

# AI Investigation: Using Binary Output Game with No Dominant Strategy to experiment using Confidence Intervals to determine inconclusive victories

## AI Investigation Part 2

*Goal: test using Confidence Intervals to determine inconclusive victories*

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```
In [ ]: import random
import math as m
```

### Define Game

Basically a game with no optimal strategy. The chance of winning is 50% regardless of what strategy is chosen.

(The game is similar to Rock Paper Scissors, except deliberately avoided the chance of draws)

```
In [ ]: CHOICE1 = ['A', 'B']
CHOICE2 = ['C', 'D']
```

```
In [ ]: def winloss(player1, player2):
        """ If player 1 plays A and player 2 plays C, then player 1 wins; if player 2 plays D
        and player 2 plays C, then player 1 loses, if player 2 plays D then player 1 wins """

        if player1 == 'A':
            if player2 == 'C':
                return True
            else:
                return False
        else:
            if player2 == 'D':
                return True
            else:
                return False
```

### Simulation

```
In [4]: RUNS = 1000000

sample = list()
victory = list()
for i in range(RUNS):
    obs1 = random.sample(CHOICE1, 1)[0]
    obs2 = random.sample(CHOICE2, 1)[0]
    sample.append(obs1)
    victory.append(winloss(obs1[0], obs2[0]))
```

**Manipulation** Sidenote: In future basic data collection should be done WITHIN Simulation to save time

```
In [5]: A_Score = 0
```

```

B_Score = 0

A_win = 0
B_win = 0

A_Count = 0
B_Count = 0

for i in range(RUNS):
    if sample[i] == 'A':
        A_Count += 1
        if victory[i]:
            A_Score += 1
            A_win += 1
        else:
            A_Score -= 1
    else:
        B_Count += 1
        if victory[i]:
            B_Score += 1
            B_win += 1
        else:
            B_Score -= 1

ScoreDict = ({'A': A_Score, 'B': B_Score})

```

Can get an estimate of p by:

$$E(X) = 2p-1 \Rightarrow p_{\text{hat}} = (x_{\text{bar}} + 1)/2$$

In [6]:

```

A_Mean = A_Score/A_Count
B_Mean = B_Score/B_Count

P_A = (A_Mean + 1)/2
P_B = (B_Mean + 1)/2

```

Constructing confidence interval using

$$\text{Mean} = 2p-1$$

$$\text{Var} = 4(p_{\text{hat}})(1-(p_{\text{hat}}))$$

(here we use p\_hat as an approximation for p)

In [7]:

```

A_99CI = ((A_Mean/2) + 0.5 + (-1) * 2.57 * m.sqrt(4*P_A*(1-P_A))/(2*(m.sqrt(A_Count))),
          (A_Mean/2) + 0.5 + (1) * 2.57 * m.sqrt(4*P_A*(1-P_A))/(2*(m.sqrt(A_Count))))

```

In [8]:

```

print(f'A_Score: {A_Score}')
print(f'B_Score: {B_Score}')
print(f'A_Count: {A_Count}')
print(f'B_Count: {B_Count}')
print(f'A_Mean: {A_Mean}')
print(f'B_Mean: {B_Mean}')
print(f'P_A: {P_A}')
print(f'P_B: {P_B}')
print(f'99% Confidence Interval: {A_99CI}')

```

```

A_Score: 111
B_Score: -55
A_Count: 50083
B_Count: 49917

```

```
A_Mean: 0.0022163209072938923
B_Mean: -0.001101829036200092
P_A: 0.5011081604536469
P_B: 0.4994490854819
99% Confidence Interval: (0.49536624368035564, 0.5068500772269382)
```

```
In [9]: print(f'P_A: {A_win/A_Count}')
        print(f'P_B: {B_win/B_Count}')
```

```
P_A: 0.5011081604536469
P_B: 0.4994490854819
```

## Algorithm

```
In [10]: def final_choice(ScoreDict, Scores, Counts):
        first = sorted(list(ScoreDict.items()), key = lambda x:x[1], reverse = True)[0][0]
        second = sorted(list(ScoreDict.items()), key = lambda x:x[1], reverse = True)[1][0]

        Means = {'A': Scores['A']/Counts['A'], 'B': Scores['B']/Counts['B']}

        Ps = {'A': (A_Mean + 1)/2, 'B': (B_Mean + 1)/2}

        # Construct 99% Confidence interval. If P of 'second' belongs in the 99% CI of first,
        # dominant strategy
        CI99 = ((Means[first]/2) + 0.5 + (-1) * 2.57 * m.sqrt(4*Ps[first]*(1-Ps[first]))/(2*(n
            (Means[first]/2) + 0.5 + (1) * 2.57 * m.sqrt(4*Ps[first]*(1-Ps[first]))/(2*(m.sc

        print(f'P_A: {Ps["A"]}')
        print(f'P_B: {Ps["B"]}')
        print(f'99% Confidence Interval: {CI99}')
```

```
        if Ps[second] > CI99[0] and Ps[second] < CI99[1]:
            return "No dominant strategy"

        return first
```

```
In [11]: Scores = {'A': A_Score, 'B': B_Score}
        Counts = {'A': A_Count, 'B': B_Count}
```

```
In [12]: final_choice(ScoreDict, Scores, Counts)
```

```
P_A: 0.5011081604536469
P_B: 0.4994490854819
99% Confidence Interval: (0.49536624368035564, 0.5068500772269382)
```

```
Out[12]: 'No dominant strategy'
```

## Final Note

Whilst this is a success, in reality changing the data into proportions is the same as using the mean to construct confidence intervals. In fact using the mean is more beneficial for games where the output are not binary (i.e. contain draws)