

05: Independence

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October 2, 2019

Two Dice

- Roll two 6-sided dice, yielding values D_1 and D_2 .
- Let event E : $D_1 = 5$
event F : $D_2 = 5$








1. Roll a 5 on one of the rolls
 2. Roll a 5 on both rolls
 3. Neither roll is 5
 4. Roll a 5 on roll 2
 5. Do not roll a 5 on one of the rolls
- A. $P(F)$
 - B. $P(E \cup F)$
 - C. $P(E^C \cup F^C)$
 - D. $P(EF)$
 - E. $P(E^C F^C)$



Two Dice

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- | | | |
|--|---|----------------------|
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| 3. Neither roll is 5 |  | C. $P(E^C \cup F^C)$ |
| 4. Roll a 5 on roll 2 |  | D. $P(EF)$ |
| 5. Do not roll a 5 on one of the rolls |  | E. $P(E^C F^C)$ |



Monty Hall, 1000 envelope version

Start with 1000 envelopes
(of which 1 is the prize).

1. You choose 1 envelope.

$$\frac{1}{1000} = P(\text{envelope is prize})$$

$$\frac{999}{1000} = P(\text{other 999 envelopes have prize})$$

2. I open 998 of remaining 999
(showing they are empty).

$$\frac{999}{1000} = P(998 \text{ empty envelopes had prize}) \\ + P(\text{last other envelope has prize})$$

$$= P(\text{last other envelope has prize})$$

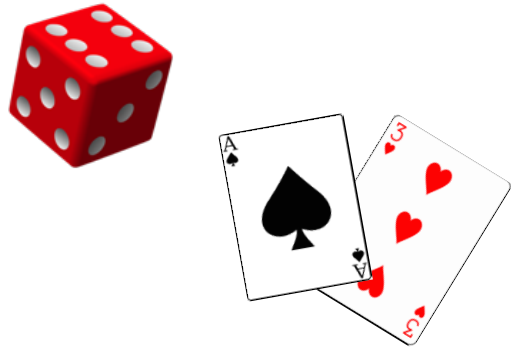
3. Should you switch?

$$P(\text{you win without switching}) = \frac{1}{\text{original \# envelopes}}$$

$$P(\text{you win with switching}) = \frac{\text{original \# envelopes} - 1}{\text{original \# envelopes}}$$

This class going forward

Last week
Equally likely
events

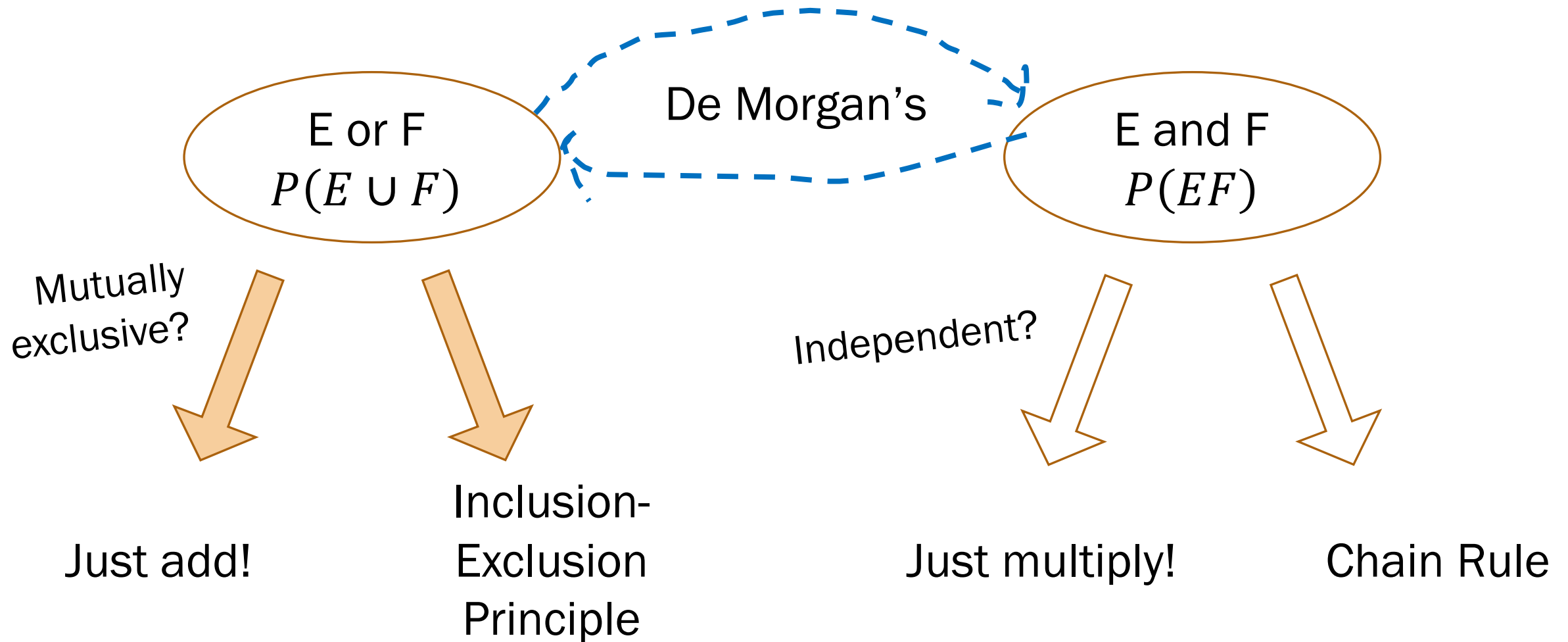


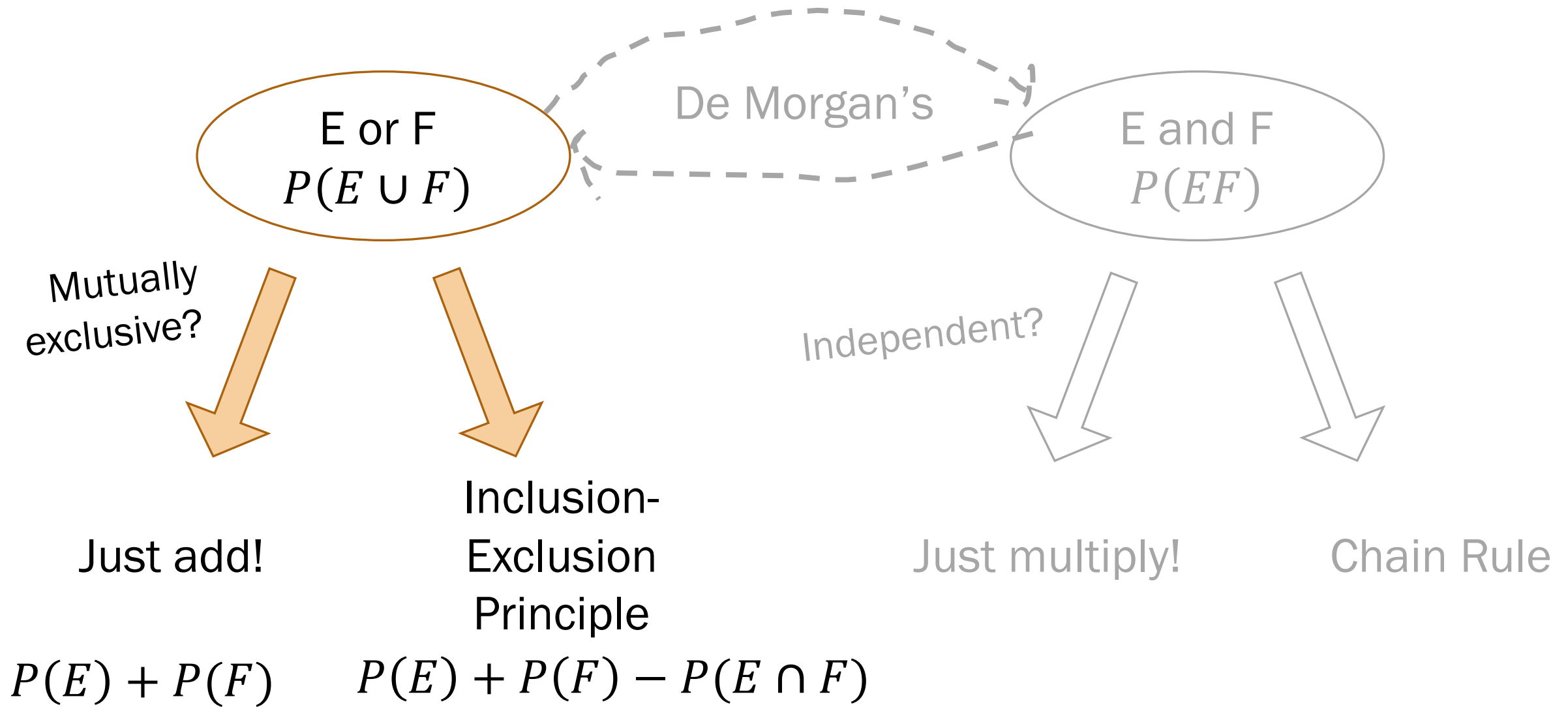
$P(E \cap F)$ $P(E \cup F)$
(counting, combinatorics)

For most of this course
Not equally likely events

$P(E \text{ given some evidence has been observed})$

Probability of events





- $P(\text{student programs in Java}) = 0.28$
- $P(\text{student programs in Python}) = 0.07$
- $P(\text{student programs in Java and Python}) = 0.05.$

What is $P(\text{student does not program in (Java or Python)})$?

