

05: Independence

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Adapted from slides by Lisa Yan

Monty Hall Problem from Let's Make a Deal

Behind one door is a car (equally likely to be any door).

Behind the other two doors are goats

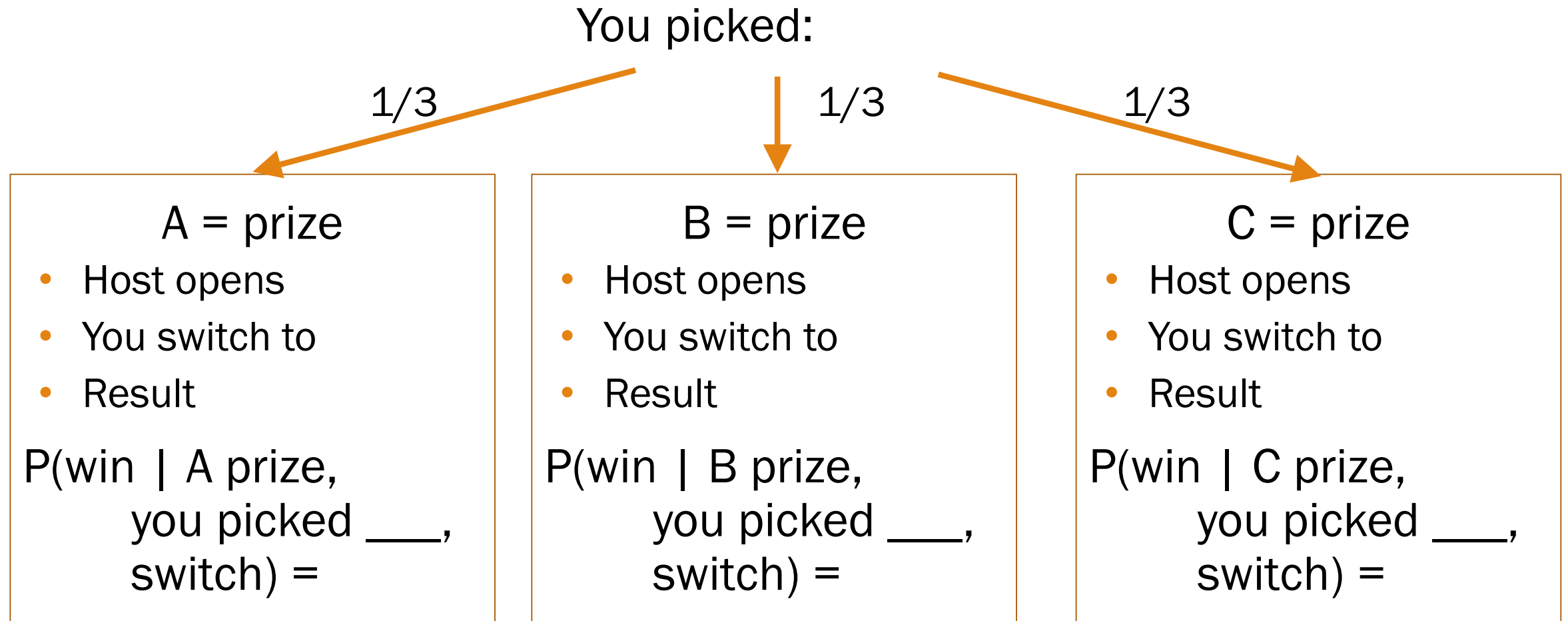
1. You choose a door
2. Host opens 1 of other 2 doors, revealing a goat
3. You are given an option to change to the other door.

Should you switch?



Doors A, B, C

What happens if you switch



$P(\text{win} \mid \text{you picked } _, \text{ switched}) =$

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).

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

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

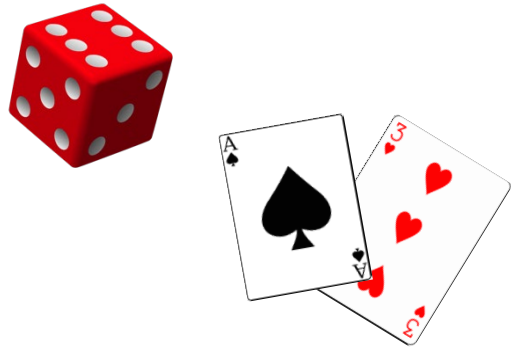
3. Should you switch?

