

ECON 220 Lab

(Week 4)

Probability and Functions in Python

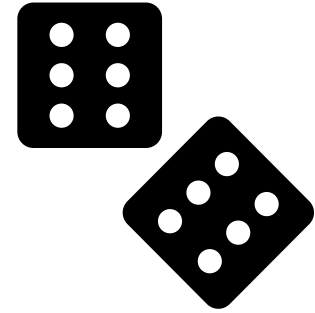
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Outline

- Empirical vs. Theoretical Probability
- Making Functions in Python

A Simple Function

```
def roll_die():  
    return random.randint(1, 6)  
  
# Example usage  
roll_die()
```



The function's name is **roll_die**. Running this will return us a **random** number, to be specific, a **random integer** between **1 to 6** (i.e. 1,2,3,4,5, and 6)

The Obvious

```
theoretical_prob = 1/6
```

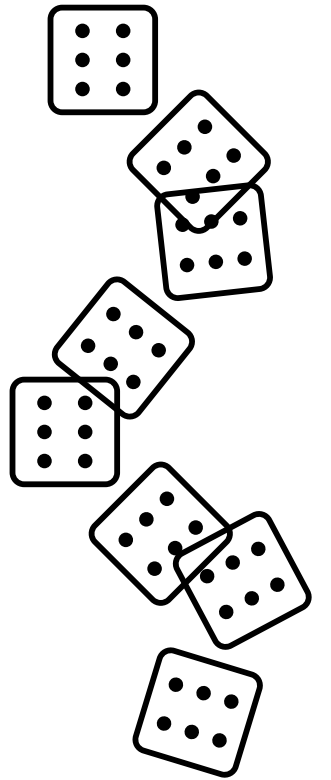
Assuming the die was fair, the theoretical probability will always be $1/6$.

That is, there is a one in six chance of rolling a specific number, say rolling a three.

A More Complicated Function

```
# Simulating multiple rolls
def simulate_rolls(n_trials, target_number):
    count = 0
    for _ in range(n_trials):
        roll = roll_die()
        print(f"The die fell on {roll}") # Debugging statement
        if roll == target_number:
            count += 1 # Correct counting
    return count / n_trials # Empirical probability
```

The function takes **two arguments** to run. (1) `n_trials` which is the number of experiments. (2) `target_number` which is the target result (i.e. any number from 1 to 6)



Running the Complicated Function

```
# Running the simulation
n_trials = 1000 # Adjust for testing
target_number = 3 # We're checking probability of rolling a 3

empirical_prob = simulate_rolls(n_trials, target_number)
```

It is good practice to declare an object that the function takes so it is easier to keep track of changes.

i.e. Running `simulate_rolls(1000,3)` will give you the same thing, but it's easier to change things when you make an object like this.

Empirical Probability vs Theoretical Probability

- Empirical Probability

- After n number of trials, the probability of rolling 3 is _____.
- This is going to vary by n !
 - The smaller the n , the less likely it is to be near the theoretical probability.
 - The larger the n , the more likely it is going to be near the theoretical probability (**LAW OF LARGE NUMBERS**)

- Theoretical Probability

- Assuming the die was **fair**, the probability of rolling 3 is _____.
- Clearly, this is $1/6$ regardless of n since there are six sides to the die, assuming fairness, are equally likely.

Illustrating the Law of Large Numbers

```
def convergence_rolls(n_trials, target_number):  
    empirical_probs = [] # Store empirical probabilities over time  
    count = 0 # Counter for occurrences of target_number  
  
    for i in range(1, n_trials + 1):  
        roll = roll_die()  
        if roll == target_number:  
            count += 1  
        empirical_probs.append(count / i) # Empirical probability at this trial  
  
    return empirical_probs # Ensure it returns a list of length n_trials
```

Similar to what we did before! Make a function. Then, we want to basically get empirical probabilities over the number of trials! We are going to get an average of the probability of rolling a specific number over the amount of trials!

Illustrating the Law of Large Numbers

```
# Running the simulation
n_trials = 1000 # Number of trials
target_number = 3 # Checking probability of rolling a 3

empirical_probs = convergence_rolls(n_trials, target_number)
```

```
plt.figure()
plt.plot(range(1, n_trials + 1), empirical_probs, label="Empirical Probability")
plt.axhline(y=theoretical_prob, color='r', linestyle='--', label="Theoretical Probability (1/6)")
plt.xlabel("Number of Rolls")
plt.ylabel("Probability of Rolling a 3")
plt.title(f"Empirical vs. Theoretical Probability of Rolling a {target_number}")
plt.legend()
plt.show()
```

Fancy Graph

Note: The result WILL VARY from person to person. Why?

Because... we used **random** remember? 😊

Clearly, as the number of rolls gets larger and larger, we see that the empirical probability converges to the theoretical probability