1. Define events
& state goal

2. Identify known
probabilities

3. Solve

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3. Solve

Let: E : Student programs
in Java

F : Student programs
in Python

Want: $P((E \cup F)^c)$

Selecting Programmers

- $P(\text{student programs in Java}) = 0.28$ $P(E)$
- $P(\text{student programs in Python}) = 0.07$ $P(F)$
- $P(\text{student programs in Java and Python}) = 0.05.$ $P(E \cap F) = P(EF)$

What is $P(\text{student does not program in (Java or Python)})$?

1. Define events
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Selecting Programmers

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What is $P(\text{student does not program in (Java or Python)})$?

1. Define events
& state goal

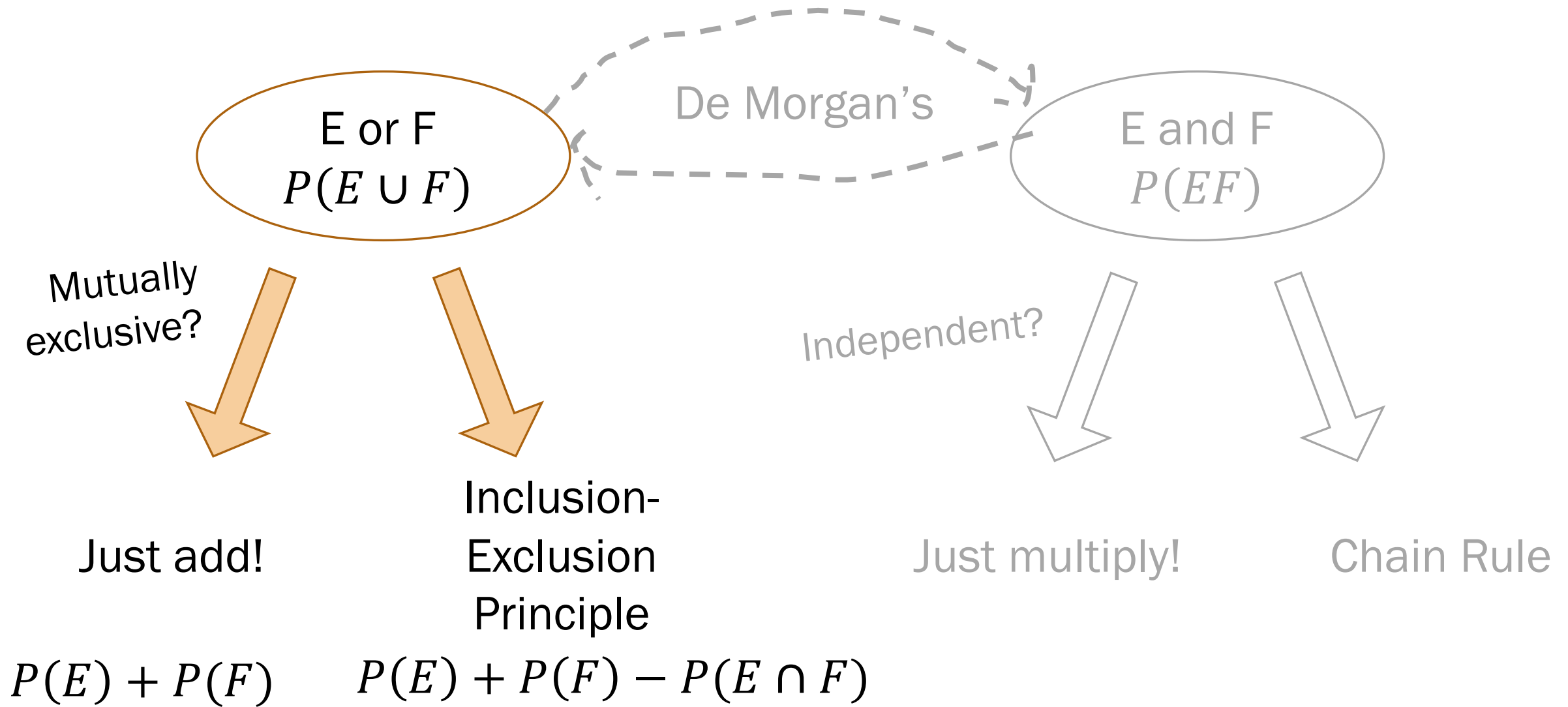
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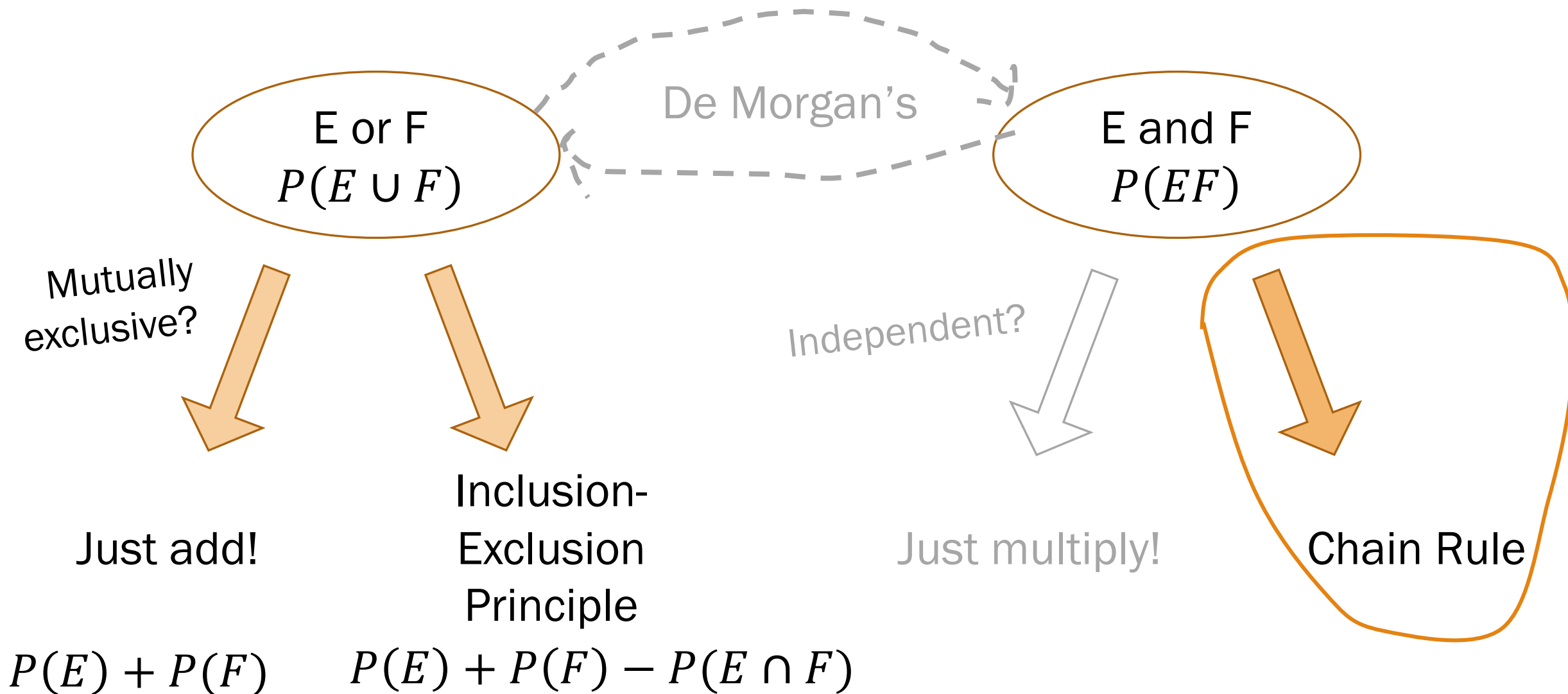
Let: E : Student programs
in Java
 F : Student programs
in Python

Want: $P((E \cup F)^c)$

$$\begin{aligned} P((E \cup F)^c) &= 1 - P(E \cup F) \\ &= 1 - [P(E) + P(F) - P(E \cap F)] \\ &= 1 - [0.28 + 0.07 - 0.05] \\ &= 0.70 \end{aligned}$$



Probability of events



Definition of conditional probability:

$$P(E|F) = \frac{P(EF)}{P(F)}$$

The Chain Rule:

$$P(EF) = P(E|F)P(F)$$

Generalized Chain Rule

$$P(E_1 E_2 E_3 \dots E_n) \\ = P(E_1) P(E_2 | E_1) P(E_3 | E_1 E_2) \dots P(E_n | E_1 E_2 \dots E_{n-1})$$



Quick check

$$P(E_1 E_2 E_3 \dots E_n) = P(E_1) P(E_2 | E_1) \dots P(E_n | E_1 E_2 \dots E_{n-1})$$
 Chain Rule

You are going to a friend's Halloween party.