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

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

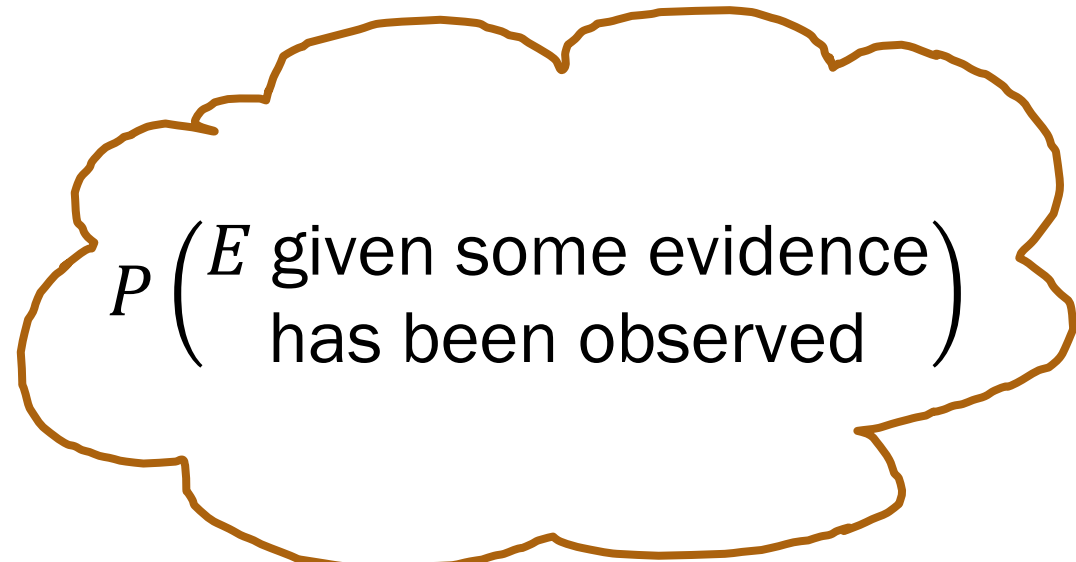
This class going forward

Last week
Equally likely
events



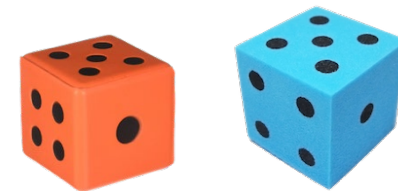
$P(E \cap F)$ $P(E \cup F)$
(counting, combinatorics)

