

1. (a) True, since A and B are independent, $P(A \cap B) = P(A) \cdot P(B)$

$$\begin{aligned} \text{Also, } P(B) &= P(B \cap A) + P(B \cap A^c) \\ &= P(B) \cdot P(A) + P(B \cap A^c) \\ \Rightarrow P(B)(1 - P(A)) &= P(B \cap A^c) \\ \Rightarrow P(B) \cdot P(A^c) &= P(B \cap A^c) \end{aligned}$$

$\therefore A^c$ and B are independent

(b) False, Suppose you flip a fair coin twice.

let A: the event that head occurs for first coin flip, (1H)

B: the event that tail occurs for first coin flip, or
head occurs for second coin flip (1T or 2H)

C: the event that head occurs for second coin flip (2H)

$$\text{First, } P(A|B,C) = \frac{P(A \cap B \cap C)}{P(B \cap C)} = \frac{(1H \text{ and } 2H)}{(2H)} = \frac{\frac{1}{4}}{\frac{1}{2}} = \left(\frac{1}{2}\right)$$

$$P(A|C) = \frac{P(A \cap C)}{P(C)} = \frac{(1H \text{ and } 2H)}{(2H)} = \frac{\frac{1}{4}}{\frac{1}{2}} = \left(\frac{1}{2}\right)$$

so, A and B are conditionally independent given C.

$$\text{Next, } P(A|B, C^c) = \frac{P(A \cap B \cap C^c)}{P(B \cap C^c)} = \frac{(1H \text{ and } (1T \text{ or } 2T) \text{ and } (\neq 2H))}{(1T \text{ or } 2H) \text{ and } (\neq 2H)} = \emptyset$$

$$\because \text{not } A \cap B = 1H \text{ and } 2H, \text{ but cannot add } (\neq 2H) \quad C^c = \emptyset$$

$$P(A|C^c) = \frac{P(A \cap C^c)}{P(C^c)} = \frac{(H \text{ and } (\neq 2H))}{(2T)} = \frac{1H \text{ and } 2T}{2T} = \frac{\frac{1}{4}}{\frac{1}{2}} = \left(\frac{1}{2}\right)$$

Since $P(A|B, C^c) \neq P(A|C^c)$,

A and B are not conditionally independent given C^c .

c) True, Any Events (i.e. A and B) in a partition are disjoint.

$$\text{So, } P(A \cap B) = 0.$$

If A and B are independent, $P(A \cap B) = P(A) \cdot P(B)$,

Since $P(A) \neq 0$, $P(B) \neq 0$ but $P(A \cap B) = 0$ so equation is incorrect.

Therefore, events in a partition cannot be independent.

d) True, Since $P(A|B) = 1$, $\frac{P(A \cap B)}{P(B)} = 1$, $P(A \cap B) = P(B)$

$$\begin{aligned} P(B^c|A^c) &= \frac{P(B^c \cap A^c)}{P(A^c)} = \frac{P((A \cup B)^c)}{P(A^c)} \quad (\because \text{by de Morgan's law}) \\ &= \frac{1 - P(A \cup B)}{1 - P(A)} = \frac{1 - (P(A) + P(B) - P(A \cap B))}{1 - P(A)} \\ &= \frac{1 - (P(A) + P(B) - P(A \cap B))}{1 - P(A)} = \frac{1 - P(A)}{1 - P(A)} \quad (1) \end{aligned}$$

e) True, $P(B|A \cup B) = P(B \cap A|A \cup B) + P(B \cap A^c|A \cup B)$

$$= \frac{P(B \cap A \cap (A \cup B))}{P(A \cup B)} + \frac{P(B \cap A^c \cap (A \cup B))}{P(A \cup B)}$$

