

Independence

Today, start with a cool program



G₁

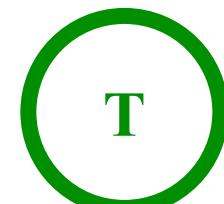
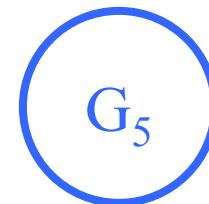
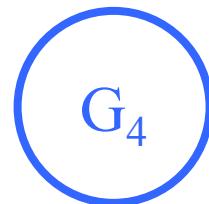
G₂

G₃

G₄

G₅

T

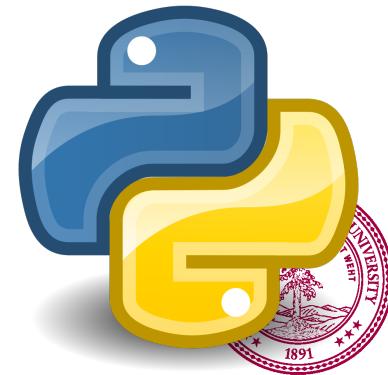


The screenshot shows a Mac OS X window titled "dna.txt" with the file path "DNA.txt — dna". The content of the file is a list of 100,000 samples, each represented by a sequence of six binary values (True or False). The first few lines of the data are:

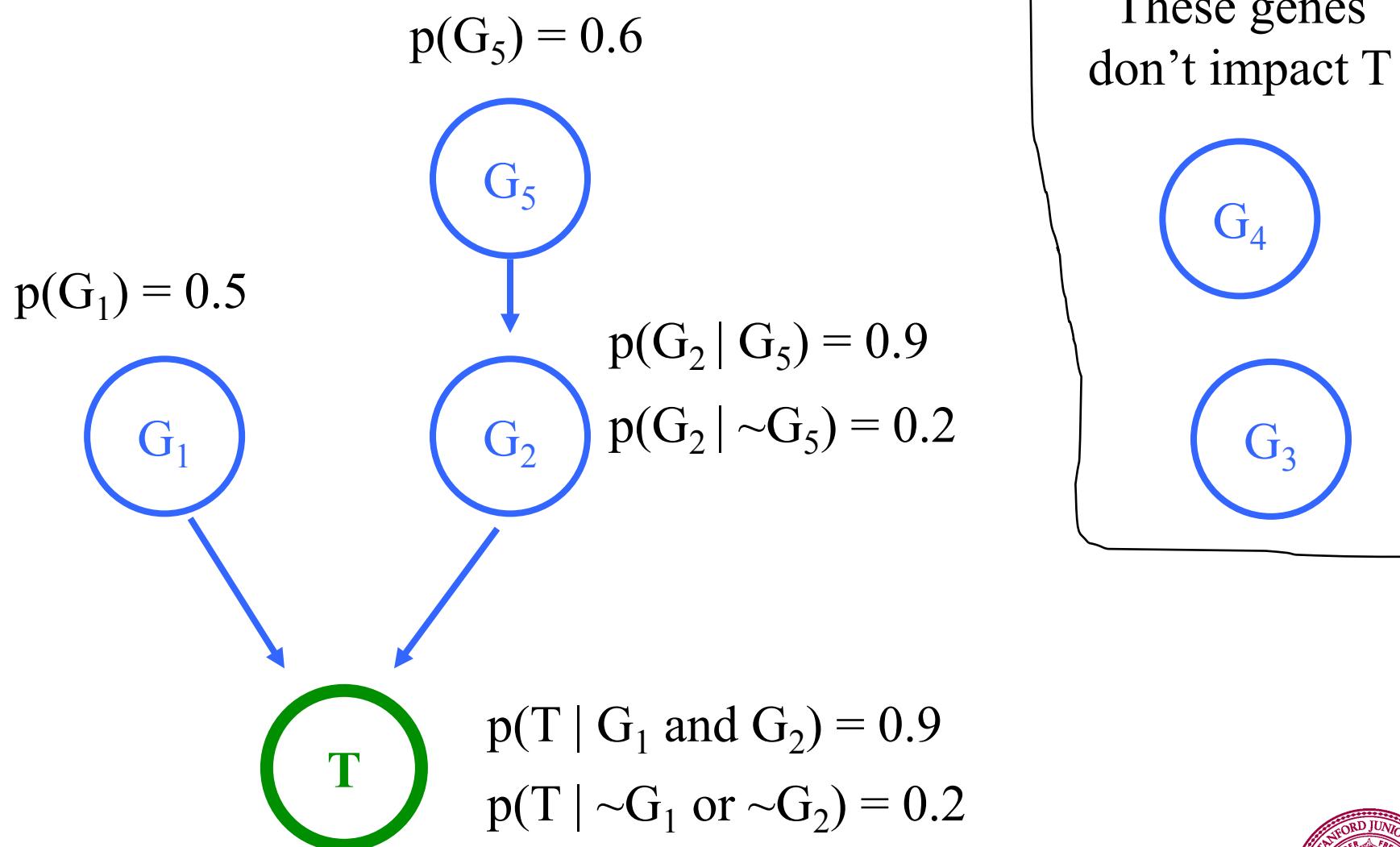
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1 False, True, False, False, True, False