Let C = there is candy
 M = there is music

W = you wear a costume
 E = no one wears your costume

An awesome party means that all of these events must occur.

What is $P(\text{awesome party}) = P(CMWE)$?

- A. $P(C)P(M|C)P(W|CM)P(E|CMW)$
- B. $P(M)P(C|M)P(W|MC)P(E|MCW)$
- C. $P(W)P(E|W)P(CM|EW)$
- D. A, B, and C
- E. None/other



Quick check

$$P(E_1 E_2 E_3 \dots E_n) = P(E_1) P(E_2 | E_1) \dots P(E_n | E_1 E_2 \dots E_{n-1})$$

Chain Rule

You are going to a friend's Halloween party.

Let C = there is candy
 M = there is music

E = no one wears your costume
 W = you wear a costume

An awesome party means that all of these events must occur.

What is $P(\text{awesome party}) = P(CMEW)$?

A. $P(C)P(M|C)P(E|CM)P(W|CME)$

B. $P(M)P(C|M)P(E|MC)P(W|MCE)$

C. $P(W)P(E|W)P(CM|EW)$

D. A, B, and C

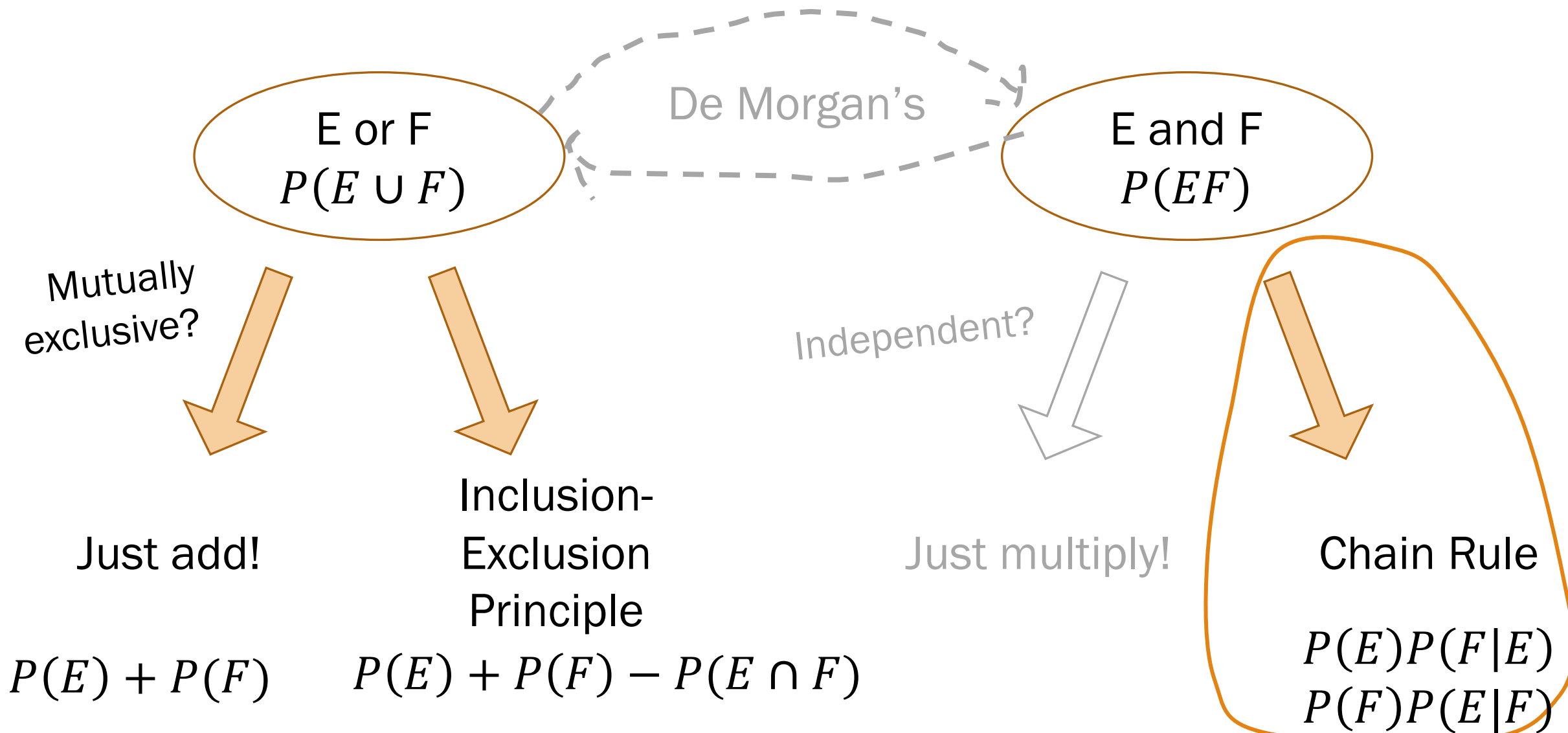
E. None/other



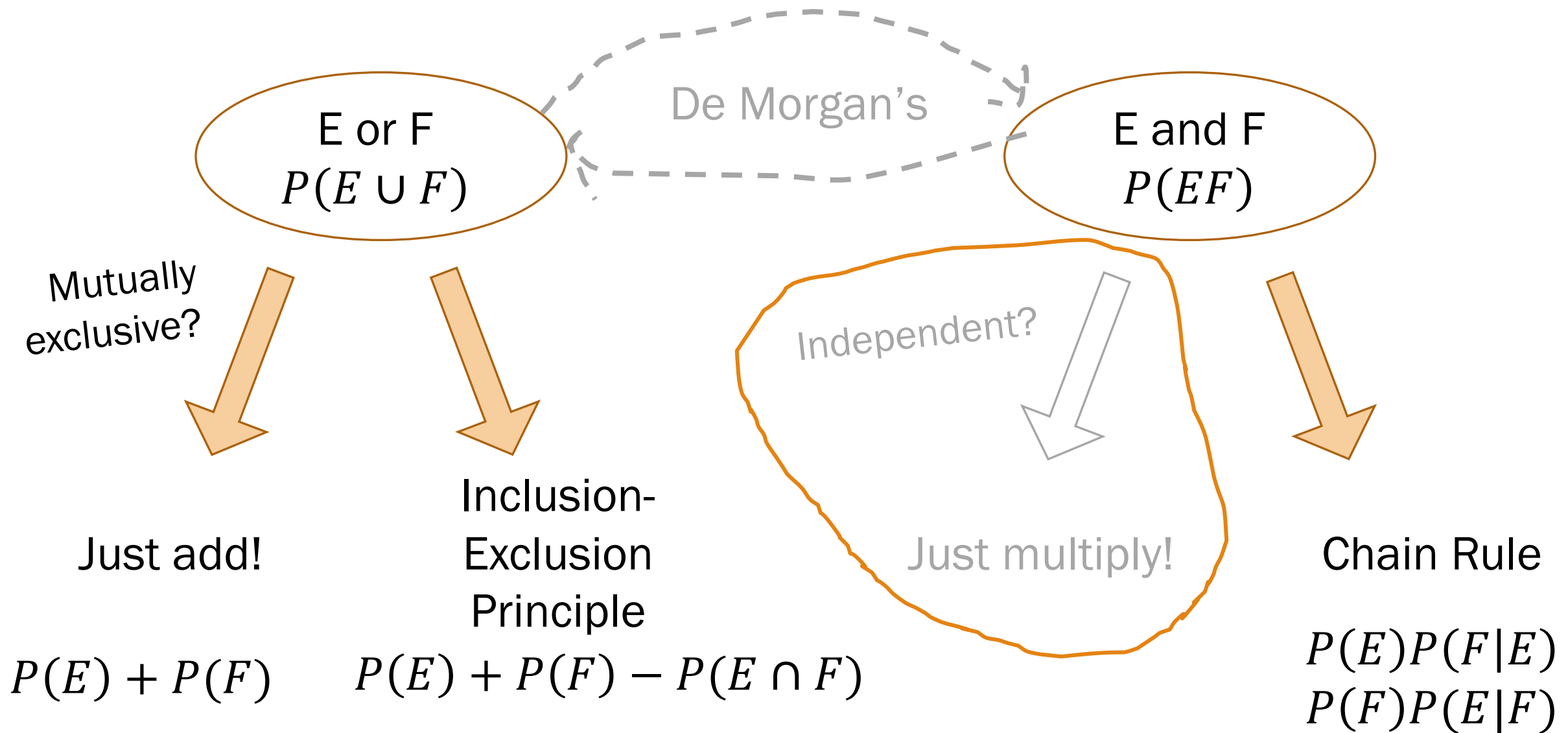
Chain Rule is a way of introducing “order” and “procedure” into probability.