For most of this course
Not equally likely events



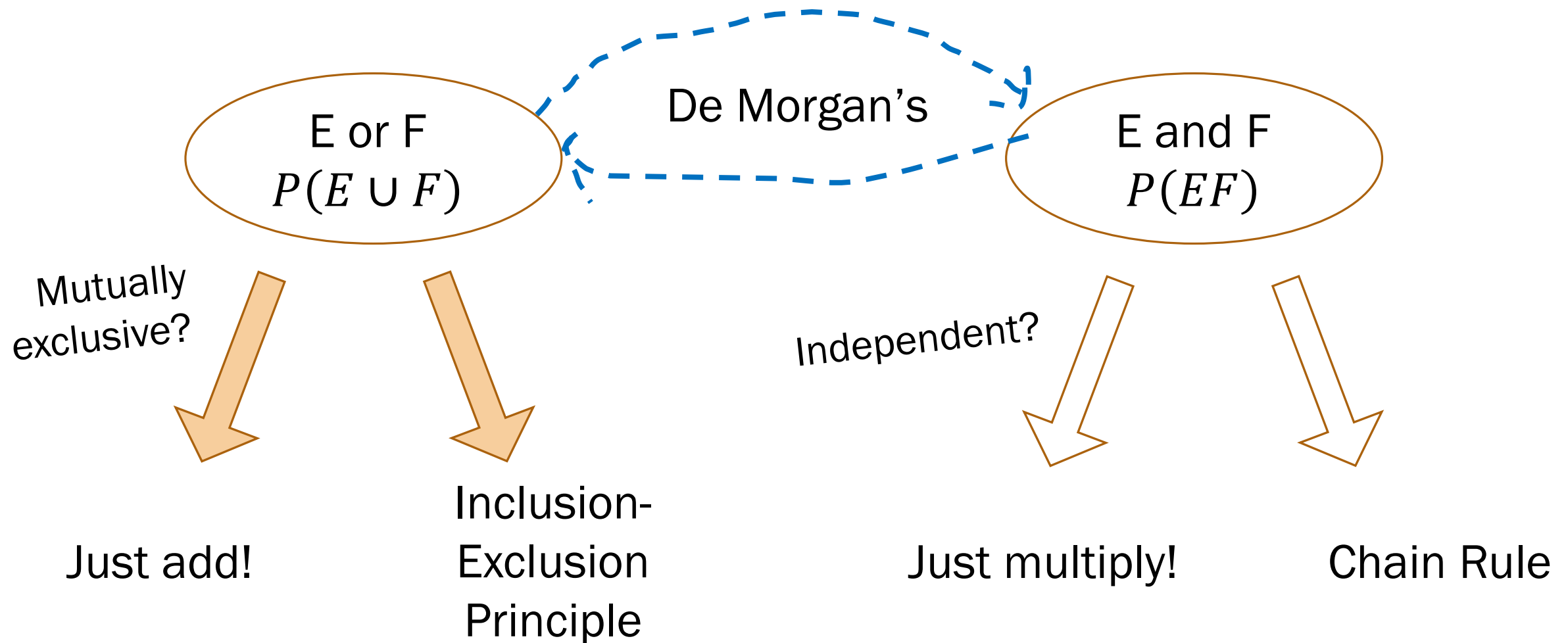
Two Dice

- 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 or both
 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 or both
- A. $P(F)$
 - B. $P(E \cup F)$
 - C. $P(E^C \cup F^C)$
 - D. $P(EF)$
 - E. $P(E^C F^C)$

Probability of events



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

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

1. Define events
& state goal

2. Identify known
probabilities

3. Solve

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

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

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

Definition of conditional probability:

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

The Chain Rule:

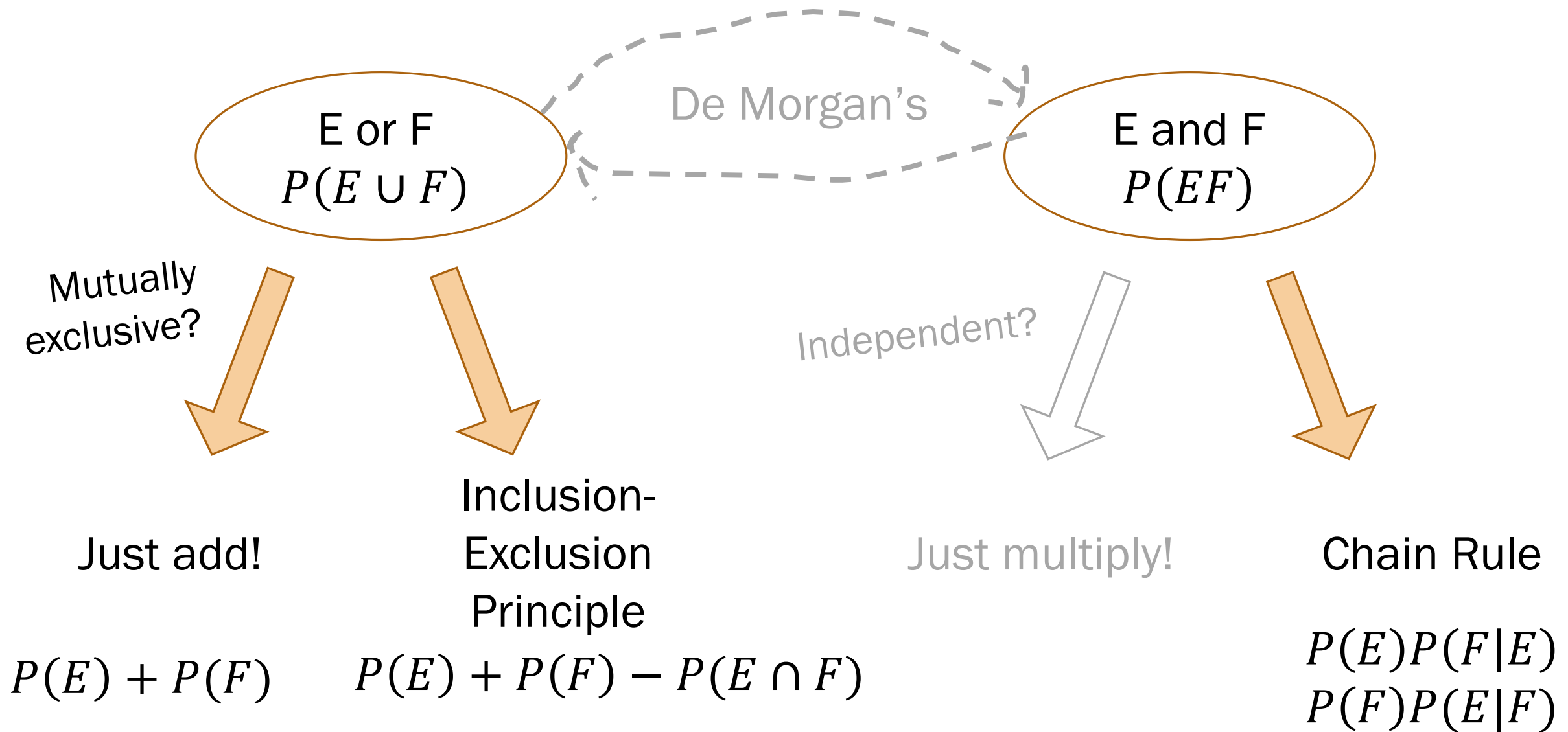
$$P(EF) = P(E|F)P(F)$$

Generalized Chain Rule

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



Probability of events



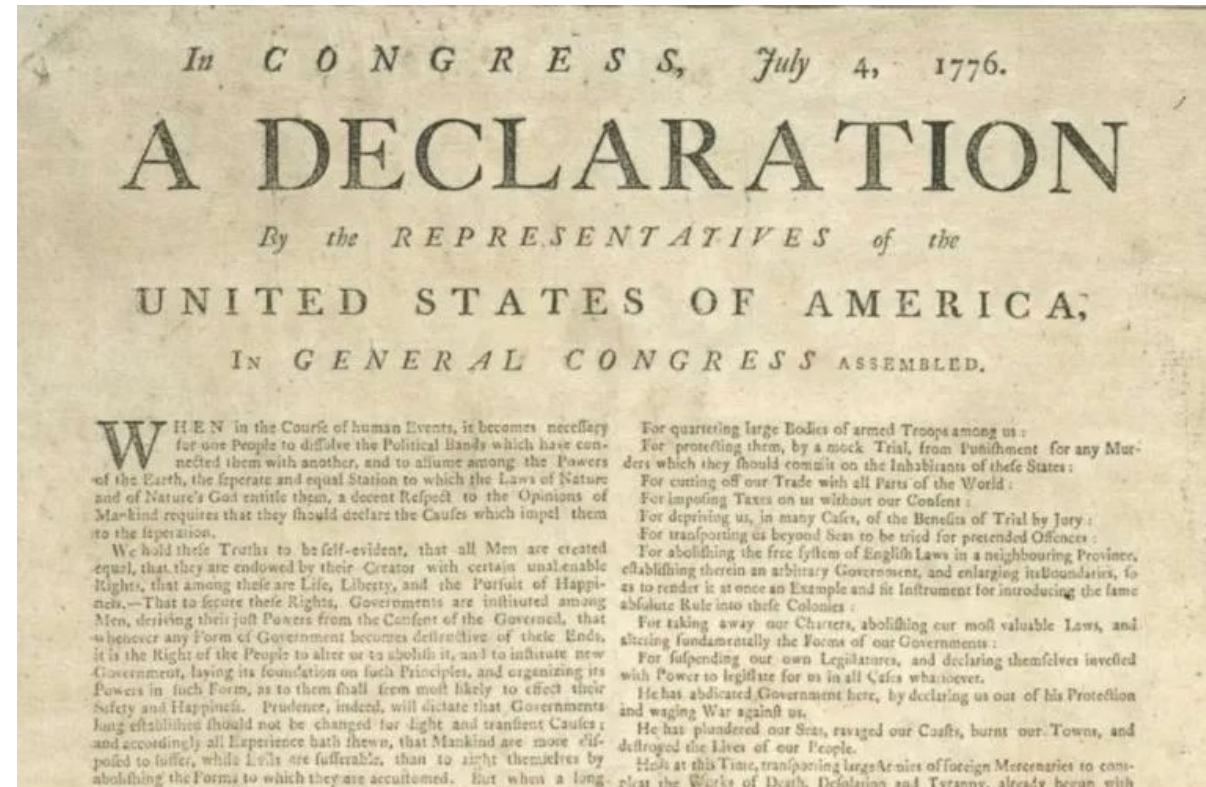
Today's plan

➔ Independence

Independent trials

De Morgan's Laws

Conditional independence (if time)