2 True, True, False, True, True, False
3 True, True, False, True, True, True
4 False, True, False, True, True, False
5 False, True, False, False, True, False
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14 False, False, True, True, False, False
15 True, True, False, False, True, True
16 True, False, True, True, False, False
17 True, True, True, True, True, True
18 True, False, True, False, False, True
19 False, True, False, True, True, True
20 False, False, True, False, False, False
21 False, False, False, True, True, False
22 False, True, False, False, True, False
23 True, True, False, True, True, True
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25 True, False, False, False, False, True
26 False, False, True, True, False, True
27 False, False, False, True, False, False
28 False, True, True, False, False, True
29 False, True, False, False, True, True
30 False, False, False, False, False, True
31 False, True, False, True, True, False
32 True, False, False, True, False, False
33 True, True, False, True, True, True
34 True, True, False, False, True, True
35 True, True, False, True, True, True
36 False, False, False, True, False, False
--
```

6 observations per sample

100,000
samples



Discovered Pattern



We've gotten ahead of ourselves



Source: The Ho

Start at the beginning



Source: The Ho

And vs Condition

$P(AB)$ vs $P(A|B)$

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

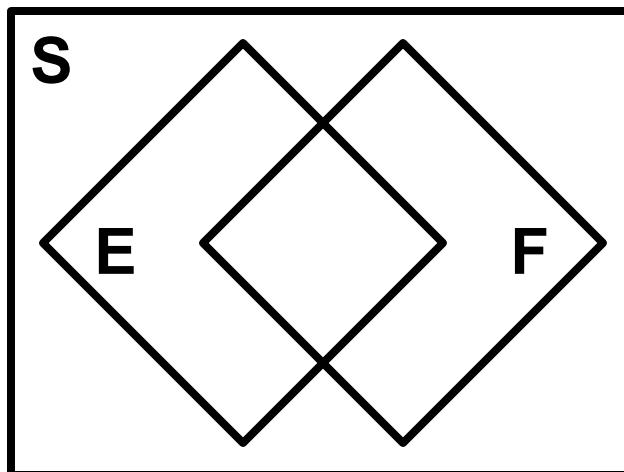


Sets Review



Set Operations Review

- Say E and F are subsets of S

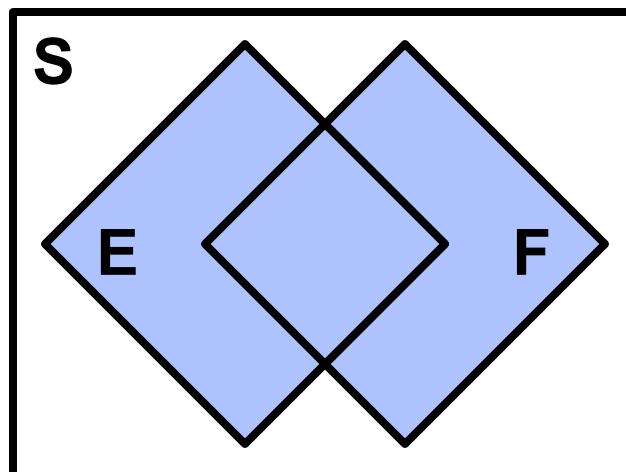


Set Operations Review

- Say E and F are events in S

Event that is in E or F

$$E \cup F$$



- $S = \{1, 2, 3, 4, 5, 6\}$ die roll outcome
- $E = \{1, 2\}$ $F = \{2, 3\}$ $E \cup F = \{1, 2, 3\}$

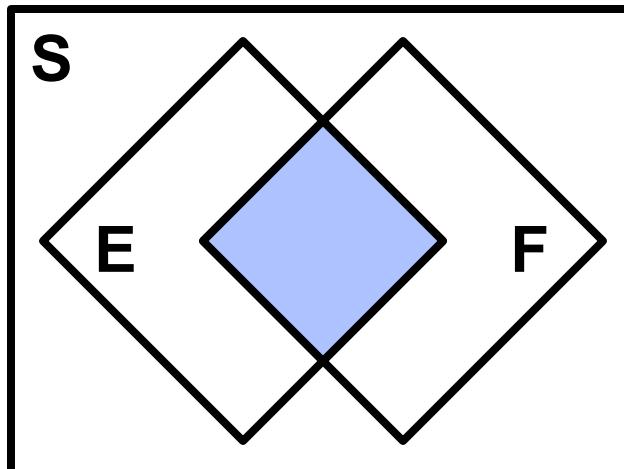


