# **K-Armed Bandits**

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# **K\_ARMED\_BANDITS MODULE**

 $\textbf{class} \ \, \textbf{k\_armed\_bandits.BanditsGame}(K, T, seed=None, non\_stationary\_type='stationary', verbose=True)$ 

Bases: object

Creates an instance of the bandits testbed.

Generates K NormalBandit objects to populate the BanditsGame

... .. attribute:: K

the number of NormalBandits to populate

**type** int

#### verbose

optional parameter used to provide greater detail to stdout

**Type** 

boolean

### run\_greedy(T):

runs an instance of the BanditsGame using the greedy algorithm

### run\_epsilon\_greedy(T, epsilon):

runs an instance of the BanditsGame using the epsilon-greedy algorithm

#### run\_optimistic\_initial\_values(T, initial\_value):

runs an instance of the BanditsGame using the optimistic-greedy algorithm

#### run\_gradient\_bandit(T, alpha):

runs an instance of the BanditsGame using the gradient-bandit algorithm

# run\_epsilon\_greedy(T, epsilon)

Runs an instance of the BanditsGame using the epsilon-greedy algorithm.

## **Parameters**

- T(int) The number of steps to run the algorithm.
- **epsilon** (*float*) The probability of exploring a random action.

#### **Returns**

- numpy.ndarray An array of rewards obtained at each step.
- *numpy.ndarray* An array indicating whether the optimal action was chosen at each step.

#### run\_gradient\_bandit(T, alpha)

Runs an instance of the BanditsGame using the gradient-bandit algorithm.

#### **Parameters**

• **T** (*int*) – The number of steps to run the algorithm.

• **alpha** (*float*) – The learning rate for updating preferences.

#### **Returns**

- numpy.ndarray An array of rewards obtained at each step.
- *numpy.ndarray* An array indicating whether the optimal action was chosen at each step.

#### run\_greedy(T)

Runs an instance of the BanditsGame using the greedy algorithm.

#### **Parameters**

**T** (*int*) – The number of steps to run the algorithm.

#### **Returns**

- numpy.ndarray An array of rewards obtained at each step.
- *numpy.ndarray* An array indicating whether the optimal action was chosen at each step.

# run\_optimistic\_initial\_values(T, initial\_value)

Runs an instance of the BanditsGame using the optimistic-greedy algorithm.

#### **Parameters**

- **T** (*int*) The number of steps to run the algorithm.
- initial\_value (float) The initial optimistic value for all actions.

#### Returns

- numpy.ndarray An array of rewards obtained at each step.
- *numpy.ndarray* An array indicating whether the optimal action was chosen at each step.

# class k\_armed\_bandits.NormalBandit(means, variance=1, verbose=True)

Bases: object

A class to represent an individual bandit with a normal reward distribution.

. . .

#### mean

the average reward to be expected from this bandit

# Type

float

#### variance

the variance associated with the normal reward distribution

#### **Type**

float

#### verbose

optional parameter used to provide greater detail to stdout

#### **Type**

boolean

#### pull():

outputs a randomly generated reward based on the bandit's distribution

#### **pull**(*t*)

Operates this instance of NormalBandit.

A random reward is generated from its defined normal reward distribution, which may vary depending on the time step t.

#### **Parameters**

t (int) – The current time step.

#### Returns

A float value representing the reward.

#### Return type

float

k\_armed\_bandits.plot\_results(avg\_rewards\_greedy, avg\_rewards\_epsilon\_greedy,

```
avg_rewards_optimistic, avg_rewards_gradient, opt_action_props_greedy, opt_action_props_epsilon_greedy, opt_action_props_optimistic, opt_action_props_gradient, best_epsilon, best_alpha, avg_rewards_per_trial_greedy, avg_rewards_per_trial_optimistic, avg_rewards_per_trial_epsilon_greedy, avg_rewards_per_trial_gradient)
```

Plots the results of different algorithms.

#### **Parameters**

- **avg\_rewards\_greedy** (*numpy.ndarray*) Average rewards obtained using the greedy algorithm.
- avg\_rewards\_epsilon\_greedy (numpy.ndarray) Average rewards obtained using the epsilon-greedy algorithm.
- avg\_rewards\_optimistic (numpy.ndarray) Average rewards obtained using the optimistic initial values algorithm.
- avg\_rewards\_gradient (numpy.ndarray) Average rewards obtained using the gradient bandit algorithm.
- **opt\_action\_props\_greedy** (*numpy.ndarray*) Proportion of times the optimal action was chosen using the greedy algorithm.
- opt\_action\_props\_epsilon\_greedy (numpy.ndarray) Proportion of times the optimal action was chosen using the epsilon-greedy algorithm.
- **opt\_action\_props\_optimistic** (*numpy.ndarray*) Proportion of times the optimal action was chosen using the optimistic initial values algorithm.
- **opt\_action\_props\_gradient** (*numpy.ndarray*) Proportion of times the optimal action was chosen using the gradient bandit algorithm.
- **best\_epsilon** (*float*) The best epsilon value found during parameter tuning.
- **best\_alpha** (*float*) The best alpha value found during parameter tuning.

```
k_{armed\_bandits.run\_simulation}(K, T, n\_problems, algorithm, seed=None, non\_stationary\_type='stationary', **kwargs)
```

Runs multiple instances of the BanditsGame and computes average rewards and optimal action proportions.

# **Parameters**

- **K** (*int*) The number of bandits.
- T(int) The number of steps to run each game.
- **n\_problems** (*int*) The number of independent problems to simulate.
- algorithm (str) The algorithm to use ('greedy', 'epsilon\_greedy', 'optimistic\_initial\_values', 'gradient\_bandit').
- **seed** (int, optional) Random seed for reproducibility (default is None).

- **non\_stationary\_type** (*str*, *optional*) Type of non-stationary behavior ('stationary' or specified in non\_stationary\_types).
- **\*\*kwargs** (*dict*) Additional parameters specific to the algorithm.

### Returns

- numpy.ndarray Average rewards across all problems at each step.
- *numpy.ndarray* Proportion of times the optimal action was chosen across all problems at each step.

# NON\_STATIONARY MODULE

## non\_stationary.abrupt\_change(bandits)

Permutes the means corresponding to each bandit.

At each time step, with probability 0.005, permutes the means corresponding to each of the reward distributions.

#### **Parameters**

bandits (list of NormalBandit) - List of bandits whose means need to be permuted.

non\_stationary.drift\_change(previous\_mean, initial\_mean)

Applies a small random drift to the mean.

#### **Parameters**

- **previous\_mean** (*float*) The mean at the previous time step.
- initial\_mean (float) The initial mean of the bandit.

#### Returns

The updated mean after applying the drift.

# Return type

float

 $non\_stationary.generate\_means(initial\_mean, T, non\_stationary\_type)$ 

Generates a sequence of means for the given non-stationary type.

### **Parameters**

- initial\_mean (float) The initial mean of the bandit.
- **T** (*int*) The number of time steps.
- **non\_stationary\_type** (*str*) The type of non-stationary behavior ('drift', 'mean\_reverting', 'abrupt').

#### Returns

An array of means for each time step.

#### Return type

numpy.ndarray

 $\verb|non_stationary.mean_reverting_change||(previous\_mean, initial\_mean)||$ 

Applies a mean-reverting change to the mean.

#### **Parameters**

- **previous\_mean** (*float*) The mean at the previous time step.
- initial\_mean (float) The initial mean of the bandit.

#### Returns

The updated mean after applying the mean-reverting change.

Return type float

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