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

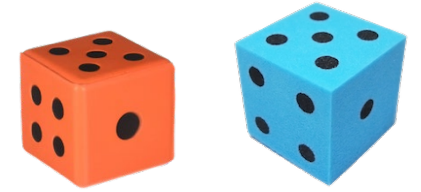
$$= P(E)$$

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 : $D_1 + D_2 = 5$



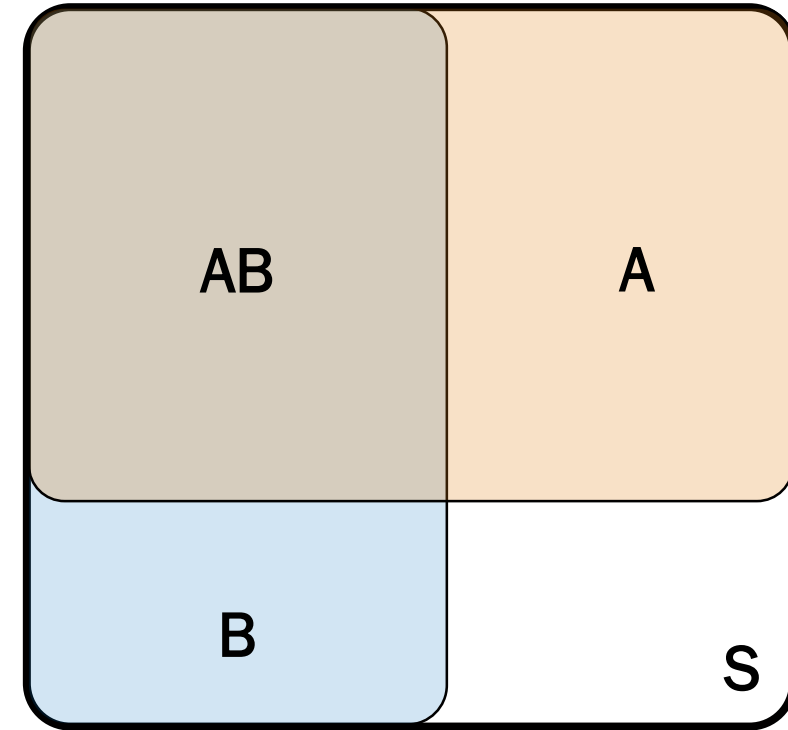
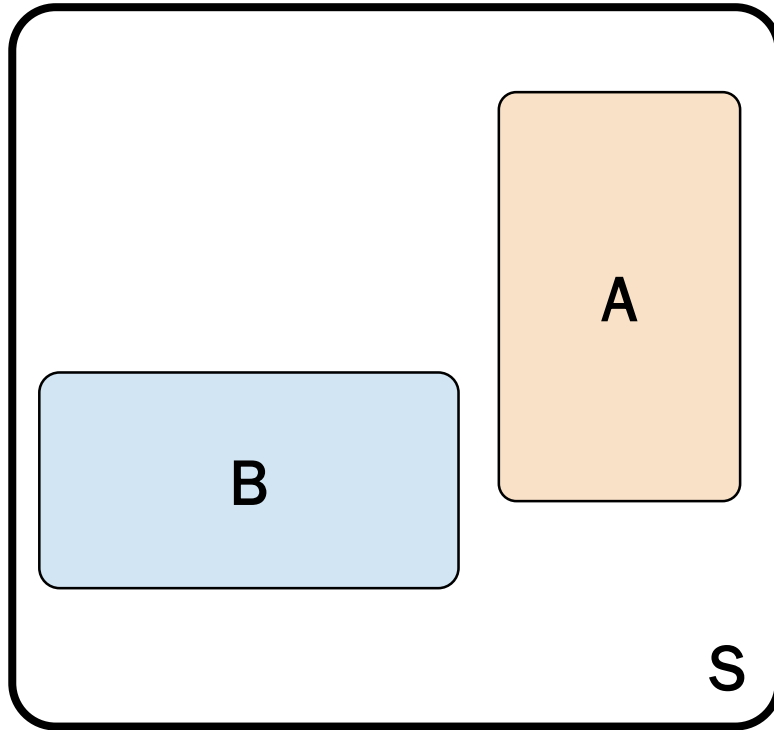
$$G = \{(1,4), (2,3), (3,2), (4,1)\}$$