Probability of events



Probability of events



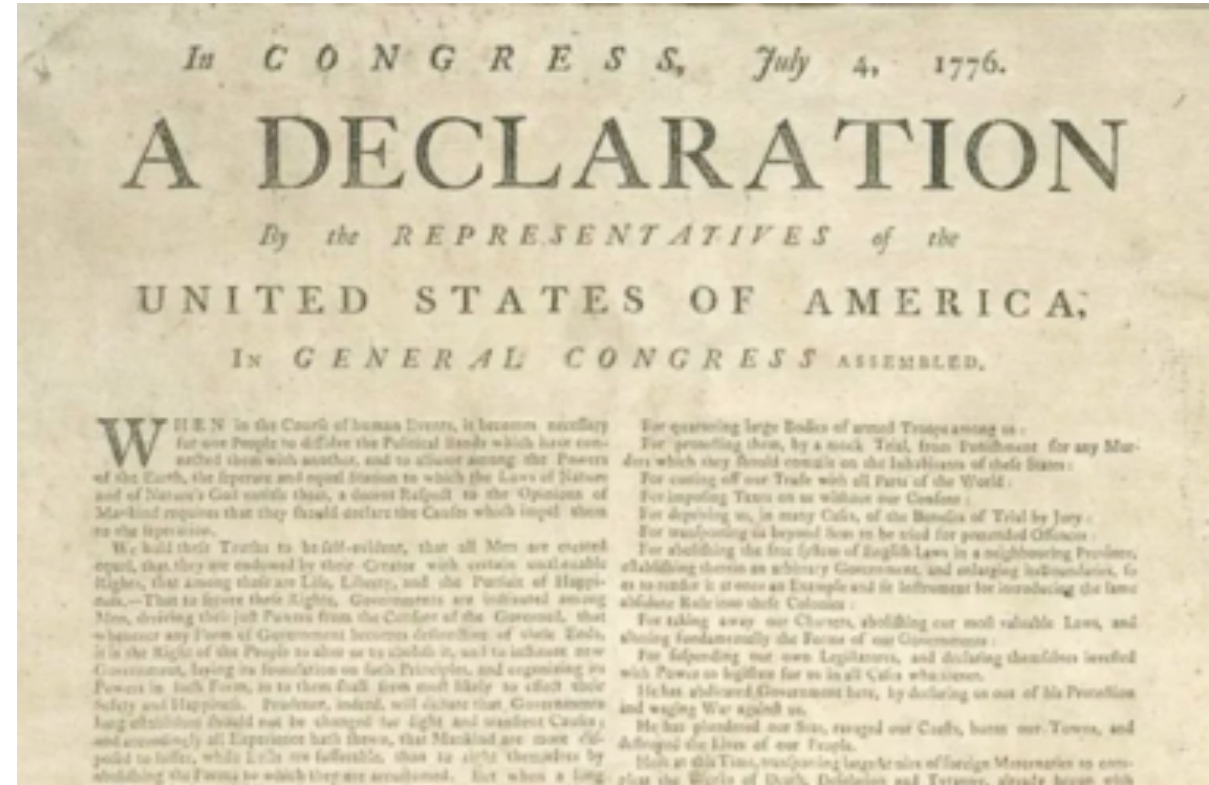
Today's plan

➔ Independence

Independent trials

De Morgan's Laws

Conditional independence (if time)



On this day in 1958, Guinea declared independence from France.

Independence

Two events E and F are defined as independent if:

$$P(EF) = P(E)P(F)$$

Otherwise E and F are called dependent events.

An equivalent definition:

$$P(E|F) = P(E)$$

Intuition through proof

Independent events E and F \iff $P(EF) = P(E)P(F)$
 $P(E|F) = P(E)$

Statement:

If E and F are independent, then $P(E|F) = P(E)$.

Proof:

$$P(E|F) = \frac{P(EF)}{P(F)}$$

Definition of conditional probability

$$= \frac{P(E)P(F)}{P(F)}$$

Independence of E and F

$$= P(E)$$

Taking the bus to cancellation city



Knowing that F happened **does not change** our belief that E happened.

Dice, our misunderstood friends

Independent events E and F \iff $P(EF) = P(E)P(F)$
 $P(E|F) = P(E)$

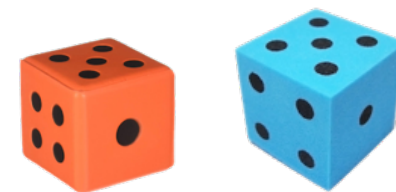
- Roll two 6-sided dice, yielding values D_1 and D_2 .

- Let event E : $D_1 = 1$

- event F : $D_2 = 6$

- event G : $D_1 + D_2 = 5$

$$G = \{(1,4), (2,3), (3,2), (4,1)\}$$



1. Are E and F independent?

$$P(E) = 1/6$$

$$P(F) = 1/6$$

$$P(EF) = 1/36$$

independent

2. Are E and G independent?

$$P(E) = 1/6$$

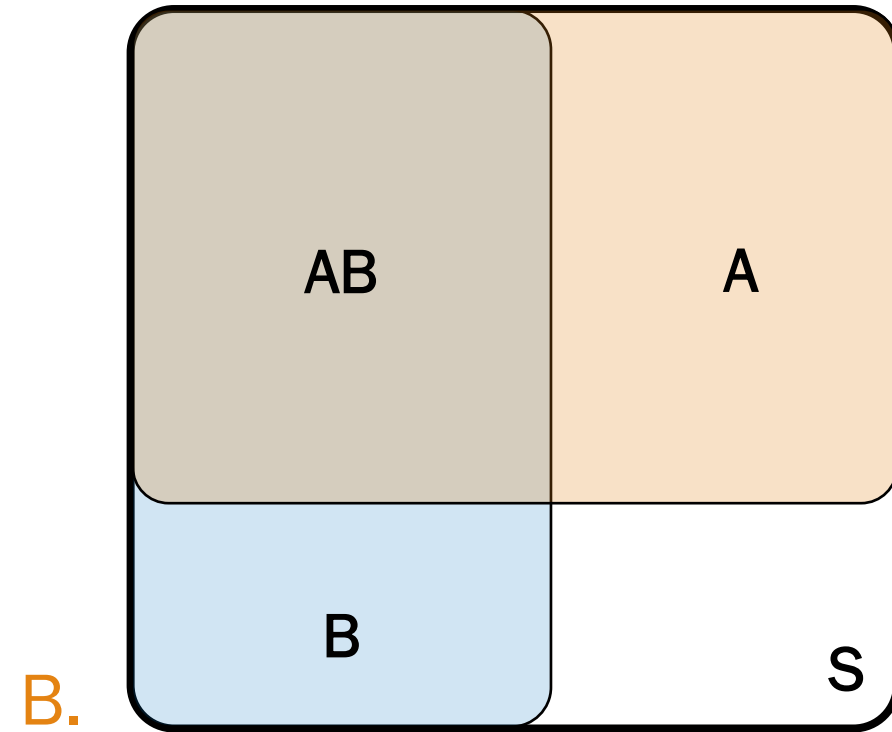
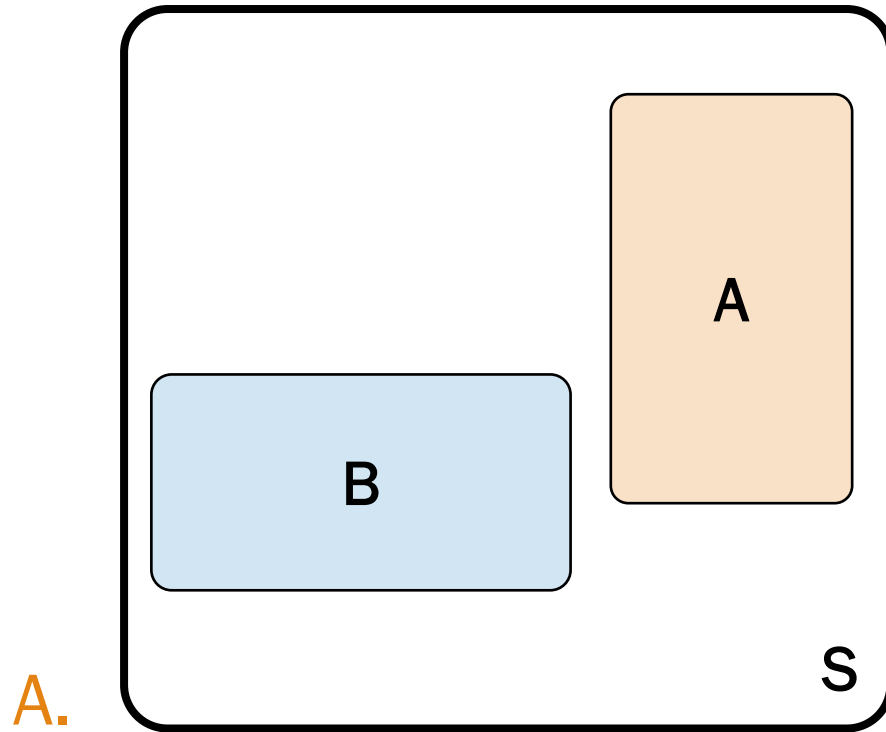
$$P(G) = 4/36 = 1/9$$

