

Algorithm Project Angry Birds

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Professor: Dr. Mir Hossein Dezfoulian

Fatemeh Saeedi 40212358020



Sahba Mirshahi 40212358040

GitHub link:

https://github.com/Farinaz-Saeedi/AngryBird

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Introduction

In this project, we implemented a program to guide birds from their home cities to enemy cities using the A* search algorithm . The project focuses on creating an efficient pathfinding solution while considering obstacles and enemy spies . The main goal is to ensure that the birds not only hit their targets successfully but also cause maximum damage . This project is implemented in C++ and utilizes concepts such as object-oriented programming , dynamic programming , and greedy algorithms . In the following sections , we provide details about the project structure , implementation , and workflow . We hope you enjoy reading this report!

Project Structure

Class Organization:

The project is designed using object-oriented programming principles and consists of several classes, explained as follows:

City: Represents each city on the map and serves as the parent class. Two classes Enemy (for pig cities) and Home (for bird cities), inherit from it.

Bird: Represents each bird and stores its attributes, including damage, name, distance, and others.

Scenario : Represents a scenario , and seven scenario classes inherit from it . Each scenario has its own class and is executed based on user requests .

Controller: The most important class, where the main functions for managing the program are implemented.

Input Format:

In the following, the input files and their formats are explained.

Birds.txt: This file stores information about the birds. The user can refer to this file to become familiar with the birds and ultimately select the desired ones. The data format is as follows: Bird Name - Total Distance - Uncontrolled Distance - Spy Tolerance - Damage - Type

```
Venomtail 2500 500 3 100 A1
Skyrage 2500 500 4 25 A3
Ironbeak 2500 500 2 130 A2
Stonewing 5000 500 2 90 B1
Fury 5000 500 0 300 B2
Blaze 3000 700 2 110 C1
Viper 2900 900 1 10 C
Titan 10000 500 2 1488 D1
```

• The information of the birds

Cities.txt: In this file, the user must enter the cities along with the details.

The input format is as follows:

First, the number of cities, followed by each line containing:

- City Name
- Longitude Latitude
- City Type (Normal/Enemy/Home)
- Presence of Spy (0/1)
- Defense Level (for enemy cities)
- Capacity (for home cities)

```
11
city0 600 0 Normal 0
city1 0 800 Home 0 3
city2 200 600 Home 1 4
city3 500 400 Home 1 3
city4 500 -200 Normal 0
city5 400 -100 Normal 1
city6 200 -50 Normal 0
city7 100 200 Normal 0
city8 -400 100 Enemy 1 4
city9 -400 -300 Enemy 1 2
city10 -200 0 Normal 0
```

An example to write cities.txt

Scenario.txt: The user must enter the scenario number (from 1 to 7). The program then accesses the corresponding scenario file where the scenario-specific information is entered. The input formats for each scenario are:

<u>Scenario 1:</u> Number of birds - Bird name (Titan)

Scenario 2: Number of birds, followed by each line:

Bird Name - Slingshot (Home) Name

Scenario 3: Number of birds type, followed by each line:

Bird Name - number of birds

Scenario 4: Number of birds , followed by each line : Bird Name - Base Name

Scenario 5: Number of nights (5), number of birds, then for each bird:

Bird Name-Slingshot Name

Scenario 6: Number of birds, then for each bird:

Bird Name - Slingshot Name

<u>Scenario 7:</u> Number of nights (7) - Desired Damage - Number of birds , then for each bird : Bird Name - Cost - Slingshot Name

SpiesInScen5.txt: This file belongs to Scenario 5, where new spies are discovered each night. The user must enter the spy information for each night in this file. The input format is as follows:

Each line contains the Night Number-Name of the Cities that have a spy

Key Functions of Each Class

Bird Class:

This class stores all the attributes of a bird, such as name, damage, distance, path, and type (A/B/C/D). Setter and getter functions are implemented for managing these attributes. The type of birds is handled using an enum.

City Class:

This class represents a city, whether it is Normal, Enemy, or Home. The type of the city is managed using an enum. It stores the common information shared between all cities, such as the city name, coordinates (longitude and latitude), and the city type.

Enemy Class:

This class inherits from the City class . It represents an enemy city . In addition to the common city information , it has a defense level attribute that it manages . The class also provides a pushReachBird function to add birds that have reached their destination to a particular vector , and a setBirdPath function to set the paths assigned to those birds .

Home Class:

This class also inherits from the City class and stores the capacity of each slingshot . Since this class represents a slingshot , a vector is defined to keep track of the birds in each slingshot place . push , delete , and getter functions are also implemented for it . The reduceCapacity function in this class decreases the slingshot capacity by one unit for further implementation in the desired scenarios .

Scenario Class:

This class serves as the parent class , and Scenarios 1 to 7 inherit from it . Two virtual functions , readInputs and printOutput, are implemented in this class and overridden by each scenario . The readInputs function is used to read the information for each scenario , while the printOutput function is used to print the outputs of each scenario , including damage and the path of each bird .

