#### #17: Functions in Python and Conditional Probability

March 4th, 2020

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#### **Control Flow II: Functions!**

Up until today, we have explored just one technique to control the flow of code: a **for-loop**.

Here's the for-loop syntax in a simple form:

```
print("Three dice rolls:")
for i in range(3):
   value = random.randint(1, 6)
   print("You rolled a: " + value)
print("Bye!")
```

The second control flow statement we'll explore is a \_\_\_\_\_:

```
def rollDie():
   value = random.randint(1, 6)
   return value
```

Three key ideas when using **functions**:

- 1. [Syntax]:
- 2. [Indentation]:
- 3. [Control Flow]:
- 4. [Return Value]:

Puzzle #1: After running the previous code in this handout, what does this cell do?

```
1 rollDie()
```

Puzzle #2: After running the previous code in this handout, what does this cell do?

```
1 result = rollDie()
2 result
```

Puzzle #3: After running the previous code in this handout, what does this cell do?

```
1 result = rollDie() + rollDie() + rollDie()
2 result
```

```
Puzzle #6:

1 rollDie(25)

Puzzle #7:

1 rollDie(1000000)
```

## **Using Functions to Simplify Complex Problems**

We are building complex problems! Suppose we don't want to remember the probability of drawing a queen, we can define P queen():

```
1 def P_queen():
2 return (1/13)
```

What about drawing multiple queens in a row with replacement:

```
1 def P_queen( ):
```

What about returning a simulation of events?

```
1
  def simulate_rolling_dice( times, dieSides ):
2
    data = []
    for i in range( times ):
3
       roll = random.randint(1, dieSides)
4
       d = { "roll": roll }
5
        data.append(d)
6
7
    df = pd.DataFrame( data )
8
    return df
```

# **Conditional Probability and Tables**

**Definition:** Conditional Probability: The *conditional probability* of an event *B* is the probability that the event will occur given that an event *A* has already occurred.

**Conditional Probability**: 
$$P(B \mid A) = \frac{P(A \text{ and } B)}{P(A)} = \frac{P(A \cap B)}{P(A)} = \frac{P(A \cap B)}{P(A)}$$

## $\star$ If a situation involves 2 or more conditions, a table is helpful. $\star$

**Example 1:** Suppose an old box of pills contains 30 Ibuprofen and 70 Tylenol. Half the ibuprofen and 10% of the Tylenol are expired. Fill in the table below:

	Expired	Not Expired	Totals
Ibuprofen			
Tylenol			
Totals			

- a) What's the probability of randomly drawing a pill from the box and getting one that's expired?
- **b)** What's the probability of randomly drawing a pill from the box and getting an expired Ibuprofen?
- c) In all contingency tables, the numbers in the margins are called *marginal* probabilities (when divided by the overall total) & the numbers in the inside cells are called *joint* probabilities. Your answer to \_\_\_\_\_ above is a marginal probability and your answer to \_\_\_\_\_ above is a joint probability because it gives the probability of getting both a \_\_\_\_\_ & a \_\_\_\_\_. (Fill in the blanks, with either a or b.)
- **d)** Show that P(A|B) isn't the same as P(B|A). Can you think of some obvious examples that would convince others without having to show them by doing calculations?

<sup>\*</sup> This expression is only valid when P(A) is not equal to o.

<sup>\*</sup> In the case where events *A* and *B* are *independent* (when event *A* has no effect on the probability of event *B*), the conditional probability of event *B* given *A* is just the probability of event *B*: *P*(*B*).

**Example 2:** Suppose jurors make the right decisions about guilt and innocence 95% of the time and that 80% of all defendants are truly guilty.

	Acquitted	Convicted	Totals
Innocent			
Guilty			
Totals			100

- a) What's the probability that a person is convicted given that they are innocent?
- **b)** What's the probability that a person is innocent, given that they are convicted?

**Example 3:** For years ELISA has been the most widely used test for screening donated blood for AIDS. (See http://www.stat.ucla.edu/cases/elisa/)

If a blood sample has the AIDS virus, then there is a 99% probability that ELISA will correctly give a positive result. If the blood sample does not have the AIDS virus then there is still a 6% probability that ELISA will incorrectly give a positive result. About 1% of the blood samples ELISA tests are truly contaminated.

	Result of Test		
	Positive	Negative	Totals
Blood has AIDS virus			
Blood does NOT have AIDS virus			
Totals			10,000

- a) What's the probability someone tested positive given that the person has the AIDS virus?
- **b)** What's the probability someone has the AIDS virus given the person tested positive?