(Since, $P(B|A) = P(B|(A \cup B) \cap A)$, change formula as below)

$$= \frac{P((B \cap A) \cap (A \cup B))}{P((A \cup B) \cap A)} \cdot \frac{P((A \cup B) \cap A)}{P(A \cup B)} + \frac{P((B \cap A^c) \cap (A \cup B))}{P((A \cup B) \cap A^c)} \cdot \frac{P((A \cup B) \cap A^c)}{P(A \cup B)}$$

$$= P(B|(A \cup B) \cap A) \cdot P(A|A \cup B) + P(B|(A \cup B) \cap A^c) \cdot P(A^c|A \cup B)$$

(Since $(A \cup B) \cap A = A$, $(A \cup B) \cap A^c = B \cap A^c$)

$$= P(B|A) \cdot P(A|A \cup B) + P(B|B \cap A^c) \cdot P(A^c|A \cup B)$$

$$\text{Also, } P(B|B \cap A^c) = \frac{P(B \cap B \cap A^c)}{P(B \cap A^c)} = 1$$

$$= P(B|A) \cdot P(A|A \cup B) + P(A^c|A \cup B)$$

$$\geq P(B|A) \cdot P(A|A \cup B) + P(B|A) \cdot P(A^c|A \cup B) \quad (\because P(B|A) \leq 1)$$

$$= P(B|A) (P(A|A \cup B) + P(A^c|A \cup B)) = P(B|A)$$

2. (a) For F_A to be a valid σ -algebra, we need to prove below

i) $\bigcup_{A \in F_A}$, set of $A \cap F$ should be partition of every A .
and contains total A .

$$\therefore A = \bigcup_{A \in F_A} A \cap F.$$

ii) If $B \in F_A$, then $B^c \in F_A$,

If sample space is A , then $B^c = A - B$.

If $B \in F_A$, there is some set $S \in F$ such that $B = A \cap S$

Since F is a σ -algebra, $S^c \in F$.

$$A = (A \cap S) + (A \cap S^c) = B + (A \cap S^c)$$

$$\Rightarrow A - B = A \cap S^c.$$

$$\therefore B^c = A - B = A \cap S^c \in F_A$$

iii) if $B_1, B_2 \in F_A$, then $B_1 \cup B_2 \in F_A$. Also, if $B_1, B_2, \dots \in F_A$, then $\bigcup_{i=1}^{\infty} B_i \in F_A$

If $B_1, B_2 \in F_A$, then there exist $S_1, S_2 \in F$ such that $B_1 = A \cap S_1$, $B_2 = A \cap S_2$

Since F is a σ -algebra, $S_1 \cup S_2 \in F$, so $A \cap (S_1 \cup S_2)$ is in F_A

$$\therefore A \cap (S_1 \cup S_2) = (A \cap S_1) \cup (A \cap S_2) = B_1 \cup B_2 \in F_A$$

(*) By the same argument for the finite case,

$$\text{if } B_1, B_2, \dots \in F_A, \text{ then } \bigcup_{i=1}^{\infty} B_i \in F_A.$$

\rightarrow Let $B = A \cap S$, then $P_A(S \cap A) = P_A(B) = \frac{P(B)}{P(A)}$

i) $P_A(B) \geq 0$ for any event $B \in F_A$. ($\because P(B) \geq 0, P(A) > 0$)

ii) $P_A(\text{sample space}) = 1$, $P_A(A) = \frac{P(A)}{P(A)} = 1$

iii) If $B_1, B_2, \dots, B_n \in F_A$ are disjoint, then $P\left(\bigcup_{i=1}^n B_i\right) = \sum_{i=1}^n P(B_i)$

$$\text{i.e. } P(B_1 \cup B_2) = P(B_1) + P(B_2) - P(B_1 \cap B_2) = P(B_1) + P(B_2)$$

$$P_A\left(\bigcup_{i=1}^n B_i\right) = \frac{P\left(\bigcup_{i=1}^n B_i\right)}{P(A)} = \frac{\sum_{i=1}^n P(B_i)}{P(A)} = \sum_{i=1}^n P_A(B_i).$$

By the same argument for the finite case,

for a countably infinite disjoint sets $B_1, B_2, \dots \in F_A$,

$$P\left(\lim_{n \rightarrow \infty} \bigcup_{i=1}^n B_i\right) = \lim_{n \rightarrow \infty} \sum_{i=1}^n P(B_i).$$

