

# Movie Recommendation System using K-Arm Bandit Problem

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- Each possible movie is modelled as an arm
- Out of 100 movies ( $k = 100$ ) the system presents 10 choices to the user. The user selects a movie based on which the expected reward for that movie is updated.
- We use the epsilon greedy approach to control exploration and exploitation.

## Initialisation:

- We have an array,  $N$  which is set to all zeroes – stores the number of times a movie has been chosen.
- We have an array,  $Q$  set to all zeroes – this is the expected reward for each movie
- Let  $\epsilon = 0.3$

## Training:

- Generate a random number  $r$  between 0 and 1.
- If  $r < 0.3$ :
  - Exploration: select 10 movies at random out of the 100 and present them as choices to the user
- If  $r \geq 0.3$ :
  - Exploitation: Select 10 movies based on the expected reward array  $Q$ . We do this by choosing the movies with the highest values of  $Q$ ; i.e top 10 values from  $Q$ .
- **User Choice:**
  - User chooses one of the 10 movies
  - The number of times that movie has been chosen is incremented by 1 in the  $N$  array :  $N_{\text{movie}} += 1$
  - We now update the expected reward for that movie,  $Q$  using the following formula:  
 **$Q_{\text{new}} = Q_{\text{old}} + (R - Q_{\text{old}}) / N_{\text{movie}}$**

This process is repeated for  $T$  timesteps.