1. Are E and F independent?

2. Are E and G independent?

Independence?

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



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:

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

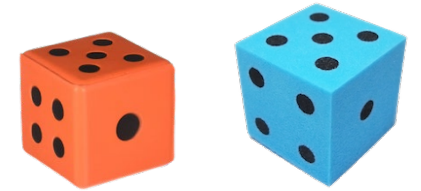
Independent trials:

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?

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

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

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

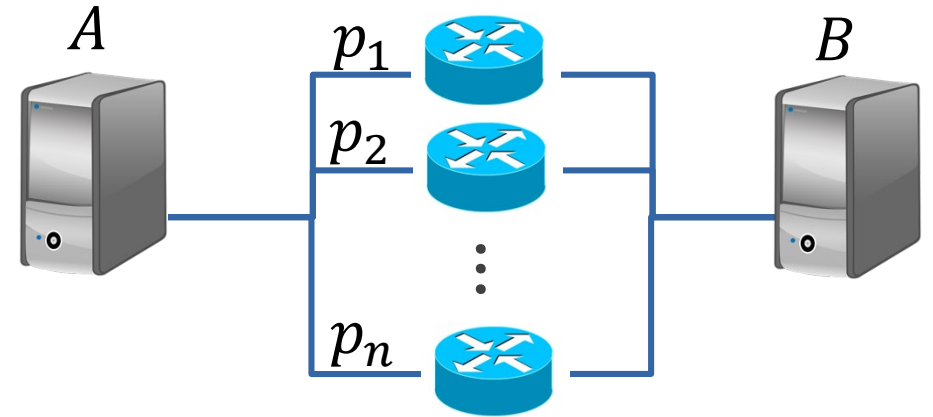
Pairwise independence is not sufficient to prove independence of >2 events!

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 $P(E)$?



(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)
2. $P(n$ tails on n coin flips)
3. $P(\text{first } k \text{ heads, then } n - k \text{ tails})$
4. $P(\text{exactly } k \text{ heads on } n \text{ coin flips})$