$$P(EG) = 1/36 \neq P(E)P(G)$$

dependent

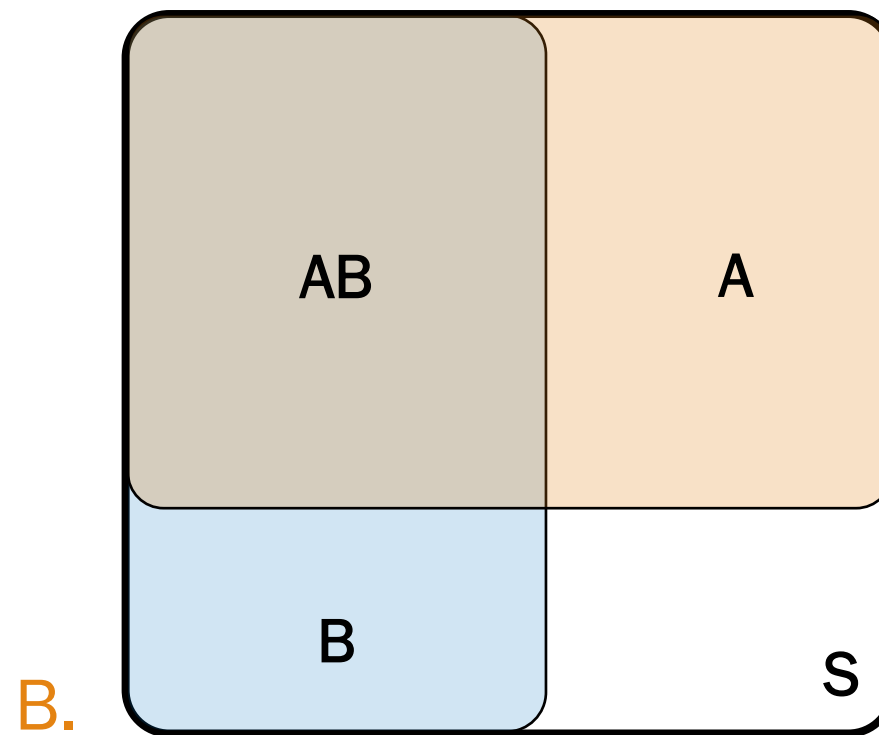
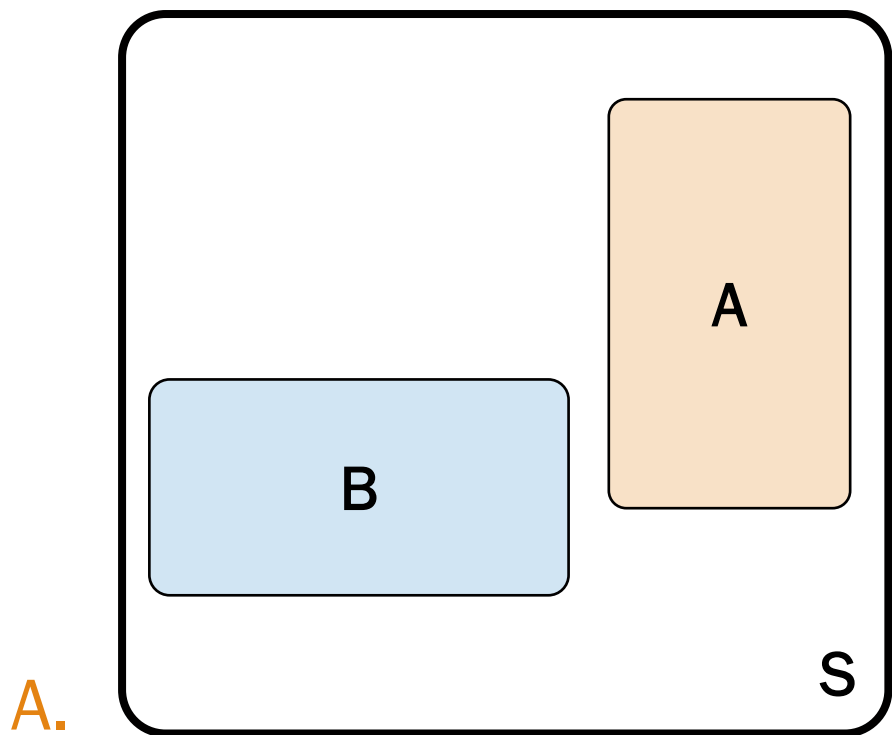
Independence?

Independent events E and F \iff $P(EF) = P(E)P(F)$
 $P(E|F) = P(E)$



Independence?

Independent events E and F \iff $P(EF) = P(E)P(F)$
 $P(E|F) = P(E)$



(Def. 1, assuming equally likely outcomes)

