.caught = True
120
121
122 1 usage
123 class Ball:
124     def __init__(self, ):
125         self.ball_x_pos = ball_x_pos
126         self.ball_y_pos = ball_y_pos
127         self.ball_angle = ball_angle
128         self.ball_velocity = ball_velocity
129         self.rect_ball = ball_rect_ball
130         self.x_vel = 0
131         self.y_vel = 0
132
133     def draw(self, time):
134         self.ball_x_pos -= self.ball_velocity * math.cos(self.ball_angle) * time
135         self.ball_y_pos += self.ball_velocity * math.sin(self.ball_angle) * time - 4.9 * time ** 2
136         self.x_vel = self.ball_velocity * math.cos(self.ball_angle)
137         self.y_vel = self.ball_velocity * math.sin(self.ball_angle) - 9.8 * time
138         pygame.draw.circle(screen, (255, 0, 0), to_pygame(self.ball_x_pos, self.ball_y_pos), radius=10)
139         self.rect_ball = pygame.Rect(self.ball_x_pos - 5, -self.ball_y_pos + SCREEN_HEIGHT - 35, 10, 10)
140         pygame.draw.rect(screen, (0, 0, 0), self.rect_ball, width=1)
141
142 4 usages (4 dynamic)
143     def reset(self):
144         self.ball_x_pos = ball_x_pos
145         self.ball_y_pos = ball_y_pos
146         self.ball_angle = ball_angle
147         self.ball_velocity = ball_velocity
148         self.rect_ball = ball_rect_ball
149         self.x_vel = 0
150         self.y_vel = 0
151
152 def main(genomes, config):
153     pygame.time.wait(5000)
154     pygame.display.set_caption("SCUDEM")
155     pygame.init()
156     run = True
157     time = 0
158     round = 0
159     catches = 0
160     ball = Ball()
161     dogs = []
162     ge = []
163     nets = []
164     outputs = []
165     for _, g in genomes:
166         net = neat.nn.FeedForwardNetwork.create(g, config)
167         nets.append(net)
168         dogs.append(Dog((randint(0, 255), randint(0, 255), randint(0, 255))))
169         g.fitness = 0
170         ge.append(g)
171
172     # Just for manual play
173
174     while run:
175         time += 1 / 60
176         screen.fill((255, 255, 255))
177         for event in pygame.event.get():
178             if event.type == pygame.QUIT:
179                 run = False
180
181                 pygame.quit()
182                 quit()
183
184                 pygame.display.set_caption(str(catches))
185                 if len(dogs) == 0:
186                     run = False
187                     print(catches)
188                     break
189                 ball.draw(time)
190                 for x, dog in enumerate(dogs):
191                     dog.distance = distance_calc(dog, ball)
192                     if distance_calc(dog, ball) < dog.min_distance:
193                         dog.min_distance = distance_calc(dog, ball)
194                     output = nets[x].activate(
195                         (dog.angle1, dog.angle2, dog.distance, ball.ball_x_pos, ball.ball_y_pos, ball.x_vel, ball.y_vel,
196                          dog.dog_x_pos, dog.dog_y_pos, time)) #
197
198                     if output[0] > 0.5:
199                         if dog.angle1 >= 0 and dog.length1 * math.sin(dog.angle1 - dog.delta_angle) + dog.length2 * math.sin(
200                             dog.angle1 + dog.angle2) > 0:
201                             dog.angle1 -= dog.delta_angle
202
203                     elif output[0] < -0.5:
