Birds: Game Design

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Game Overview

- Game Rules
 - Birds pursue and eat food
 - Hawks pursue and eat Birds
 - Birds flee Hawks
 - Player strategically places food to guide Birds to the Nest
- Scoring
 - Some number of birds reach the nest
 - Birds eat some number of food items
- Al Mechanism
 - Decisive Pathfinding
 - Allow Birds to find Food
 - Allow Hawks to find Birds
 - Allow Birds to flee Hawks

AI Overview

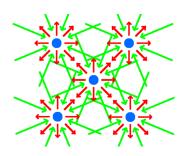
- Extension of Collaborative Diffusion
- Model reality by diffusing metrics about pursued agents throughout environment
- Agents can affect different metric layers
- Agents make decisions based on metrics in their immediate neighborhood
 - Agents compute a weighted summation of various metric layers
 - Agents of different types can weight metrics differently (including negative weights)
 - Agents move to the maximum valued cell after weighted summation

Why Use Diffusion?

- Path Finding for arbitrary number of entities / goals
- Complexity is constant as entities and goals increase
- By using multiple diffusion layers and weights, complex behaviors emerge
 - Division of Labor
 - Simultaneous Goals
 - Dynamic goal-changing based on circumstance

Example: Flocking behavior





Basic algorithm

```
For each diffusion layer: //Food, Explore, Bird, Etc.

diffusionMatrix[pos.ofeveryentityinlayer] := maxDiffusion

//maxDiffusion == 1, usually. This is called the 'seed'

All other cells are set to 0

For Numlterations: //how many times to diffuse the entire layer

For each x, y in diffusionMatrix:

If environment[x][y] is not an obstacle:

Set diffusionMatrix[x][y] = Sum of Neighboring Cells *

diffusionRate
```

Diffusion Algorithm

Sum of Neighbors Operator:

$$S(X): S_{i,j} = X_{i+1,j} + X_{i,j+1} + X_{i-1,j} + X_{i,j-1} + X_{i-1,j-1} + X_{i+1,j+1} + X_{i+1$$

- Input for game map of size $m \times n$:
 - Metric Seed: $M_{m \times n}: M_{i,j} = \left\{ \begin{array}{ll} 0 & \text{if no metric seed exists at cell } (i,j) \\ 1 & \text{otherwise} \end{array} \right.$

The originating metric values to be diffused

 $\qquad \qquad \mathsf{Obstacle\ Mask:}\ \ O_{m\times n}: O_{i,j} = \left\{ \begin{array}{cc} 0 & \text{ for obstacle\ cells}\ (i,j) \\ 1 & \text{ otherwise} \end{array} \right.$

A mask to hide obstacles. Has value of 0 where obstacles exist, 1 everywhere else

▶ Diffusion Matrix: $D \in \mathbb{R}_{m \times n}$

Existing diffusion array to be further diffused. Initially all zeros

- ▶ Diffusion Rate: $d: d \in \mathbb{R}, d < 1$
 - Diffusion rate scalar, controls how much one cell bleeds into another
- Iteration count: c : c ∈ Z, c > 0
 Numer of diffusion iterations to apply. Controls rate at which metric diffuses through environment
- Return Value: Diffusion Matrix: $\hat{D} \in \mathbb{R}_{m \times n}$: Metric seed values diffused
- Precomputed Values:
 - $\qquad \qquad \qquad \mathbf{Metric \ Mask:} \ \ \bar{M}_{m \times n} : \bar{M}_{i,j} = \left\{ \begin{array}{ll} 1 & \text{ if } M_{i,j} = 0 \\ 0 & \text{ otherwise} \end{array} \right.$

A mask to hide metric seed cells.

Need only be computed when M changes a 0 valued cell to non-zero value.

- Has value of 0 where metric seeds exist, 1 everywhere else
- NeighborCoefficient: n = 8 count of neighboring cells used in diffusion
- Diffusion Algorithm (\odot deonotes element-wise matrix multiplication): FOR k = 1 : c $\hat{D} \leftarrow \frac{d}{c}(S(D) \odot \bar{M}) + M$

Diffusion Example

Environment:				•		
Metric Seed Matrix:						
		L 0	0	0	0	0 7
		0	0	0 1 0	0	0
	M =	0	0	1	0	0
		0	0	0	0	0
		Lo	0	0	0	0
Metric Mask:		Γ1	1	1	1	1 7
		1	1	1 1 0 1	1	1
	$\bar{M} =$	1	1	0	1	1
		1	1	1	1	1
		4	- 1	- 1	- 1	-

Simple environment depicting a food item placed on a nonwrapping game map with obstacles

A numeric matrix with 0 at empty cells and 1 at cells containing food

A mask to prevent modifying values in food containing cells. Helps ensure that values of diffusion array are bounded

A mask to prevent diffusion through obstacle cells

Agent Utilization of Metrics

- ► Agents are given an array of metric matrices representing the metrics of interest in their neighborhood of the game map
- ▶ Neighborhood is an array of 3×3 matrices
- Agents apply a scalar weight to each layer representing their strategy
- Agents perform element-wise sum the weighted matrices
- Agents select the maximum valued cell as their destination

Game Representation - Behavior

```
'Metrics':{
  'FoodMetric': {'rate': 0.2.'iters': 1}.
    'BirdMetric': {'rate': 0.9.'iters': 15}.
    'HawkMetric': {'rate': 0.7, 'iters': 15},
    'NestMetric':{'rate':0. 'iters':0}
'Entities':{
'Food':{ 'Eats':[],
 'Weights':{},
 'MapChar': 'F',
 'Image': 'apple.png',
 'Affects':{'FoodMetric':1},
 'Moves': False.
 'StartSkill':0
'Bird': { 'Eats': ['Food'],
'Weights':{
 'FoodMetric':1,
    'BirdMetric':0
    'HawkMetric':-1
},
'MapChar':'B',
 'Image':'bird.png',
 'Affects':{'BirdMetric':'skill'}.
 'Moves':True.
 'StartSkill':1
'Hawk': { 'Eats': ['Bird', 'Food'],
 'Weights':{
  'FoodMetric':1,
'BirdMetric':1,
    'HawkMetric':0
 ,
'MapChar':'H',
'Image': 'hawk.png',
 'Affects':{'HawkMetric':1},
'Moves':True.
 'StartSkill':0
'Nest': { 'Eats': ['Bird'],
'Weights':{},
 'MapChar':'N',
 'Image': 'nest.png',
 'Affects':{'NestMetric':'skill'},
 'Moves': False.
 'StartSkill':0
 ,
InsertEntity':[{'entity':'Food','label':'food','count':15},{'entity':'Bird','label':'bird','count':12}, {'entity':'Hawk','la'
 'StartPaused': True.
'Win': {'NestMetric':5}
                                                                                     4 □ > 4 □ > 4 □ > 4 □ >
```

Game Representation - Map

```
0000000000000000
0............
0....B....O
0...........
0............
O...H.......
0.....
0....0
0.......F0....0
0....0
0....0
0....0
0.....
0....0...0
0.....
```

Extensions and Enhancements

- Referencing Affected Layers Subtract own effect
- Extend to 3d
- Continuous movement over discrete map
- Genetic algorithm to evolve weights