Announcements

Section

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

This quarter

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

Today's plan

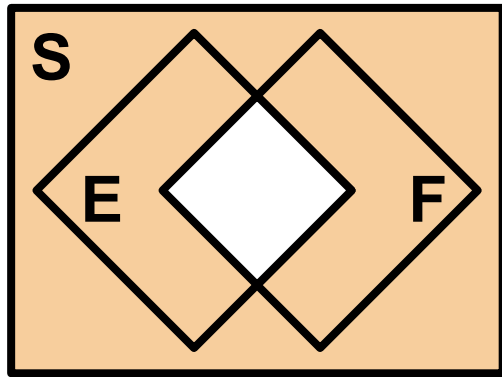
Independence

Independent trials

 De Morgan's Laws

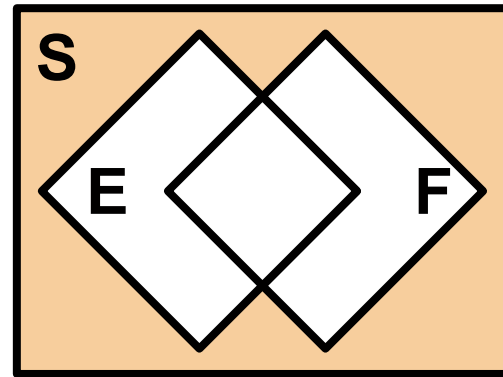
Conditional independence

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 $P(E)$?

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


$$P(S_i) = p_1$$
$$P(S_i^C) = 1 - p_1$$

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 $P(E)$?


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

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 $P(E)$?

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

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

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 $P(E)$?

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

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 $P(E)$?

$$\begin{aligned} P(E) &= P(F_1 F_2 \cdots F_k) \\ &= 1 - P\left((F_1 F_2 \cdots F_k)^c\right) && \text{Define } F_i = \text{bucket } i \text{ has at least one string in it} \\ &= 1 - P\left(F_1^c \cup F_2^c \cup \cdots \cup F_k^c\right) && \text{Complement} \\ &= 1 - P\left(\bigcup_{i=1}^k F_i^c\right) = 1 - \sum_{r=1}^k (-1)^{(r+1)} \sum_{i_1 < \cdots < i_r} P\left(F_{i_1}^c F_{i_2}^c \cdots F_{i_r}^c\right) && \text{De Morgan's Law} \end{aligned}$$

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

Today's plan

Independence

Independent trials

De Morgan's Laws

➔ Conditional independence (if time)

Conditional Paradigm

For any events A, B, and E, you can condition consistently on E, and all formulas still hold:

Axiom 1

$$0 \leq P(A|E) \leq 1$$

Corollary 1 (complement)

$$P(A|E) = 1 - P(A^c|E)$$

Commutativity

$$P(AB|E) = P(BA|E)$$

Chain Rule

$$P(AB|E) = P(B|E)P(A|BE)$$

Bayes' Theorem

$$P(A|BE) = \frac{P(B|AE)P(A|E)}{P(B|E)}$$

Independence relationships
can change with conditioning.

A and B
independent

does NOT always
mean

A and B
independent
given E.

Conditional Independence

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

Two events A and B are defined as conditionally independent given E if:

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

An equivalent definition:

- A. $P(A|B) = P(A)$
- B. $P(A|BE) = P(A)$
- C. $P(A|BE) = P(A|E)$

Netflix and Condition

Let E = a user watches Life is Beautiful.

Let F = a user watches Amelie.

What is $P(E)$?

$$P(E) \approx \frac{\# \text{ people who have watched movie}}{\# \text{ people on Netflix}} = \frac{10,234,231}{50,923,123} \approx 0.20$$

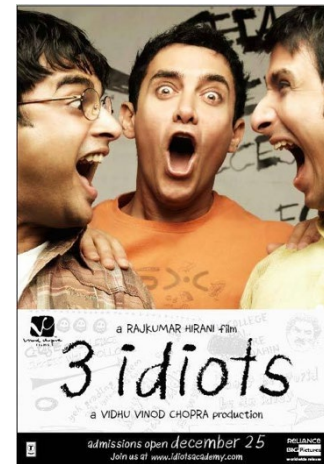
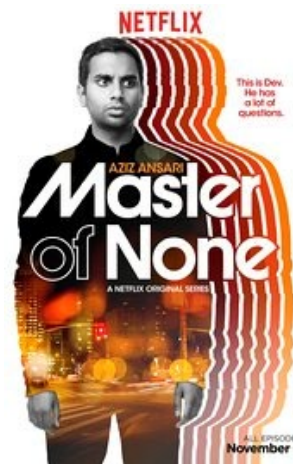


What is the probability that a user watches Life is Beautiful, given they watched Amelie?

$$P(E|F) = \frac{P(EF)}{P(F)} = \frac{\# \text{ people who have watched both}}{\# \text{ people who have watched Amelie}} \approx 0.42$$

Netflix and Condition

Let E be the event that a user watches the given movie.
Let F be the event that the same user watches Amelie.



$$P(E) = 0.19$$

$$P(E) = 0.32$$

$$P(E) = 0.20$$

$$P(E) = 0.09$$

$$P(E) = 0.20$$

$$P(E|F) = 0.14$$

$$P(E|F) = 0.35$$

$$P(E|F) = 0.20$$

$$P(E|F) = 0.72$$

$$P(E|F) = 0.42$$

Independent!