Set Operations Review

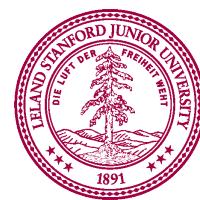
- Say E and F are events in S

Event that is in E and F

$$E \cap F \text{ or } EF$$



- $S = \{1, 2, 3, 4, 5, 6\}$ die roll outcome
- $E = \{1, 2\}$ $F = \{2, 3\}$ $E F = \{2\}$
- **Note:** mutually exclusive events means $E F = \emptyset$

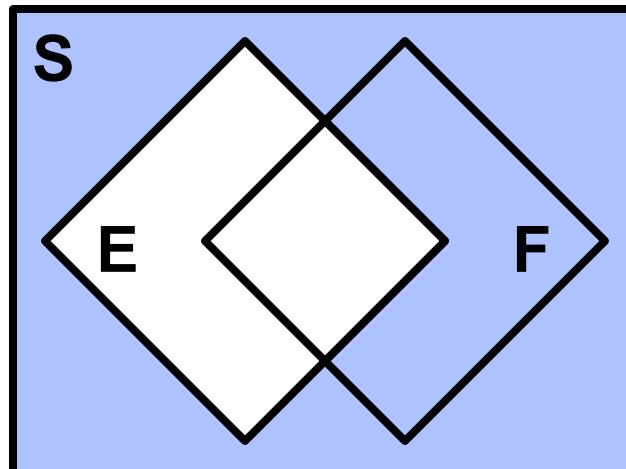


Set Operations Review

- Say E and F are events in S

Event that is not in E (called complement of E)

$$E^c \text{ or } \sim E$$

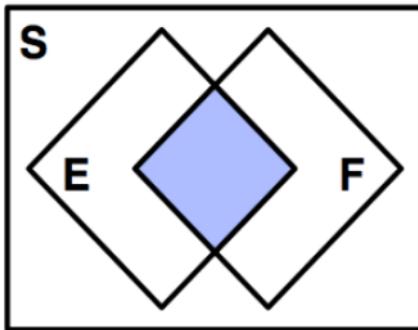


- $S = \{1, 2, 3, 4, 5, 6\}$ die roll outcome
- $E = \{1, 2\}$ $E^c = \{3, 4, 5, 6\}$

