# Introduction a first course in Programming to Data Analysis Using python.

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# **Contingency table**

		Diagnose	
		Disease	No-Disease
Risk Factor	Smoke	а	b
	Not Smoke	С	d

What it is  $P(Disase \mid Smoke) = \frac{a}{a+b}$  note that marginal distribution. Note that  $P(Disease \cap Smoke) = \frac{a}{(a+b+c+d)}$ . Also note that

 $P(Smoke) = \frac{a+b}{(a+b+c+d)}$ . Note the result of divide the last two probabilities.



# **Conditional Probability**

The probability of event given a "information". Probability of A occur given B occurs.

$$P(A \mid B) = \frac{P(A \cap B)}{P(B)} \tag{1}$$

Note that  $P(A \cap B)$  it is equal to  $P(B \cap A)$ .

$$P(A \mid B)P(B) = P(B \mid A)P(A)$$
 (2)



# **Bayes theorem**

Notice that not is same: *The probability that occur A given B, that occur B given A.* However we can compute one of the another.

$$P(A \mid B) = \frac{P(A)P(B \mid A)}{P(B)} \tag{3}$$



# **Bowls problems**

Derive the following problem.



# Law of total probability

The sample space defined as  $\Omega$  if we split omega in  $\omega$  k subsets in order that each *subset* no overlap with others.  $\bigcup_{i=1}^k \omega_i = \Omega$  y  $\bigcap_{i=1}^k \omega_i = \emptyset$  For instance the sample space defined as  $\Omega = \{a, b, c, d, e, f\}$   $\omega_1 = \{a, f\}$   $\omega_2 = \{b, c, d\}$   $\omega_3 = \{e\}$ .



# $\textbf{Split} \ \Omega$

Ω  $\omega_{12}$  $\omega_{10}$  $\omega_3$  $\omega_5$  $\omega_{11}$  $\omega_{4}$  $\omega_2$  $\omega_7$  $\omega_9$  $\omega_6$  $\omega_8$  $\omega_1$ Pontificia Universidad JAVERIANA

# P(A)

#### total law probability

We need remember by set theory that a event A could be rewrite as  $A = (A \cap B) \cup (A \cap C)$ . if  $(B \cup C) = \Omega$  for this case we can rewrite

$$A = (A \cap \omega_1) \cup (A \cap \omega_2)...(A \cap \omega_k)$$

$$P(A) = P(A \cap \omega_1) + ... + P(A \cap \omega_k)$$

$$P(A) = P(A \mid \omega_1)P(\omega_1) + P(A \mid \omega_2)P(\omega_2) + ... + P(A \mid \omega_k)P(\omega_k)$$
(4)

note that by total law P(A)



## Monty hall

Reach the famous in 1990

There are three closed doors, and you must select one to win a car, behind of the only one, there is a car, and behind the other two there are goats. and after select the door, another door with a goat it is showed, you must remain in the selected door or switch? initial probality it is the 1/3.

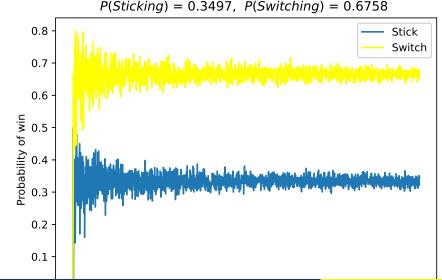


# **Monty Hall simulation**

This so not intuitive that generated controversy in the academic community.



The following Figure 16 show the relation among the probability of two coincidences among two person in the sample.



# Monty Hall problem

Python code

```
def monty_game():
 doors=[1,2,3]
 doors_variable = doors.copy()
 winner = random.randint(1,3)
 select_one = random.randint(1,3)
 values = [winner,select_one]
 switch = list( set(doors) - set(values))
 select_two = random.randint(switch[0], switch[-1])
 s1,s2 = 0,0
 if winner == select_one:
   s1 += 1
 else:
   s2 +=1
 return [s1,s2]
```



# Monty Hall problem

python code

```
games=1000
s1 total = ∏
s2 total = []
for trials in range(1,games):
 values_s1 = sum([monty_game()[0] for _ in range(trials)]) /
     trials
 s1_total.append(values_s1)
 values_s2 = sum(monty_game()[1] for _ in range(trials)) /
     trials
 s2_total.append(values_s2)
plt.plot(np.arange(1,games),s1_total)
plt.plot(np.arange(1,games),s2_total)
print(s1_total[-1], s2_total[-1])
```

# Monty Hall

#### **Bayesian solution**

The event  $D_i$  the i winner door and  $M_i$  monty open j door, for i, i = 1, 2, 3.

$$P(D_i \mid M_j) = \frac{P(M_j \mid D_i)P(D_i)}{P(M_j)}$$
 (5)

you select the first door and monty the second, therefore the question is  $P(D_3 \mid M_2)$ . Notice that  $P(M_2) = \sum_{i=1}^3 P(M_2 \mid D_i)$ . given the rules of games,  $P(M_2 \mid D_2) = 0$ , and  $P(M_2 \mid D_3) = 1$ , if monty could select random in two choices  $P(M_2 \mid C_1) = 1/2$  and finally, switch strategy have a probability of 2/3.

### **Buffon Needled**

Here we have a column

we have another column

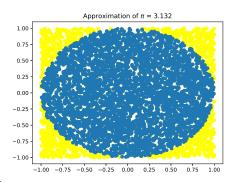


#### $\pi$ Number

```
import numpy as np
import matplotlib.pyplot as plt
dots = 5000
c1.c2 = -1.1
x = np.random.uniform(c1, c2,
    size=dots)
y = np.random.uniform(c1, c2,
    size=dots)
coordenates_circle =
    (x**2)+(y**2) < 1
circle_y=y[coordenates_circle]
circle_x=x[coordenates_circle]
pi = 4*sum(coordenates_circle)
    / dots
plt.scatter(x,y, color='yellow')
plt.scatter(circle_x,circle_y)
```

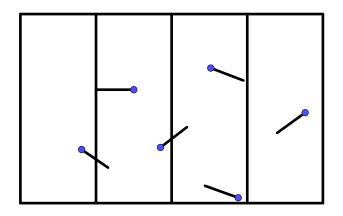
What is the probability of a drop lands in the circle?

$$P(hit) = \frac{\pi r^2}{4r^2} \tag{6}$$



## **Needles**

Proposed by and resolved by the naturalist. take in mind the





## Uniform distribution

 $\times \sim U(a,b)$  in the interval(a,b).

$$f(x) = \frac{1}{b-a} \tag{7}$$

the function is defined in the open interval a < x < b. Remember that:

$$F(x) = \int_{-\infty}^{x} f(u) du$$
 (8)

import numpy as np np.random.uniform(a,b ,size=(k,p)) # [)

#Draw k list with p elements with numbers [a,b)

Choose a point in the interval (a,b),  $F(x) = \int_{-\infty}^{x} f(u)du$  (8) You can calculate what it is the probability that a point is in (c,d)