$$P(AB) = P(A)P(B)$$

0

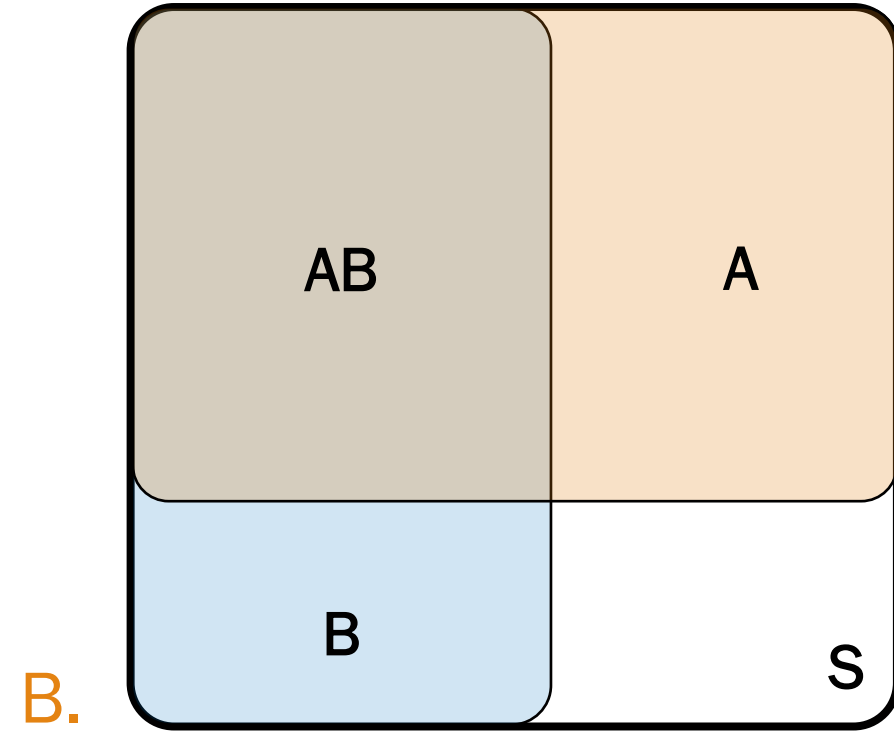
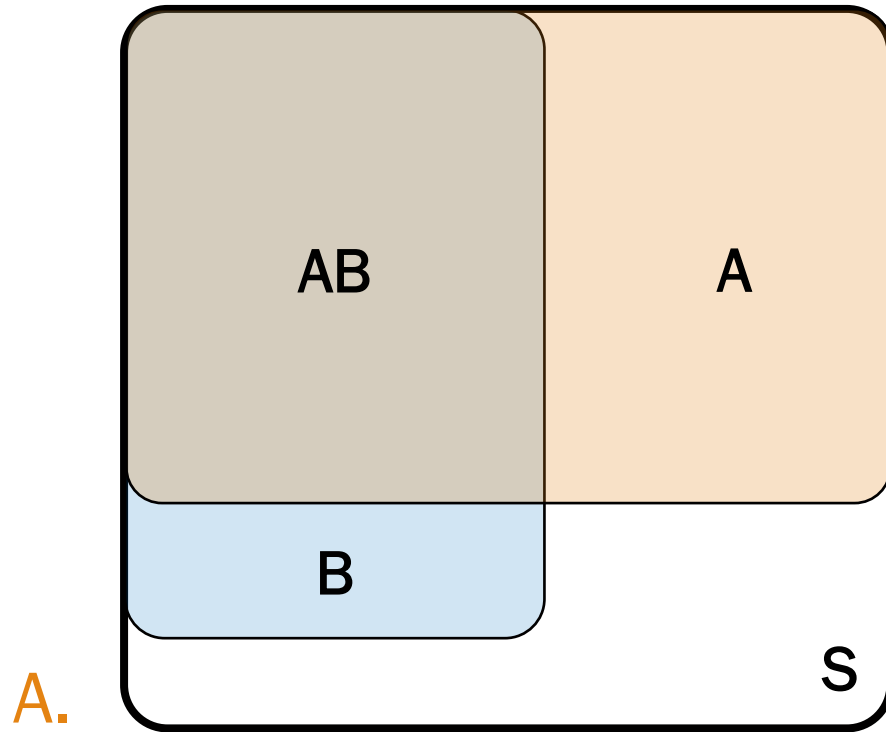
$$\frac{|AB|}{|S|} = \frac{|A|}{|S|} \times \frac{|B|}{|S|}$$

$$\frac{|AB|}{|S|} = \frac{|A|}{|S|} \times \frac{|B|}{|S|}$$



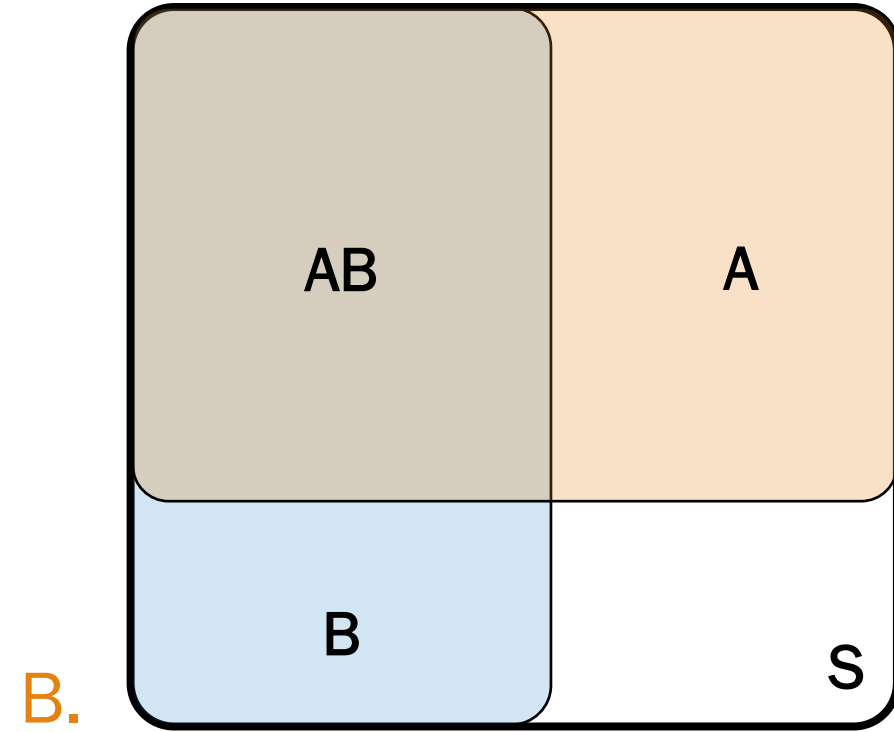
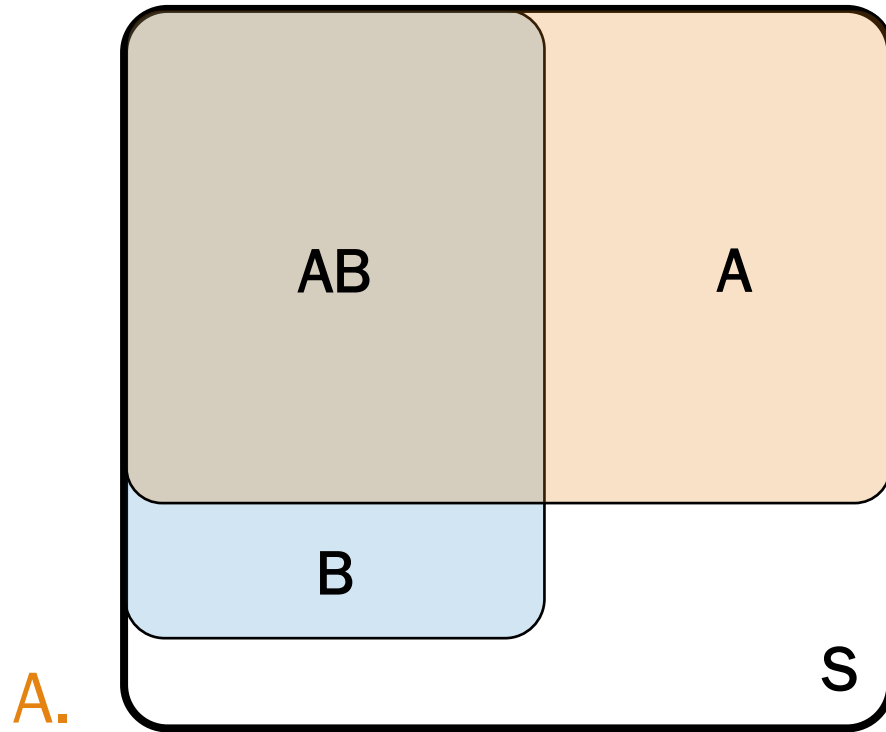
Independence?

Independent events E and F \Leftrightarrow $P(EF) = P(E)P(F)$
 $P(E|F) = P(E)$



Independence?

Independent events E and F \iff $P(EF) = P(E)P(F)$
 $P(E|F) = P(E)$



(Def. 2, assuming equally likely outcomes)

$$P(A|B) = P(A)$$

$$\frac{|AB|}{|B|} \neq \frac{|A|}{|S|}$$

$$\frac{|AB|}{|B|} = \frac{|A|}{|S|}$$



Independence of complements

Independent events E and F \iff $P(EF) = P(E)P(F)$
 $P(E|F) = P(E)$

Statement:

If E and F are independent, then E and F^C are independent.

Proof:

$$\begin{aligned}P(EF^C) &= P(E) - P(EF) \\ &= P(E) - P(E)P(F) \\ &= P(E)[1 - P(F)] \\ &= P(E)P(F^C)\end{aligned}$$

E and F^C are independent

Intersection

Independence of E and F

Factoring

Complement

Definition of independence



Knowing that F **didn't happen** does not change our belief that E happened.

Today's plan

Independence

→ Independent trials

De Morgan's Laws

Conditional independence (if time)

Generalizing independence

Three events E , F , and G are independent if:

$$\left\{ \begin{array}{l} P(EFG) = P(E)P(F)P(G), \text{ and} \\ P(EF) = P(E)P(F), \text{ and} \\ P(EG) = P(E)P(G), \text{ and} \\ P(FG) = P(F)P(G) \end{array} \right.$$

n events E_1, E_2, \dots, E_n are independent if:

$$\left\{ \begin{array}{l} \text{for } r = 1, \dots, n: \\ \quad \text{for every subset } E_1, E_2, \dots, E_r: \\ \quad \quad P(E_1, E_2, \dots, E_r) = P(E_1)P(E_2) \cdots P(E_r) \end{array} \right.$$


Independent trials:

Outcomes of n separate flips of a coin are all independent of one another.