2-(b) Yes, P should satisfy the conditions below

i) $P(B) \geq 0$ for any event $B \in F$. (⊗)

(\because numerator and denominator are both non-negative)

ii) $P(\text{sample space}) = 1$, $P(\emptyset) = \frac{\# \text{ of data points with value in } \emptyset}{N} = 1$

iii) If $S_1, S_2, \dots, S_n \in F$ are disjoint, then $P\left(\bigcup_{i=1}^n S_i\right) = \sum_{i=1}^n P(S_i)$

$$P\left(\bigcup_{i=1}^n S_i\right) = \frac{\# \text{ of data points with value } \bigcup_{i=1}^n S_i}{N}$$

$$= \frac{\# \text{ of data points with value in } S_1 + \# \text{ of } \dots S_2 + \dots + \# \text{ of } \dots S_n}{N}$$

$$= \sum_{i=1}^n P(S_i)$$

By the same argument for the finite case,
for a countably infinite disjoint sets $S_1, S_2, \dots \in F$,

$$P\left(\lim_{n \rightarrow \infty} \bigcup_{i=1}^n S_i\right) = \lim_{n \rightarrow \infty} \sum_{i=1}^n P(S_i)$$

3. (a) Yes, since employees have not been in contact with each other, it is reasonable to assume that Test i only depends on particular employee i .

(b) Define $I_i (1 \leq i \leq 10)$ = each employee i being ill
 $T_i (1 \leq i \leq 10)$ = each employee i test is positive

The event that at least one test is positive is $\left(\bigcup_{i=1}^{10} T_i \right)$.

$$P\left(\bigcup_{i=1}^{10} T_i\right) = 1 - P\left(\left(\bigcup_{i=1}^{10} T_i\right)^c\right) = 1 - P\left(\bigcap_{i=1}^{10} T_i^c\right) \text{ (by DeMorgan's laws)}$$

~~State~~ $P\left(\bigcap_{i=1}^{10} T_i^c\right) = \prod_{i=1}^{10} P(T_i^c) \quad (\because T_i \text{ is independent})$

$$= \prod_{i=1}^{10} \left(P(I_i) P(T_i^c | I_i) + P(I_i^c) P(T_i^c | I_i^c) \right)$$

$$\begin{cases} P(I_i) = 0.01 \\ P(I_i^c) = 0.99 \end{cases}, \quad \begin{cases} P(T_i | I_i) = 0.98 \\ P(T_i^c | I_i) = 0.02 \end{cases}, \quad \begin{cases} P(T_i | I_i^c) = 0.05 \\ P(T_i^c | I_i^c) = 0.95 \end{cases}$$

$$= (0.01)(0.02) + (0.99)(0.95)^{10}$$

$$\approx 0.543$$

$$\boxed{\therefore P\left(\bigcup_{i=1}^{10} T_i\right) \approx 1 - 0.543 \approx 0.457}$$

$$(c) P\left(\bigwedge_{i=1}^{10} I_i^c \mid \bigvee_{j=1}^{10} T_j\right) = \frac{P\left(\bigwedge_{i=1}^{10} I_i^c, \bigvee_{j=1}^{10} T_j\right)}{P\left(\bigvee_{j=1}^{10} T_j\right)} = 0.457 \quad (3-b)$$

$$P\left(\bigwedge_{i=1}^{10} I_i^c, \bigvee_{j=1}^{10} T_j\right) = P\left(\bigwedge_{i=1}^{10} I_i^c\right) \cdot P\left(\bigvee_{j=1}^{10} T_j \mid \bigwedge_{k=1}^{10} I_k^c\right)$$

$$= \prod_{i=1}^{10} P(I_i^c) \times \left(1 - P\left(\bigvee_{j=1}^{10} T_j^c \mid \bigwedge_{k=1}^{10} I_k^c\right)\right)$$