204                         if dog.angle1 <= math.pi and dog.length1 * math.sin(
205                             dog.angle1 + dog.delta_angle) + dog.length2 * math.sin(dog.angle1 + dog.angle2) > 0:
206                             dog.angle1 += dog.delta_angle
207
208                     else:
209                         pass
210
211                     if output[1] > 0.5:
212                         if dog.angle2 >= -math.pi/2 and dog.length1 * math.sin(dog.angle1) + dog.length2 * math.sin(
213                             dog.angle1 + dog.angle2 - dog.delta_angle) > 0:
214                             dog.angle2 -= dog.delta_angle
215
216                     elif output[1] < -0.5:
217                         if dog.angle2 <= math.pi/2 and dog.length1 * math.sin(dog.angle1) + dog.length2 * math.sin(
218                             dog.angle1 + dog.angle2 + dog.delta_angle) > 0:
219                             dog.angle2 += dog.delta_angle
220
221                 for x, dog in enumerate(dogs):
222                     if ball.ball_y_pos > -50:
223                         if ball.rect_ball.clip(dog.mouth_rect):
224                             ge[x].fitness += 10000
225                             dog.caught = True
226                             catches += 1
227                             dogs.pop(x)
228                             nets.pop(x)
229                             ge.pop(x)
230
231                             elif ball.rect_ball.collideobjects(dog.lines) and dog.caught == False:
232                                 ge[x].fitness -= 100 + dog.min_distance/100
233                                 dogs.pop(x)
234                                 nets.pop(x)
```

AI Model

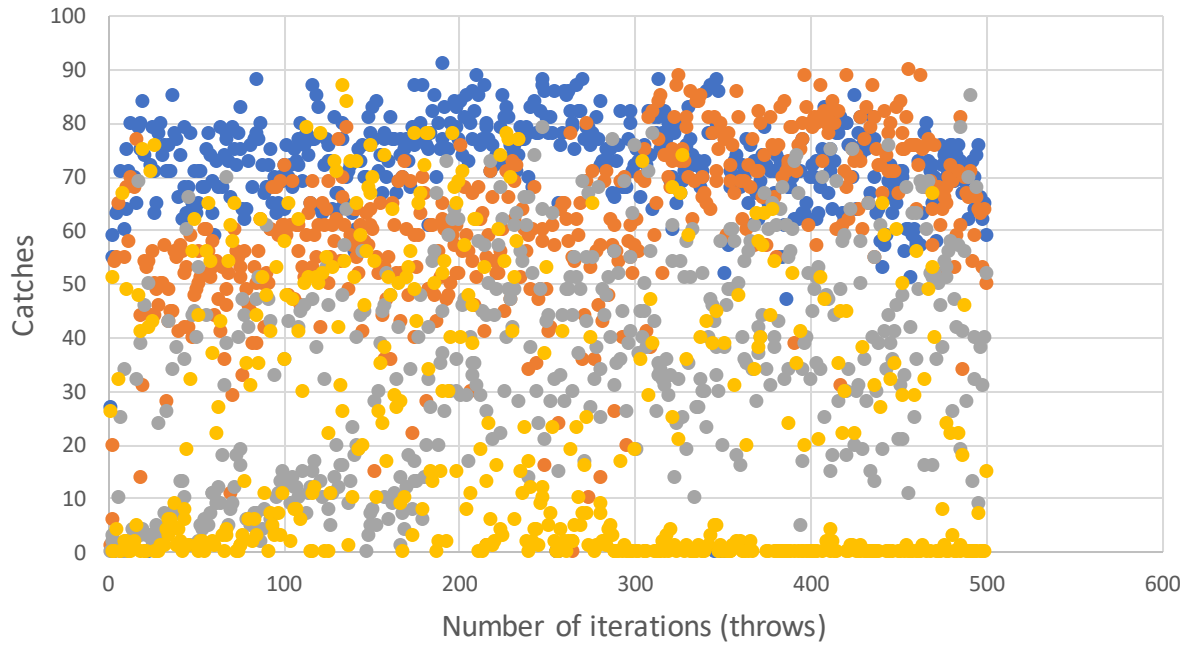


AI Manipulation

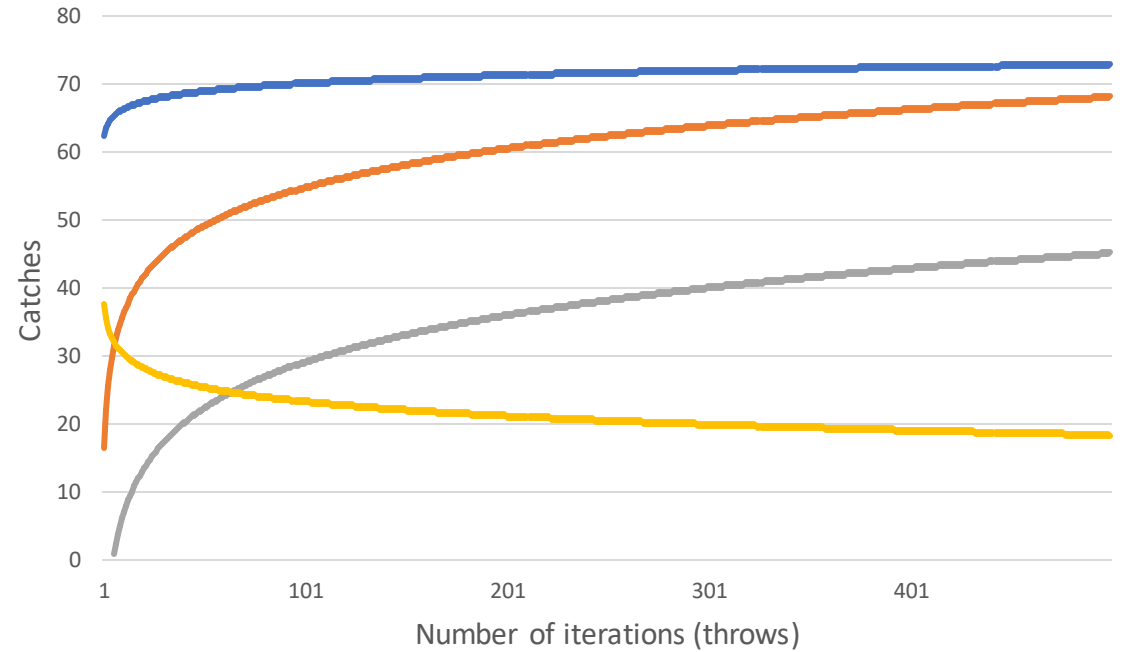
- Variables Modified for Food:
 - Velocity
 - Starting Position of Food (Distance and Height)
- Variable Modified for Fritz:
 - Iterations
 - Number of Fritz's per iteration
- Additional Assumptions
 - Food will always cross through domain of Fritz's movement
 - Once food touches ground or Fritz, food is marked as not caught

Results

Catches Per Iteration



Catches Per Iteration



- Constant initial velocity and initial food position
- Varied initial velocity, constant initial food position
- Constant initial velocity, varied initial food position
- Varied initial velocity and initial food position

Conclusion

- How precisely does the dog's brain have to estimate the orientation of his head and eyes in order to best predict the location of the item?
- How much time will the dog have to orient its mouth to the correct position?
- Are certain foods more difficult to catch due to different effects of air friction?
- Does the height at which the object is thrown matter?
- What are the most important aspects of the dog's perceptions and behaviors for the dog's success?
- Use your analysis of the situation to decide if Fritz is just clumsy or if his owner is being mean to the dog on the Internet.

Future Directions

- Improve model for multiple variable changes
- Delay reaction time
- Introduce noise for different-sized foods
- Change dog dimensions and mouth size
- Add air friction inside the model
- Expand model to 3D

Citations

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