Each flip in this case is a trial of the experiment.

Dice, increasingly misunderstood (still our friends)

- Each roll of a 6-sided die is an **independent trial**.
- Two rolls: D_1 and D_2 .
- Let event E : $D_1 = 1$
event F : $D_2 = 6$
event G : $D_1 + D_2 = 7$ $G = \{(1,6), (2,5), (3,4), (4,3), (5,2), (6,1)\}$

1. Are E and F  independent?
2. Are E and G independent?
3. Are F and G independent?
4. Are E, F, G independent?

$$P(E) = 1/6$$

$$P(F) = 1/6$$

$$P(EF) = 1/36$$



Dice, increasingly misunderstood (still our friends)

- Each roll of a 6-sided die is an **independent trial**.
- Two rolls: D_1 and D_2 .
- Let event E : $D_1 = 1$
event F : $D_2 = 6$
event G : $D_1 + D_2 = 7$ $G = \{(1,6), (2,5), (3,4), (4,3), (5,2), (6,1)\}$

1. Are E and F independent?
2. Are E and G independent?
3. Are F and G independent?
4. Are E, F, G independent?

$$P(E) = 1/6$$

$$P(F) = 1/6$$

$$P(EF) = 1/36$$

$$P(E) = 1/6$$

$$P(G) = 1/6$$

$$P(EG) = 1/36$$

$$P(F) = 1/6$$

$$P(G) = 1/6$$

$$P(FG) = 1/36$$

$$P(EFG) = 1/36$$

$$\neq \frac{1}{6} \cdot \frac{1}{6} \cdot \frac{1}{6}$$

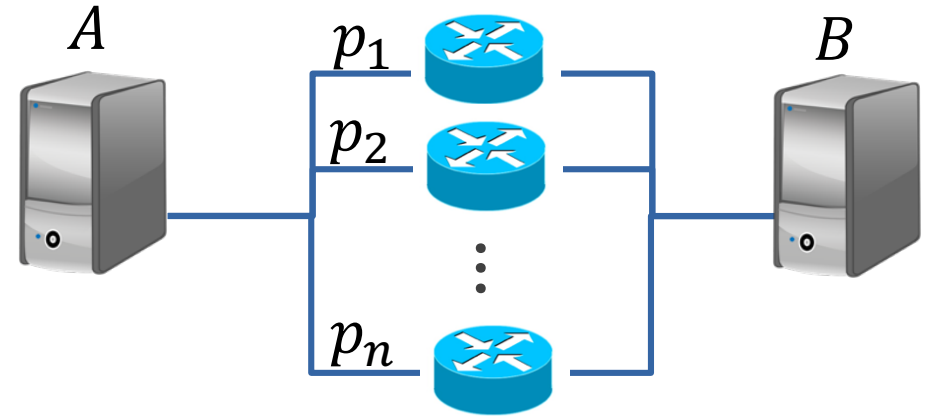


Pairwise independence is not sufficient to prove independence of >2 events!

Network reliability

Consider the following parallel network:

- n independent routers, each with probability p_i of functioning (where $1 \leq i \leq n$)
- E = functional path from A to B exists.



What is $P(E)$?

$$\begin{aligned} P(E) &= P(\geq 1 \text{ one router works}) \\ &= 1 - P(\text{all routers fail}) \\ &= 1 - (1 - p_1)(1 - p_2) \cdots (1 - p_n) \\ &= 1 - \prod_{i=1}^n (1 - p_i) \end{aligned}$$

👉 ≥ 1 : take complement

(biased) Coin Flips

Suppose we flip a coin n times.

- A coin comes up heads with probability p .
- Each coin flip is an **independent trial**.

1. $P(n \text{ heads on } n \text{ coin flips})$

$$p^n$$

2. $P(n \text{ tails on } n \text{ coin flips})$

$$(1 - p)^n$$

(biased) Coin Flips

Suppose we flip a coin n times.

- A coin comes up heads with probability p .
- Each coin flip is an **independent trial**.

1. $P(n$ heads on n coin flips)

$$p^n$$

2. $P(n$ tails on n coin flips)

$$(1 - p)^n$$

3. $P(\text{first } k \text{ heads, then } n - k \text{ tails})$

4. $P(\text{exactly } k \text{ heads on } n \text{ coin flips})$



(biased) Coin Flips

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$$p^n$$

2. $P(n$ tails on n coin flips)

$$(1 - p)^n$$

3. $P(\text{first } k \text{ heads, then } n - k \text{ tails})$

$$p^k (1 - p)^{n-k}$$

4. $P(\text{exactly } k \text{ heads on } n \text{ coin flips})$

$$\binom{n}{k} p^k (1 - p)^{n-k}$$

of mutually
exclusive
outcomes

$P(\text{a particular outcome's}$
 k heads on n coin flips)



Break for jokes/
announcements

Announcements

Section

Starts: today
Late signups/changes: by end of day
Solutions: end of week

Problem Set 1

Gradescope: entry code M7B45K
Assignment portal: available

Concept checks

Due date: every Tuesday 1:00pm
You can edit your response, so don't
be afraid of submitting multiple times.

This quarter

Beginning: fast-paced
Later: deep into concepts
Counting: the hardest part!

Today's plan

Independence

Independent trials

 De Morgan's Laws

Conditional independence

Augustus De Morgan

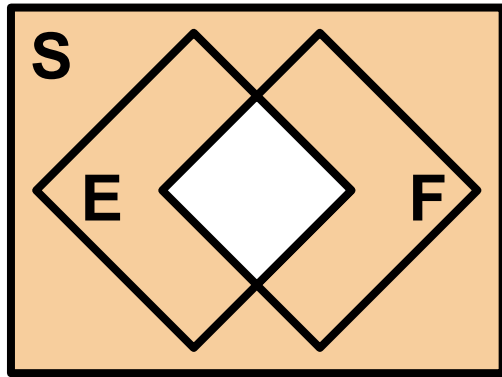
Augustus De Morgan (1806–1871):

British mathematician who wrote the book *Formal Logic* (1847).



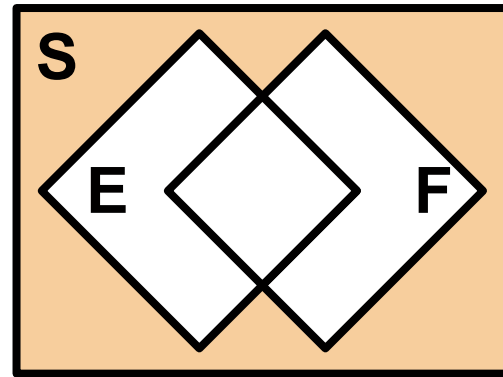
He looked remarkably similar to Jason Alexander (George from Seinfeld)
(but that's not important right now)

De Morgan's Laws



$$(E \cap F)^C = E^C \cup F^C$$

$$\left(\bigcap_{i=1}^n E_i \right)^C = \bigcup_{i=1}^n E_i^C$$



$$(E \cup F)^C = E^C \cap F^C$$

$$\left(\bigcup_{i=1}^n E_i \right)^C = \bigcap_{i=1}^n E_i^C$$

In probability:

$$P(E_1 E_2 \cdots E_n) = 1 - P(E_1^C \cup E_2^C \cup \cdots \cup E_n^C)$$

Great if E_i^C mutually exclusive!

$$P(E_1 \cup E_2 \cup \cdots \cup E_n) = 1 - P(E_1^C E_2^C \cdots E_n^C)$$

Great if E_i independent!



De Morgan's: AND \leftrightarrow OR

Hash table fun

- m strings are hashed (unequally) into a hash table with n buckets.
- Each string hashed is an **independent trial** w.p. p_i of getting hashed into bucket i .

1. $E =$ bucket 1 has ≥ 1 string hashed into it.