~~Assumption) T_j^c is conditionally independent of I_i^c , $i \neq j$, conditioned on $\bigwedge_{k=1}^{10} I_k^c$~~
~~Since T_j^c doesn't provide information about I_i^c .~~

$$P\left(\bigvee_{j=1}^{10} T_j^c \mid \bigwedge_{k=1}^{10} I_k^c\right) = \prod_{j=1}^{10} P(T_j^c \mid I_j^c)$$

~~Assumption2) T_i is conditionally independent of I_j^c ($i \neq j$), conditioned on I_i~~

~~Since T_i only depends on I_i .~~

$$\prod_{j=1}^{10} P(T_j^c \mid I_j^c) = \prod_{j=1}^{10} P(T_j^c \mid I_j)$$

$$\therefore P\left(\bigwedge_{i=1}^{10} I_i^c, \bigvee_{j=1}^{10} T_j\right) = \frac{\prod_{i=1}^{10} P(I_i^c) \times \left(1 - \prod_{j=1}^{10} P(T_j^c \mid I_j)\right)}{P\left(\bigvee_{j=1}^{10} T_j\right)}$$

$$= \frac{(0.99)^{10} \times \left(1 - (0.45)^{10}\right)}{0.457} \doteq [0.793]$$

4. (a) Total count: $2^5 = 32$.

(i) H:1	(ii) H:2	(iii) H:3
X=1) HTTTT	HTHTT	HTHHT
HTHTT	HTTHH	
TTHTT		
TTHTT	THTHT	
TTTHT	THTTH	
TTTHH	TTHTH	
	TTHTH	
5	6	1

