Flappy Bird Q-Learning Implementation

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1 Introduction

This report details the implementation of a Q-learning agent using deep neural networks to play Flappy Bird. The implementation uses raw pixel input which is preprocessed and fed to a CNN.

2 Architecture

2.1 Neural Network Structure

The neural network architecture consists of:

- Input Layer: 4 channels of 84x84 preprocessed frames
- Convolutional Layer 1: 32 filters of size 8x8 with stride 4
- Convolutional Layer 2: 64 filters of size 4x4 with stride 2
- Convolutional Layer 3: 64 filters of size 3x3 with stride 1
- Fully Connected Layer 1: $3136 \rightarrow 512$ neurons
- Output Layer: $512 \rightarrow 2$ neurons (representing actions)

ReLU activation functions are used throughout the network.

2.2 Input Processing

The raw game state is preprocessed by:

- Reshaping to 12x15 matrix
- Resizing to 84x84 pixels
- Normalizing pixel values
- Stacking 4 consecutive frames

3 Q-Learning Implementation

3.1 Key Parameters

• Discount Factor: 0.99

• Initial Epsilon: 0.2

• Final Epsilon: 0.0001

• Number of Iterations: 30,000

• Replay Memory Size: 10,000

• Minibatch Size: 32

• Learning Rate: 0.0001

3.2 Algorithm Components

• Experience Replay: Implemented using a deque with maximum size 10,000

• Epsilon-Greedy Policy: Linear decay from 0.2 to 0.0001

• Q-Value Updates: Using MSE loss and Adam optimizer

4 Training Process

The training process involves:

- Collecting experiences in replay memory
- Sampling random minibatches
- Computing target Q-values using the Bellman equation
- Updating network weights through backpropagation

The agent learns to map raw pixel inputs to Q-values for each possible action (flap or do nothing).

5 Results

5.1 Run 1: Training Parameters

The following parameters were used during training for Run 1:

• Learning Rate (lr): 0.0001

• Number of Actions: 2

• Discount Factor: 0.99

• Initial Epsilon: 0.2

• Final Epsilon: 0.0001

• Number of Training Iterations: 50,000

• Replay Memory Size: 10,000

• Minibatch Size: 32

5.2 Training Performance

Below is a snapshot of the training performance during Run 1:

```
Listing 1: Training Log (Run 1)
```

Episode: 760, Score: 23.6000000000000268, Avg Score: 5.16, Epsilon: 0.0061, Best Episode: 761, Score: -4.49999999999999, Avg Score: 5.10, Epsilon: 0.0059, Best

Checkpoint saved: Episode 762, Score 7.4999999999984

Episode: 762, Score: 7.499999999999994, Avg Score: 5.17, Epsilon: 0.0055, Best S Checkpoint saved: Episode 763, Score 29.30000000000264

Episode: 763, Score: 29.3000000000000264, Avg Score: 5.44, Epsilon: 0.0036, Best Checkpoint saved: Episode 764, Score 15.7999999999999

Episode: 764, Score: 15.7999999999999962, Avg Score: 5.50, Epsilon: 0.0029, Best Checkpoint saved: Episode 766, Score 32.1000000000335

Episode: 766, Score: 32.10000000000335, Avg Score: 5.80, Epsilon: 0.0006, Best

The highest score achieved during training was **32.1**, with an average score of approximately **5.80** by the end of training. The agent's performance improved significantly as epsilon decayed, enabling it to exploit the learned policy.

5.3 Testing Performance

After training, the model was tested over 10 episodes. The results are as follows:

Listing 2: Testing Log (Run 2)

Test Episode 1: Score = 42.80000000000043

Test Episode 2: Score = 14.499999999999973

Test Episode 3: Score = 106.099999999975

Test Episode 4: Score = 11.2999999999998

Test Episode 5: Score = 52.10000000000067

Test Episode 6: Score = 60.90000000000092

Test Episode 7: Score = 72.20000000000033

Test Episode 8: Score = 21.200000000000009

Test Episode 9: Score = 8.79999999999947

Test Episode 10: Score = 85.8999999999884

5.4 Performance Metrics

• Average Test Score: 47.58

• Maximum Test Score: 106.10

• Minimum Test Score: 8.80

The agent achieved an average score of 47.58 during testing, with a maximum score of 106.10.

