

# "Monty Hall" game analysis

This notebook is to examine the success rates of 3 different types of players

a - never changes his mind b - always changes his mind c- changes his mind 50% of times

```
In [1]: import matplotlib.pyplot as plt
import numpy as np
from numpy import *
```

```
In [2]: def grill_car():
    return np.random.randint(1,4) ## returns the range of [1,4)
```

```
In [3]: def door_first_choose():
    return np.random.randint(1,4)
```

```
In [4]: def host_shows_goat_door(car_door, first_choose):
    return [index for index in (1,2,3) if index != car_door and index != first_choose]
```

```
In [5]: def swithching_decision(player_type, first_choose, goat_door):
    if player_type == 'a': ## sticks with his first choice
        return first_choose

    elif player_type == 'b': ## always switches
        return [index for index in (1,2,3) if index != first_choose and index != goat_door]

    else: ## 50% switch 50% not
        fifty_fifty = np.random.randint(0,2)
        if fifty_fifty == 0: ## switch
            return [index for index in (1,2,3) if index != first_choose and index != goat_door]
        else:
            return first_choose
```

```
In [6]: def check_win(guess, car_door):
    return (guess == car_door)
```

```
In [7]: def play(player_type):

    car_door = grill_car()

    first_choose = door_first_choose()

    second_choose = swithching_decision(player_type, first_choose, host_shows_goat_door(car_door, first_choose))

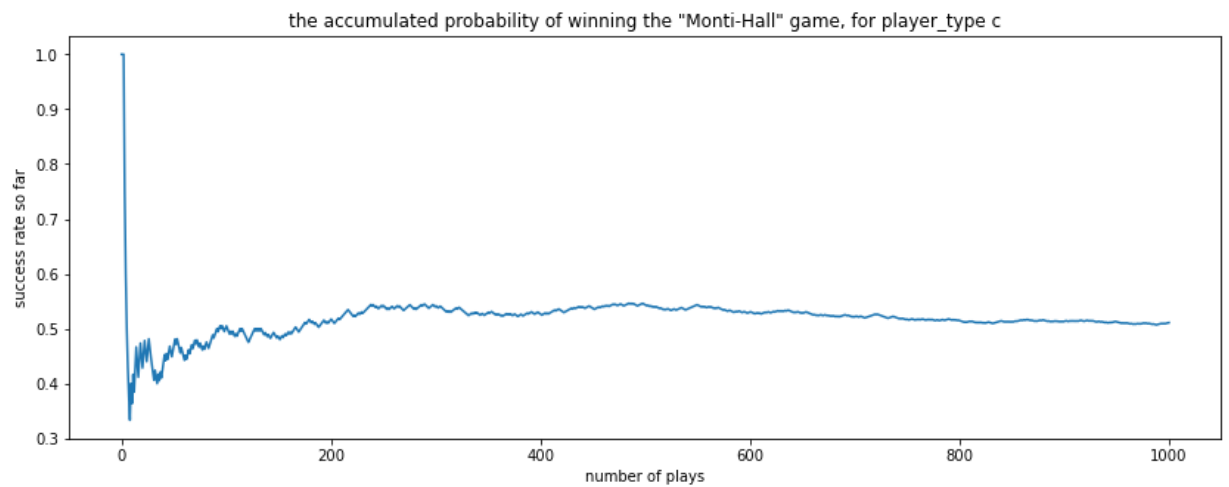