What is $P(E)$?

$$\begin{aligned}P(E) &= P(S_1 \cup S_2 \cup \dots \cup S_m) \\&= 1 - P\left((S_1 \cup S_2 \cup \dots \cup S_m)^c\right) \\&= 1 - P(S_1^c S_2^c \dots S_m^c) \\&= 1 - P(S_1^c)P(S_2^c) \dots P(S_m^c) = 1 - \left(P(S_1^c)\right)^m \\&= 1 - (1 - p_1)^m\end{aligned}$$

Define $S_i =$ string i is hashed into bucket 1
 $S_i^c =$ string i is not hashed into bucket 1

Complement

De Morgan's Law

S_i independent trials

↓

$$\begin{aligned}P(S_i) &= p_1 \\P(S_i^c) &= 1 - p_1\end{aligned}$$

More hash table fun

- m strings are hashed (unequally) into a hash table with n buckets.
 - Each string hashed is an independent trial w.p. p_i of getting hashed into bucket i .
1. $E =$ bucket 1 has ≥ 1 string hashed into it.
 2. $E =$ **at least 1** of buckets 1 to k has ≥ 1 string hashed into it.

What is $P(E)$?

WTF (not-real acronym for Want To Find)

Define $F_i =$ bucket i has at least one string in it

- $P(E) =$
- A. $P(F_1 F_2 \dots F_k)$
 - B. $1 - P(F_1^C) P(F_2^C) \dots P(F_k^C)$
 - C. $P(F_1 \cup F_2 \cup \dots \cup F_k)$
 - D. $P(F_1) + P(F_2) + \dots + P(F_k)$
 - E. None/other



More hash table fun

- m strings are hashed (unequally) into a hash table with n buckets.
 - Each string hashed is an **independent trial** w.p. p_i of getting hashed into bucket i .
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Define $F_i =$ bucket i has at least one string in it

- $P(E) =$
- A. $P(F_1 F_2 \dots F_k)$
 - B. $1 - P(F_1^C) P(F_2^C) \dots P(F_k^C)$
 - C.** $P(F_1 \cup F_2 \cup \dots \cup F_k)$
 - D. $P(F_1) + P(F_2) + \dots + P(F_k)$
 - E. None/other

Bucket events F_i are not independent



define well before complementing!



More hash table fun

- m strings are hashed (unequally) into a hash table with n buckets.
 - Each string hashed is an independent trial w.p. p_i of getting hashed into bucket i .
1. $E =$ bucket 1 has ≥ 1 string hashed into it.
 2. $E =$ **at least 1** of buckets 1 to k has ≥ 1 string hashed into it.

What is $P(E)$?

$$\text{WTF: } P(E) = P(F_1 \cup F_2 \cup \dots \cup F_k)$$

$$= 1 - P\left((F_1 \cup F_2 \cup \dots \cup F_k)^c\right)$$

$$= 1 - P(F_1^c F_2^c \dots F_k^c)$$

$$P(E) = 1 - (1 - p_1 - p_2 \dots - p_k)^m$$

Define $F_i =$ bucket i has at least one string in it

$$\begin{aligned} &= P(\text{no strings hashed to buckets 1 to } k) \\ &= \left(P(\text{string hashed outside bkts 1 to } k)\right)^m \\ &= (1 - p_1 - p_2 \dots - p_k)^m \end{aligned}$$

The **fun** never stops with hash tables

- m strings are hashed (unequally) into a hash table with n buckets.
 - Each string hashed is an **independent trial** w.p. p_i of getting hashed into bucket i .
1. $E =$ bucket 1 has ≥ 1 string hashed into it.
 2. $E =$ at least 1 of buckets 1 to k has ≥ 1 string hashed into it.
 3. $E =$ **each** of of buckets 1 to k has ≥ 1 string hashed into it.

What is $P(E)$?



The fun never stops with hash tables

- m strings are hashed (unequally) into a hash table with n buckets.
 - Each string hashed is an **independent trial** w.p. p_i of getting hashed into bucket i .
1. $E =$ bucket 1 has ≥ 1 string hashed into it.
 2. $E =$ at least 1 of buckets 1 to k has ≥ 1 string hashed into it.
 3. $E =$ each of buckets 1 to k has ≥ 1 string hashed into it.

What is $P(E)$?

WTF: $P(E) = P(F_1 F_2 \cdots F_k)$

$$= 1 - P\left((F_1 F_2 \cdots F_k)^c\right)$$

$$= 1 - P\left(F_1^c \cup F_2^c \cup \cdots \cup F_k^c\right)$$

$$= 1 - P\left(\bigcup_{i=1}^k F_i^c\right) = 1 - \sum_{r=1}^k (-1)^{(r+1)} \sum_{i_1 < \cdots < i_r} P\left(F_{i_1}^c F_{i_2}^c \cdots F_{i_r}^c\right)$$

where $P\left(F_{i_1}^c F_{i_2}^c \cdots F_{i_r}^c\right) = (1 - p_{i_1} - p_{i_2} \cdots - p_{i_r})^m$

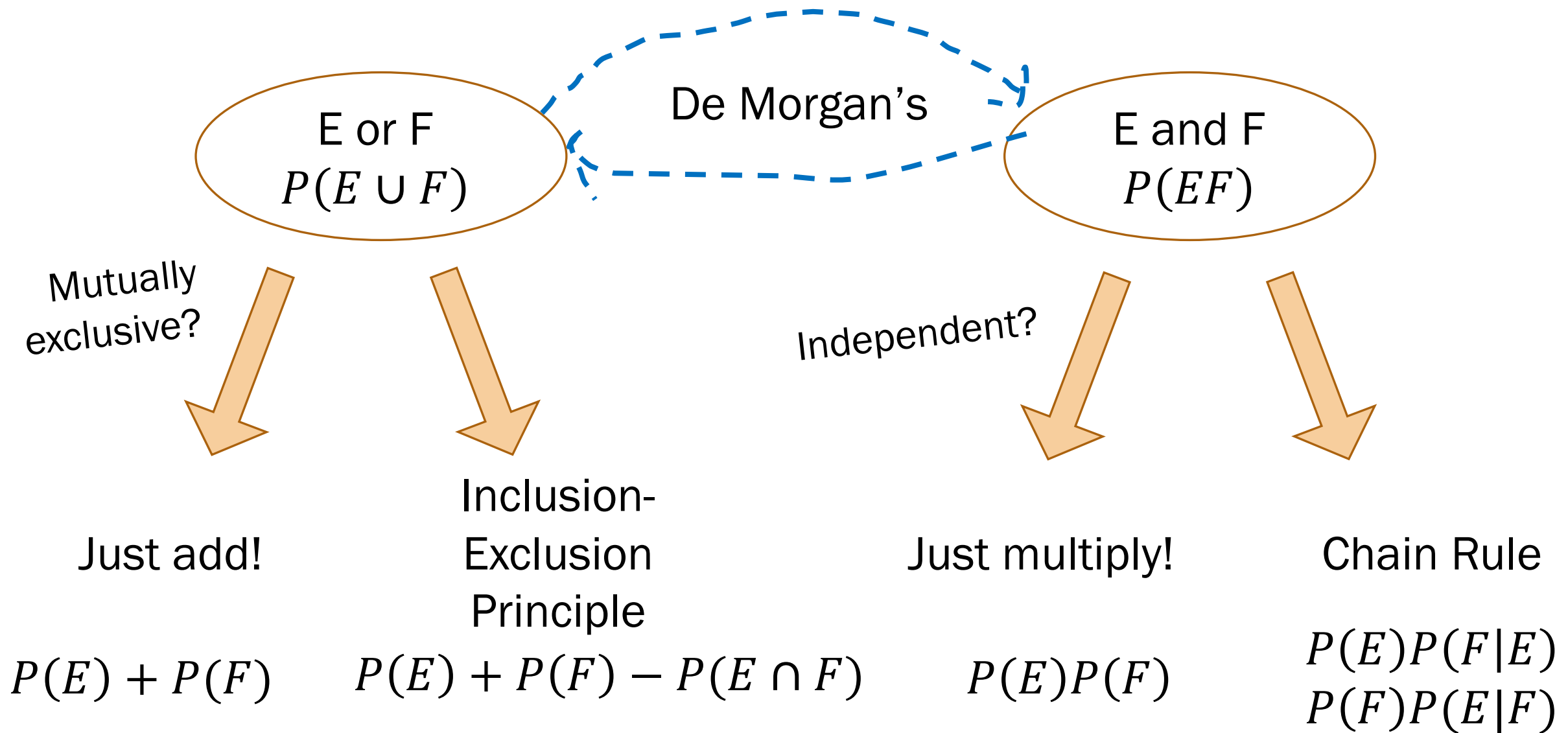
Define $F_i =$ bucket i has at least one string in it

Complement

De Morgan's Law

It is expected that this last example will
need some review!

Probability of events



Today's plan

Independence

Independent trials

De Morgan's Laws

➔ Conditional independence (if time)