The **readBird** function in this class takes the name of the birds selected by the user for the scenario and , if the bird exists in the bird list file , reads its information and stores it in the corresponding vector .

Controller Class:

The most important class is the Controller, which implements the main pathfinding algorithms and serves as the core of the project. Below, we describe some of the key functions of this class:

readCities: This function is responsible for reading city information from a file, creating the appropriate objects and storing them. After reading the number of cities, it retrieves the common information for each city and depending on the city type, reads additional information (capacity for Home cities and defense level for Enemy cities). Finally, it creates the corresponding city objects.

readScenario: This function is responsible for creating a specific scenario and reading its corresponding inputs. It takes the scenario number as input, and then makes a shared_ptr of type Scenario. After checking the scenario number, the appropriate object is created, and its inputs are read. Finally, the created object is returned.

shootDownBird: This function is responsible for killing birds. First, the target enemy is identified. Then, all birds that have been detected by this enemy are collected. The birds are sorted based on their destructive power, since the enemy prioritizes destroying the most dangerous ones first. Finally, depending on the enemy's defense level, an appropriate number of birds are killed (shot down).

run: This function is responsible for executing the entire flow of the program . First , all cities are read . Then , the desired scenario number is read , and by calling readScenario , an object of the Scenario class is created . Finally , the printOutput function of the selected scenario is executed . This displays the outputs related to the chosen scenario .

findBestPair: This function finds the best possible path for a given bird from a starting city to one of the enemy cities. For each enemy city, the path is calculated using the A* algorithm, and then the path cost and the number of spies along it are evaluated. If the current path is better than the previous ones, the path information is updated. Finally, the function returns the name of the best destination along with the result of the A* algorithm.

A-star:

Function Overview: The aStar function is an implementation of the A* pathfinding algorithm, customized for this project. Its purpose is to find the optimal path between two given cities (start and goal) for a specific bird object, while considering both geometric distances and domain-specific penalties such as the presence of spies or enemy defense systems.

Unlike the classic A* that only minimizes geometric distance, this version incorporates penalty costs, making the route not only shorter but also safer and more efficient according to the bird's capabilities.

Function Signature:

```
bool Controler::aStar (
   std::string start , // starting city name
   std::string goal , // target city name
   Bird myBird , // the bird object
   std::vector<std::shared_ptr<City>> & path , // output: final path
   ld & totalDistance , // output: real geometric distance
   ld & cost , // output: final penalized path cost )
```

Algorithm Workflow:

The algorithm begins by mapping city names to indices and checking for the trivial case where the start and goal are the same . All travel costs are initialized to infinity except for the starting city , which is set to zero . A priority queue is then used to always select the city with the lowest estimated cost , defined as the real travel cost plus the heuristic distance to the goal .

During the main loop , the city with the lowest estimated cost is extracted . If it is the goal , the path is reconstructed , the total distance is calculated , and a check ensures that the bird has enough range to complete the journey . If successful , the algorithm terminates with a valid solution . Otherwise , the algorithm explores neighboring cities : it computes the Euclidean distance for each move , applies penalties for spies and enemy defenses , and updates the cost if a better path is found . The search continues until the goal is reached . Since the heuristic is based on Euclidean distance alone , it remains admissible , while penalties are considered in the actual travel cost . This ensures that the chosen path is not just the shortest geometrically , but also the most **feasible** and **optimal** under the given constraints .

Scenario 1:

This class overrides two functions: readInputs and printOutput. The readInputs function reads the input file for Scenario 1, creates birds according to their type and quantity, assigns them to their respective homes, and updates each bird's home information. The printOutput function finds the best path and target for each bird using findBestPairFor, prints the bird's path, handles birds that cannot reach their target, executes attacks on enemy cities, and finally calculates and displays the total demolition caused by the surviving birds.

Scenario 2:

This class overrides two functions: readInputs and printOutput.

readInputs: This function reads the input data for Scenario 2 from a text file . It creates Bird objects with their respective names , assigns each bird to its home city , and updates the corresponding Home objects with these birds .

printOutput: This function simulates the activities of the birds in Scenario 2. For each bird, the best path to an enemy city is determined, it is checked whether the bird can reach the target and destroy it, and its status is updated. Then enemy attacks are executed, the total damage caused by the surviving birds is calculated, and the results are displayed in the output.

Scenario 3:

This class uses the Hungarian algorithm to find optimal assignments and ultimately execute the actions . Below , we explain the functions and their implementation details .

OptionScen3 Structure: This structure stores a possible flight option for a bird in Scenario 3. It includes the bird's index, its home city, the target city, the path from home to targe t, and the total cost of taking this path.