5.5 Run 2: Training Parameters

The following parameters were used during training for Run 2:

• Learning Rate (lr): 0.001

• Number of Actions: 2

• Discount Factor: 0.99

• Initial Epsilon: 0.2

• Final Epsilon: 0.0001

• Number of Training Iterations: 90,000

• Replay Memory Size: 10,000

• Minibatch Size: 32

5.6 Training Performance

Below is a snapshot of the training performance during Run 2:

Listing 3: Training Log (Run 2)

```
Episode: 1328, Score: 14.799999999999942, Avg Score: 7.90, Epsilon: 0.0048
Episode: 1329, Score: 43.1000000000034, Avg Score: 8.34, Epsilon: 0.0038
Episode: 1330, Score: 73.39999999999932, Avg Score: 9.00, Epsilon: 0.0012
Episode: 1331, Score: 5.99999999999999, Avg Score: 9.07, Epsilon: 0.0011
Episode: 1332, Score: 2.79999999999999, Avg Score: 9.06, Epsilon: 0.0010
Episode: 1333, Score: 26.70000000000156, Avg Score: 9.31, Epsilon: 0.0002
```

Model saved as: $saved_models/model_ep_1333_avg_9.31_20250105_153524.pth$

The highest score achieved during training was 73.40, while the average score improved to **9.31** by the end of training.

Testing Performance 5.7

After training, the model was tested over 10 episodes. The results are as follows:

Listing 4: Testing Log (Run 2)

```
Test Episode 1: Score = 10.199999999999997
Test Episode 2: Score = 142.69999999999333
Test Episode 4: Score = 12.699999999999966
Test Episode 5: Score = 4.699999999999994
Test Episode 6: Score = 174.4999999999324
Test Episode 8: Score = 6.6999999999999983
Test Episode 9: Score = 8.99999999999991
Test Episode 10: Score = 29.400000000000198
```

5.8 Performance Metrics

• Average Test Score: 40.87

• Best Test Score: 174.50

• Worst Test Score: 4.70

The model demonstrated moderate performance during testing, achieving an average score of 40.87. The highest score obtained was 174.50, while the lowest was 4.70.

5.9 Run 3: Training Parameters

The following parameters were used during training for Run 3:

• Learning Rate (lr): 0.0001

• Number of Actions: 2

• Discount Factor: 0.99

• Initial Epsilon: 0.2

• Final Epsilon: 0.0001

• Number of Training Iterations: 12,000

• Replay Memory Size: 10,000

• Minibatch Size: 32

5.10 Training Performance

Below is a snapshot of the training performance during Run 3:

Listing 5: Training Log (Run 3)

```
Episode: 1719, Score: 12.699999999999988, Avg Score: 10.01, Epsilon: 0.0050 Episode: 1720, Score: 3.599999999999988, Avg Score: 9.90, Epsilon: 0.0049 Episode: 1721, Score: 11.2999999999999963, Avg Score: 10.02, Epsilon: 0.0047 Episode: 1722, Score: 33.60000000000022, Avg Score: 10.36, Epsilon: 0.0040 Episode: 1723, Score: 2.3, Avg Score: 10.36, Epsilon: 0.0038 Episode: 1724, Score: -2.300000000000003, Avg Score: 10.33, Epsilon: 0.0037 Episode: 1725, Score: 20.40000000000134, Avg Score: 10.50, Epsilon: 0.0031 Episode: 1726, Score: 15.1999999999996, Avg Score: 10.61, Epsilon: 0.0028 Episode: 1727, Score: 4.6999999999999, Avg Score: 10.60, Epsilon: 0.0026 Episode: 1728, Score: 27.60000000000000007, Avg Score: 10.74, Epsilon: 0.0019 Episode: 1730, Score: 41.30000000000043, Avg Score: 11.10, Epsilon: 0.0009 Episode: 1731, Score: 9.09999999999971, Avg Score: 11.20, Epsilon: 0.0005
```

Model saved as: saved_models/model_ep_1731_avg_11.23_20250110_222552.pth

The highest score achieved during training was **41.30**, while the average score improved steadily to **11.23** by the end of training.

5.11 Testing Performance

After training, the model was tested over 10 episodes. The results are as follows:

Listing 6: Testing Log (Run 3)

Test Episode 8: Score = 6.499999999999984Test Episode 9: Score = 37.6000000000003Test Episode 10: Score = 176.99999999999406

5.12 Performance Metrics

• Average Test Score: 89.08

• Best Test Score: 209.90

• Worst Test Score: 6.50

6 Code Structure

Key components of the implementation:

• Neural Network Class: Implements the CNN architecture

• Frame Preprocessing: Converts raw game state to suitable input

• Training Loop: Implements Q-learning with experience replay

• Testing Framework: Evaluates trained model performance