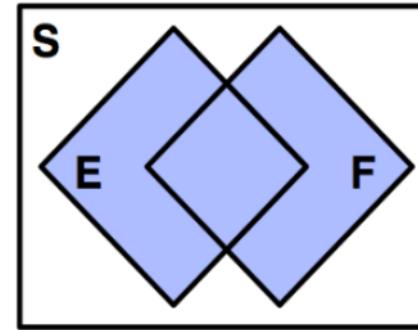


Which is the correct picture for $E^c \cap F^c$

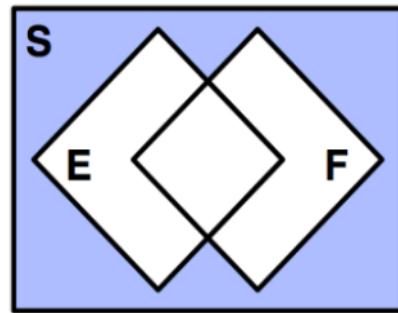
A



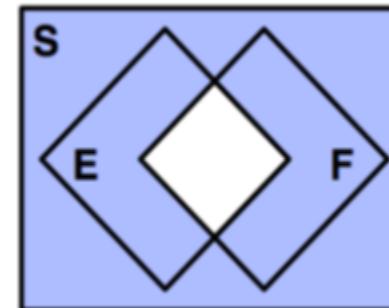
C



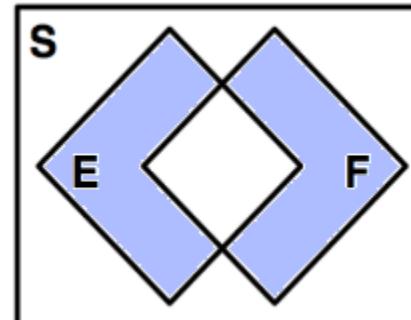
B



D



E

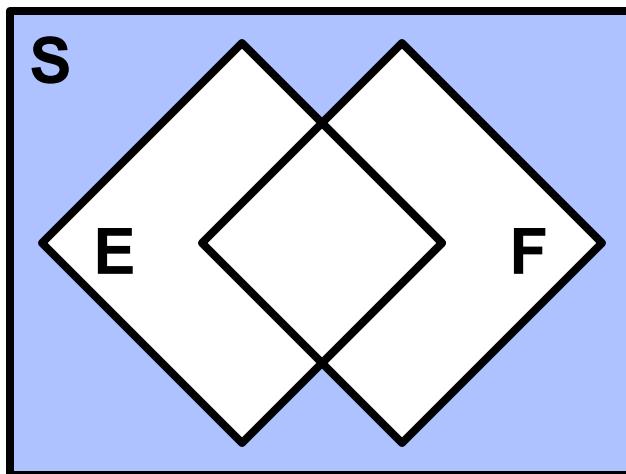


Set Operations Review

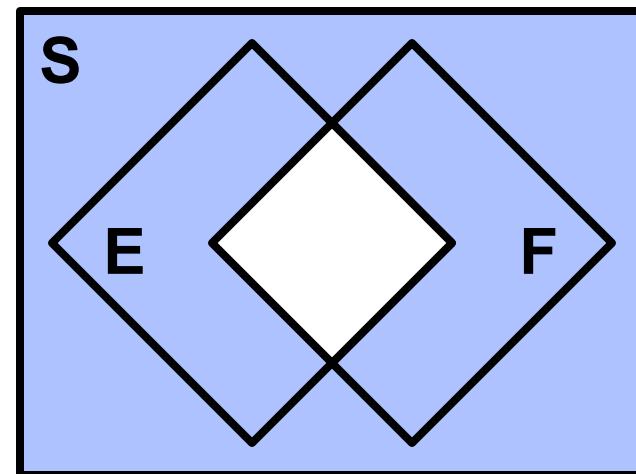
- Say E and F are events in S

DeMorgan's Laws

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



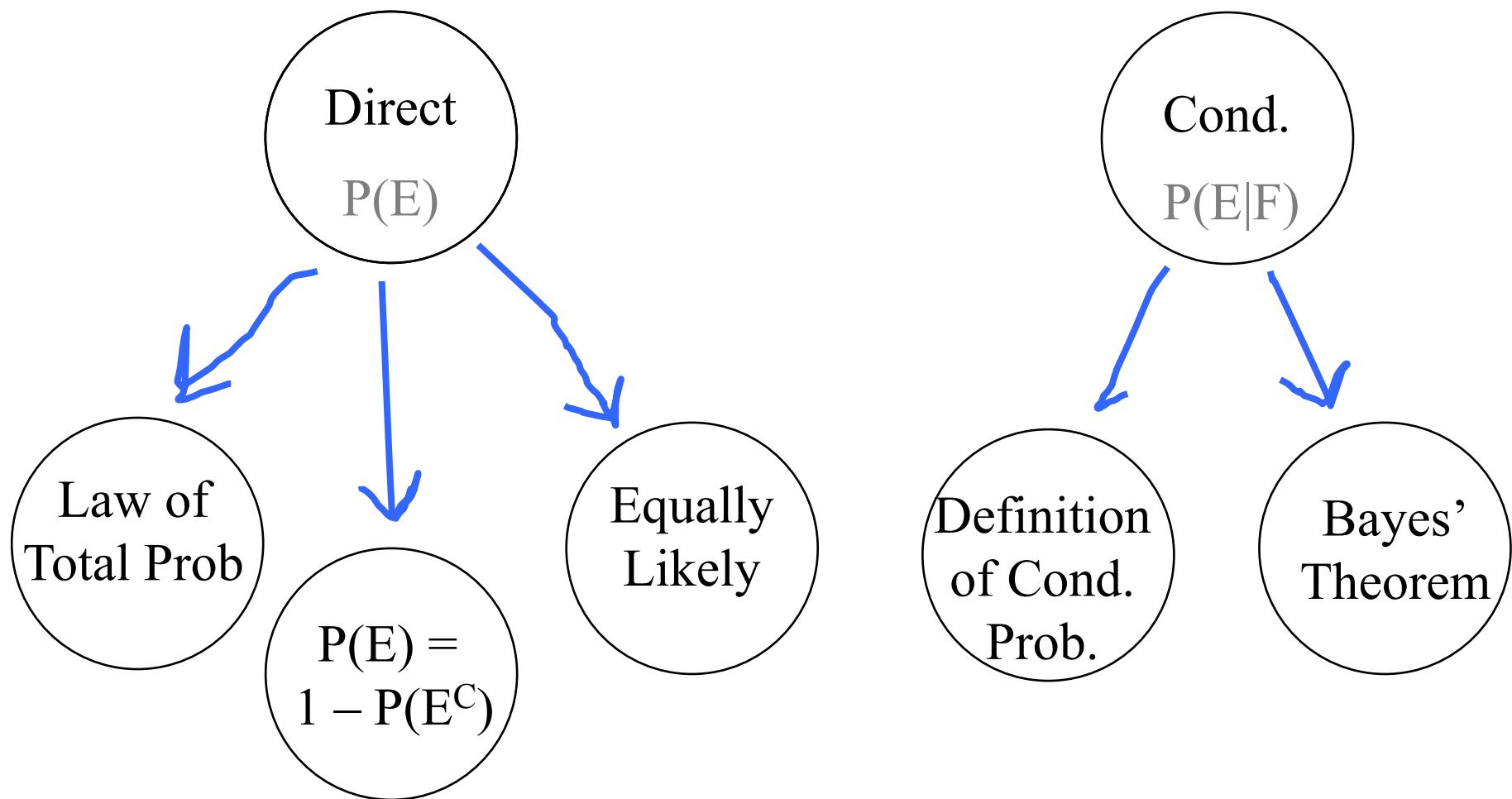
$$(E \cap F)^c = E^c \cup F^c$$



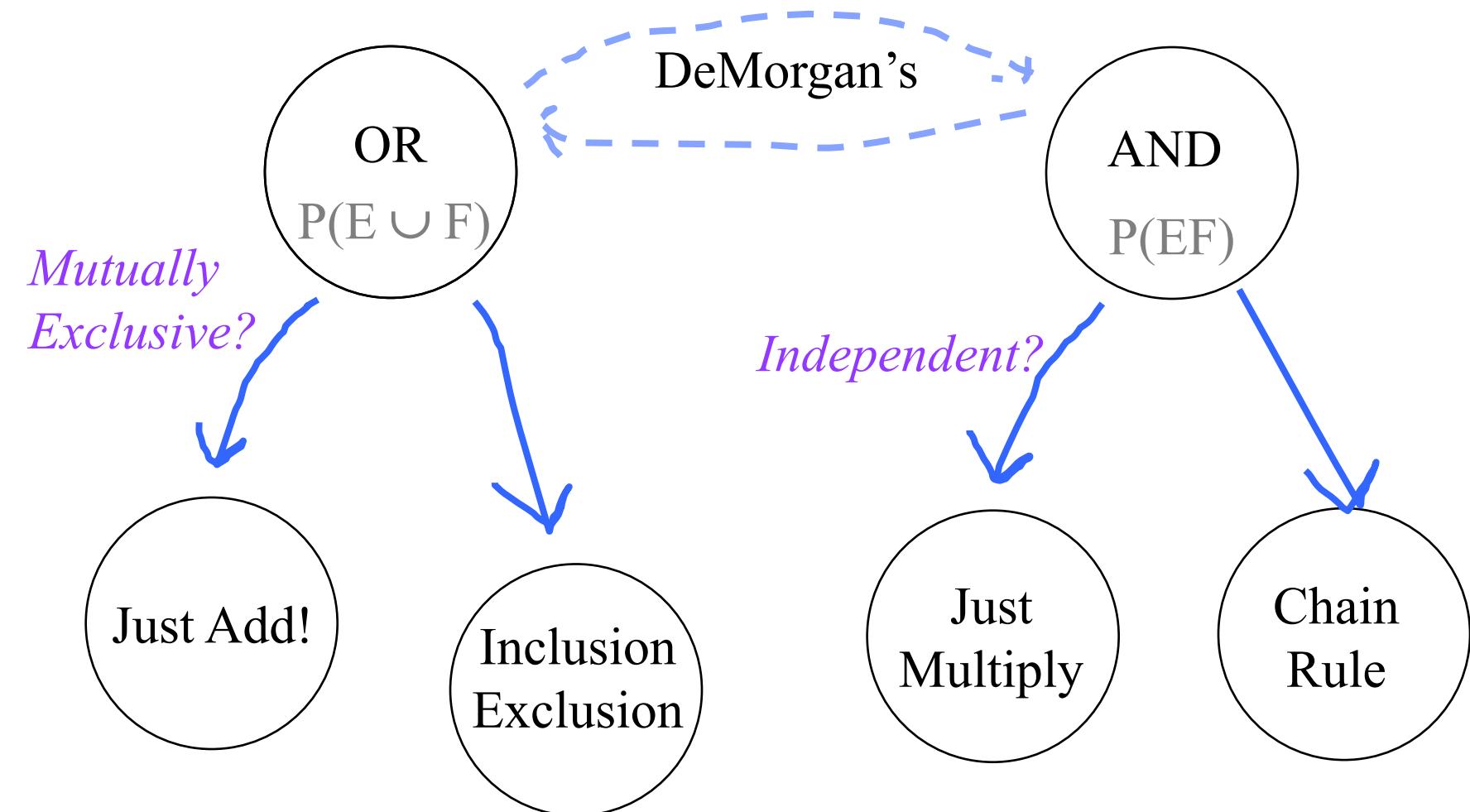
Core probability in two slides?