$$\left(\frac{12}{32}\right) \doteq 0.375$$

(i) H:2	(ii) H:3	(iii) H:4
HHHTT	HHHTH	HHTHH
THHHT	HHHTH	
TTHTT	THHTH	
TTTHH	HTHTH	
	HTHTH	
4	5	1

$$\left(\frac{11}{32}\right) \doteq 0.344$$

(i) H:3	(ii) H:4
HHHTT	HHHTH
THHHT	HTHTH
TTTHH	
3	2

$$\left(\frac{5}{32}\right) \doteq 0.156$$

(i) H:4
HHHHT
THHHH
2

$$\left(\frac{2}{32}\right) \doteq 0.063$$

(i) H:5
1

$$\left(\frac{1}{32}\right) = 0.031$$

(i) T:5
1

```
In [15]: # 4-b) probability is similar to 4-a
print(p_longest_streak(5, 100000)[1:])

[0.37544, 0.34286, 0.15678, 0.06302, 0.03089]
```

```
In [3]: # 4-b) code
import numpy as np
import matplotlib.pyplot as plt
plt.close("all")
np.random.seed(2017)

def p_longest_streak(n, tries):
    # initialize list
    result = [0] * (n+1)

    # iterate # of tries
    for _ in range(tries):
        # initialize continue_count and maxCount
        continue_cnt = 0
        maxCount = 0

        # iterate flip by n times
        for i in range(n):
            # create random coin
            coin = np.random.random()

            # if coin >= 0.5, then add to continuous count
            if coin >= 0.5:
                continue_cnt += 1

            # if not, cnt is not continuous so cnt = 0
            else:
                continue_cnt = 0

            # put into maxCount before next iteration
            maxCount = max(maxCount, continue_cnt)

        # add values into result list
        result[maxCount] += 1

    # divide by total probability to get probability
    result_prob = [result[i] / tries for i in range(len(result))]

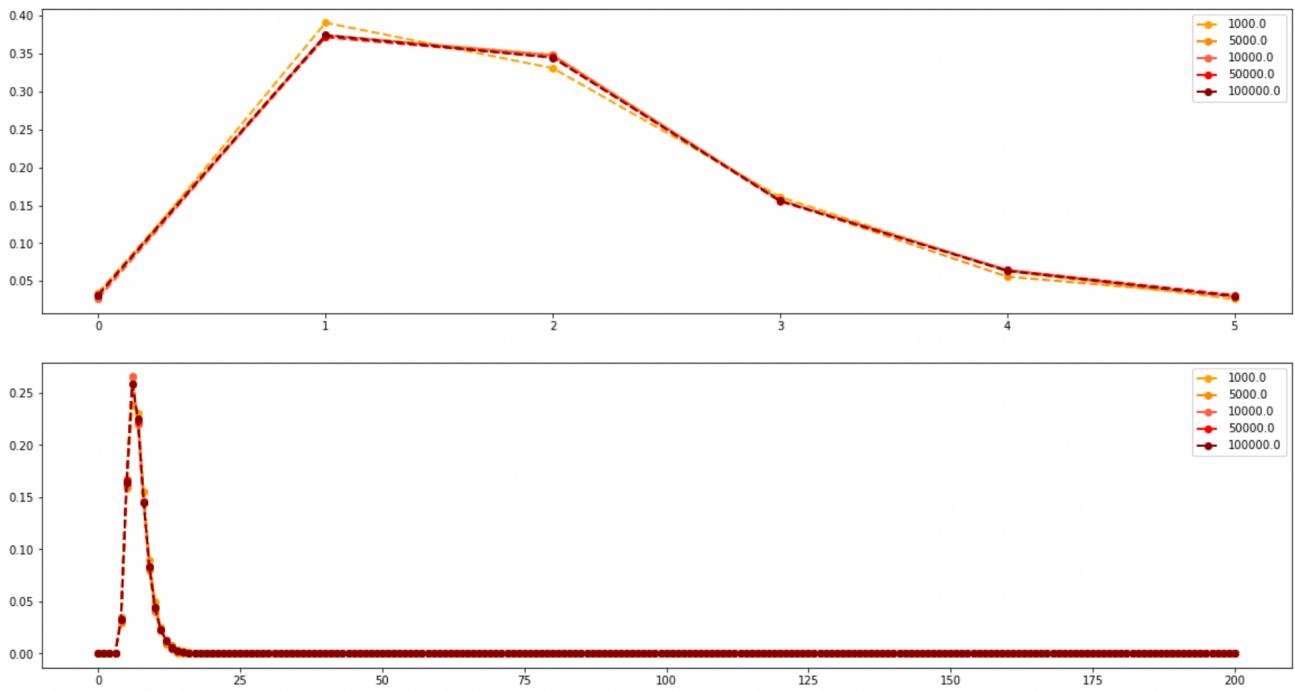
    # return list of probability
    return result_prob

n_tries = [1e3, 5e3, 1e4, 5e4, 1e5]
n_vals = [5, 200]

color_array = ['orange', 'darkorange', 'tomato', 'red', 'darkred', 'tomato', 'purple', 'grey', 'deepskyblue',
               'maroon', 'darkgray', 'darkorange', 'steelblue', 'forestgreen', 'silver']
for ind_n in range(len(n_vals)):
    n = n_vals[ind_n]
    plt.figure(figsize=(20,5))
    for ind_tries in range(len(n_tries)):
        tries = n_tries[ind_tries]
        print ("tries: " + str(tries))
        p_longest_tries = p_longest_streak(n, int(tries))
        plt.plot(range(n+1), p_longest_tries, marker='o', markersize=6, linestyle="dashed", lw=2,
                  color=color_array[ind_tries],
                  markeredgecolor= color_array[ind_tries], label=str(tries))

    plt.legend()
```

tries: 1000.0
tries: 5000.0
tries: 10000.0
tries: 50000.0
tries: 100000.0
tries: 1000.0
tries: 5000.0
tries: 10000.0
tries: 50000.0
tries: 100000.0



In [4]: # 4-c) about 0.32

```
# The program may not be generating truly random sequences,
# because it may follow the same and repeated algorithm to generate random numbers.
# (starting from the same seed)
# So the numbers it produces may not be truly random.
```

```
print ("The probability that the longest streak of ones in a Bernoulli iid sequence of length 200 has length 8 or more is")
print (np.sum(p_longest_streak(200, 100000)[8:])) # Compute the probability and print it here
```

The probability that the longest streak of ones in a Bernoulli iid sequence of length 200 has length 8 or more is
0.3200300000000004