Netflix and Condition

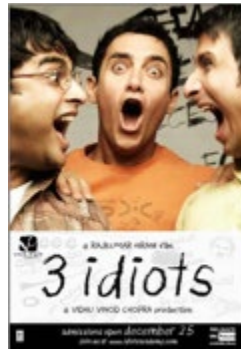
Watched:



E_1

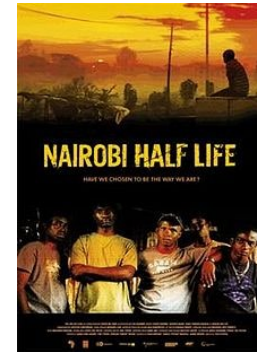


E_2



E_3

Will they watch?



E_4

What if $E_1E_2E_3E_4$ are not independent? (e.g., all international emotional comedies)

$$P(E_4|E_1E_2E_3) = \frac{P(E_1E_2E_3E_4)}{P(E_1E_2E_3)}$$

people who have watched all 4
people on Netflix

people who have watched those 3
people on Netflix

Netflix and Condition

K : likes international emotional comedies

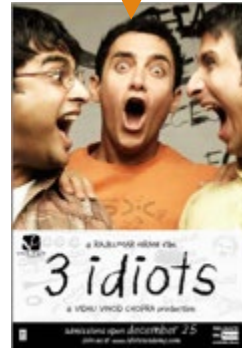
Watched:



E_1

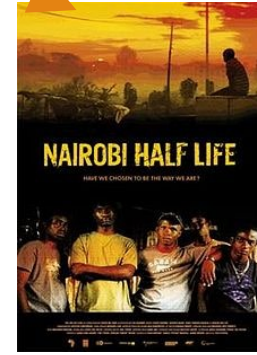


E_2



E_3

Will they watch?



E_4

What if $E_1 E_2 E_3 E_4$ are conditionally independent K ?

$$P(E_4 | E_1 E_2 E_3) = \frac{P(E_1 E_2 E_3 E_4)}{P(E_1 E_2 E_3)}$$

$$P(E_4 | E_1 E_2 E_3 K) = P(E_4 | K)$$

Conditional independence is a Big Deal

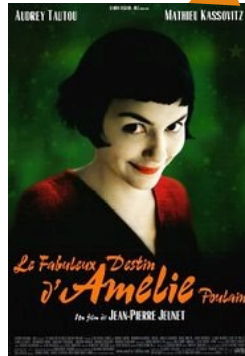
Conditional independence is a practical, real-world way of decomposing hard probability questions.

“Exploiting conditional independence to generate fast probabilistic computations is one of the main contributions CS has made to probability theory.”

–Judea Pearl wins 2011 Turing Award,
“For fundamental contributions to artificial intelligence through the development of a calculus for probabilistic and causal reasoning”

Netflix and Condition

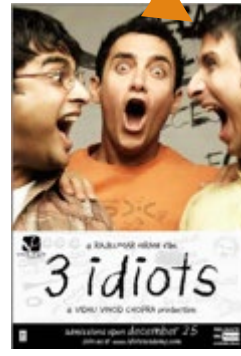
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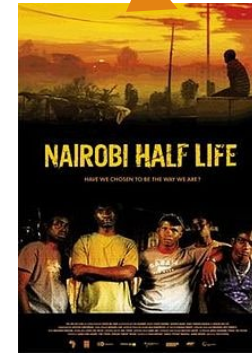
E_1



E_2



E_3



E_4

$E_1 E_2 E_3 E_4$ are
dependent

$E_1 E_2 E_3 E_4$ are
conditionally independent
given K

Dependent events can become
conditionally independent.