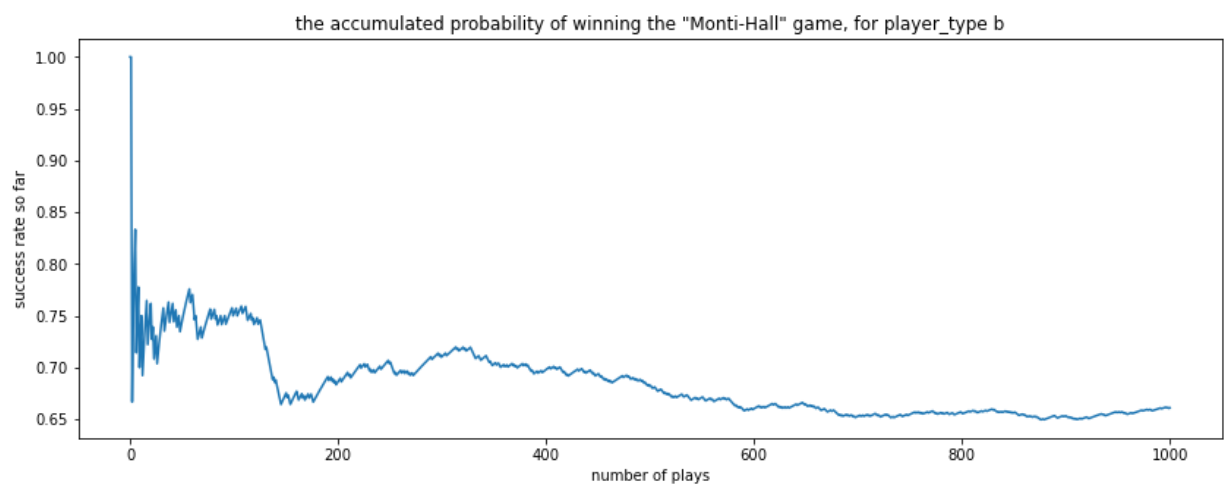
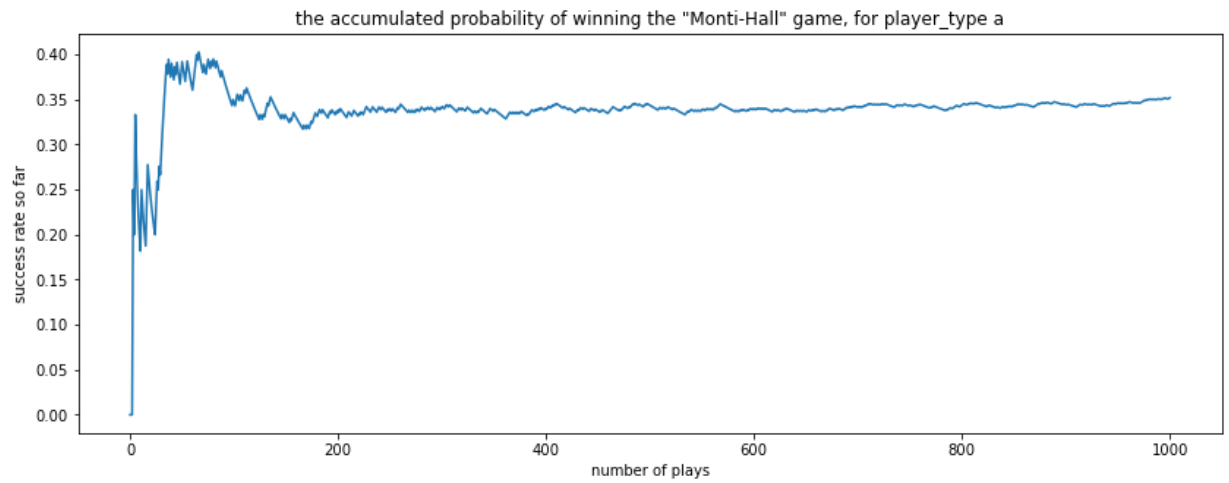
    return check_win(second_choose, car_door)
```

```
In [8]: def plot_accumulated_array(values, player_type):  
    plt.figure(figsize=(14,5))  
    plt.plot(np.linspace(0,1000,1000), values)  
    plt.xlabel('number of plays')  
    plt.ylabel('success rate so far')  
    plt.title(f'the accumulated probability of winning the "Monti-Hall" game, for  
    return
```

```
In [9]: def play_1000_times_and_plot(player_type):  
  
    ##playing 1000 times  
    results = []  
    for i in range(1000):  
        results.append(play(player_type))  
  
    ##accumulating the results  
    accumulated_results = []  
    for i in range(1000):  
        accumulated_results.append((sum(results[:i+1]))/(i+1))  
  
    ##plotting  
    plot_accumulated_array(accumulated_results, player_type)  
    return 1
```

```
In [10]: def Question2_players_a_b_c():  
    play_1000_times_and_plot('a')  
    play_1000_times_and_plot('b')  
    play_1000_times_and_plot('c')
```

In [11]: Question2\_players\_a\_b\_c()



**explanation about the graphs**

The player that had the highest success rate was player b (the one who always switch), It is predicted according to what we studied in class. Player a had an expected rate of 0.3 since it's simple probability (1 out of 3, no manipulations) Player c had combined success rate with a and b, the more times the rand number said 'change' the higher the success rate was.

This, of course, could have been predicted