The hungarianMin function implements the Hungarian (Kuhn-Munkres) algorithm to find an optimal assignment between two sets: in this context, birds and target cities. The function takes a profit matrix as input, where each row represents a bird and each column represents a target city, and the values indicate the benefit or cost of assigning a specific bird to a specific city. The algorithm works by transforming the profit matrix into a square cost matrix and iteratively finding augmenting paths to minimize the total assignment cost. At the end, it returns a vector that maps each bird (row) to the index of its assigned city (column), ensuring that the total cost is minimized and the assignment is optimal.

buildProfitMatrix: This function constructs the profit matrix. Each row corresponds to a bird, and each column corresponds to a target option. The matrix stores the cost of assigning each bird to each target. High default values are used for impossible assignments, ensuring that only valid bird-target pairs are considered in later optimization algorithms like the Hungarian method.

assignOptions : This function generates all possible assignments of birds to targets in Scenario 3 . Birds are sorted based on their demolition power . The function loops through homes that have available capacity , and for each bird , the path and cost (g-cost) to every enemy city are calculated using the A* algorithm . Valid options , including the bird , home , target , path , and cost are stored and the capacities of the homes are updated accordingly .

Scenario 4:

It follows a process similar to Scenario 2.

Scenario 5:

The readInputs function loads scenario data from a file . It reads the number of nights and iterates through all birds , assigning each one a home city corresponding to its slingshot . Each bird is then added to its home's vector , ensuring that all homes keep track of their assigned birds . This prepares the scenario for managing bird locations and launches efficiently .

Scenario Strategy: In this scenario, to maintain continuity over five nights while maximizing destruction, a specific strategy is applied. For each night, different options are generated and stored, where each option contains the bird index, the starting city, the target city, the chosen path, and the expected damage. If an option has a spy-tolerance level lower than the number of spies encountered along its path, it means the bird will definitely reach its target and cause damage. Such guaranteed options are stored in the vector **firstOptions**, and only one of them is dispatched to an enemy city each night.

After every night, new spies are discovered, which means that reaching the target is no longer guaranteed for all remaining options. Therefore, the other options are stored in an unordered_map, where the enemy city's name is used as the key, and the value is a vector containing all options directed toward that city.

To ensure guaranteed damage on a given night, the function <code>getWeakEnemy</code> from the Controller class is used to identify the weakest enemy. Then, the algorithm sends a number of birds equal to that enemy's defense level plus one, ensuring destruction. Finally, if any birds remain unused, all remaining options from <code>allOptions</code> are dispatched toward enemy cities.

Scenario 6:

The algorithm used in this class is very similar to the strategy applied in Scenario 5, with the main difference being the absence of the allOptions vector. In this class, the <code>giveBirdsID</code> function assigns a unique ID to each bird, which ensures that removing birds from the vector during any night or step does not disrupt the order. Each bird's index is retrieved using its ID through the <code>getBirdIndex</code> function, allowing safe deletion without errors.

As long as **firstOptions** contains entries , it is used to maintain the attack sequence . Once **firstOptions** is exhausted and its birds have been deployed , birds stored in the **enemyToSecondOptions** map are sent based on the enemy's defense level plus one , ensuring continued attacks . The **night** variable is incremented after each complete attack , and at the end , the total number of consecutive attack nights along with the total damage inflicted is printed .

Scenario 7:

This class handles multi-night attacks with a focus on reaching a target total damage while minimizing the cost of bird deployment. Birds are first loaded from a file with associated costs and home cities. Each bird is assigned a unique ID to ensure safe removal during the nightly attack sequence.

Nightly Attack Execution : For each night , all feasible attack options are generated by computing paths from home cities to enemy cities using the A* algorithm . Each option records expected damage , cost , path distance , and detection risk and a variable to know that if bird is detected or not . The algorithm then uses a knapsack-like strategy to select the combination of birds that maximizes damage .

After each attack, birds used are removed safely from the active list. The process continues until all nights are completed or the target total damage is reached.

KnapsackMinCost Function: This function sorts attack options by damage-to-cost ratio and selects options greedily until the target damage is met. This ensures cost-effective damage allocation across nights. This function also ensures that only the damage from birds that were not killed by the enemy is counted.

Conclusion

This project successfully modeled the problem of coordinating bird attacks on enemy cities under real-world constraints such as limited flight range, defensive penalties, and the presence of spies. By integrating algorithmic approaches like A* for pathfinding and assignment strategies for resource allocation, the system was able to determine effective routes and maximize potential damage.

The main strength of the project lies in its combination of physical constraints (distance and capacity) with strategic factors (defense levels and penalties), resulting in decisions that are both realistic and analytically useful. The modular design also enables scalability, allowing the framework to be extended with new heuristics, optimization techniques, or larger scenarios.

Nevertheless, the approach is not without limitations. The accuracy of results depends on carefully tuned heuristics and cost parameters, and very large-scale scenarios may lead to higher computational overhead.

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