So Far

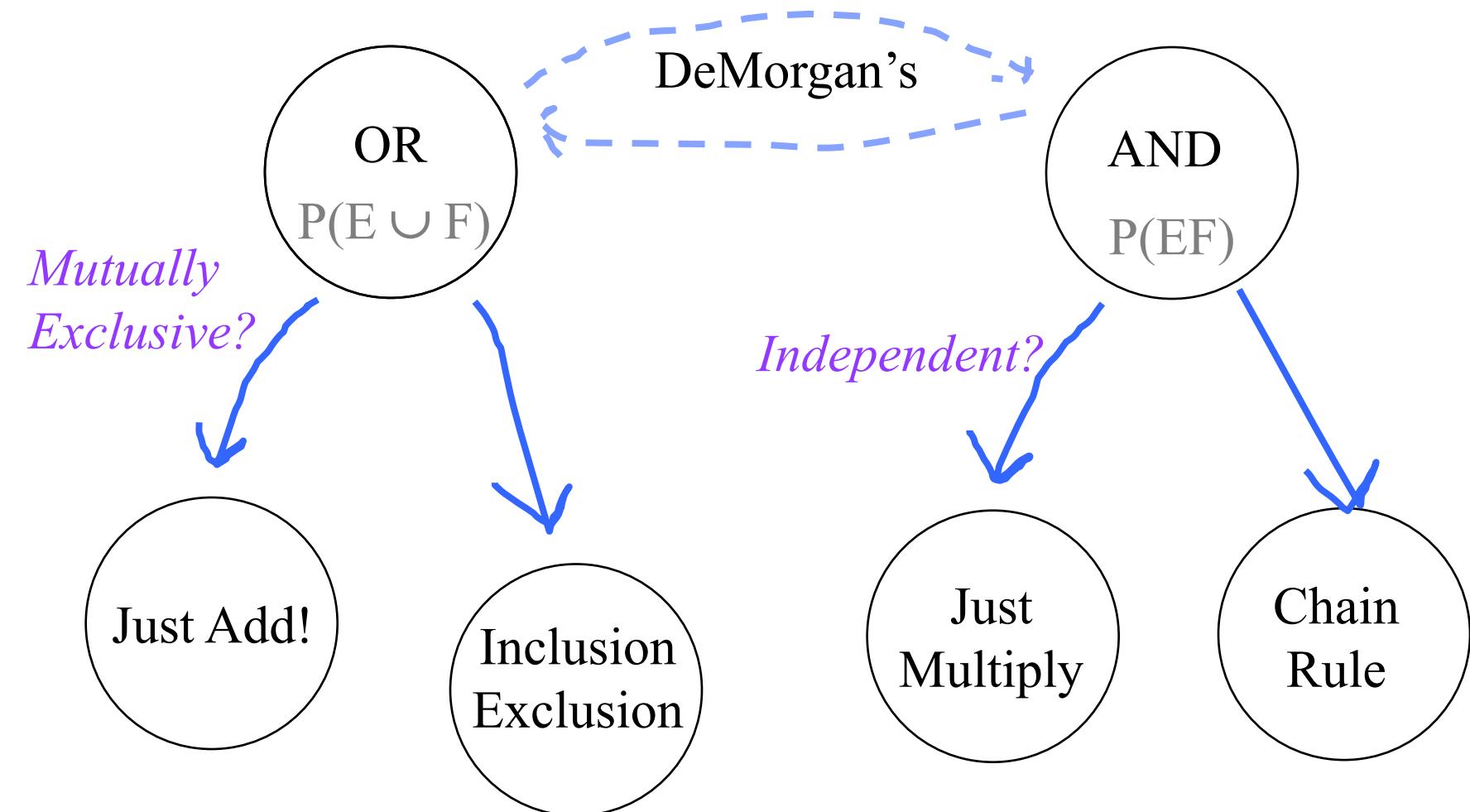


Today

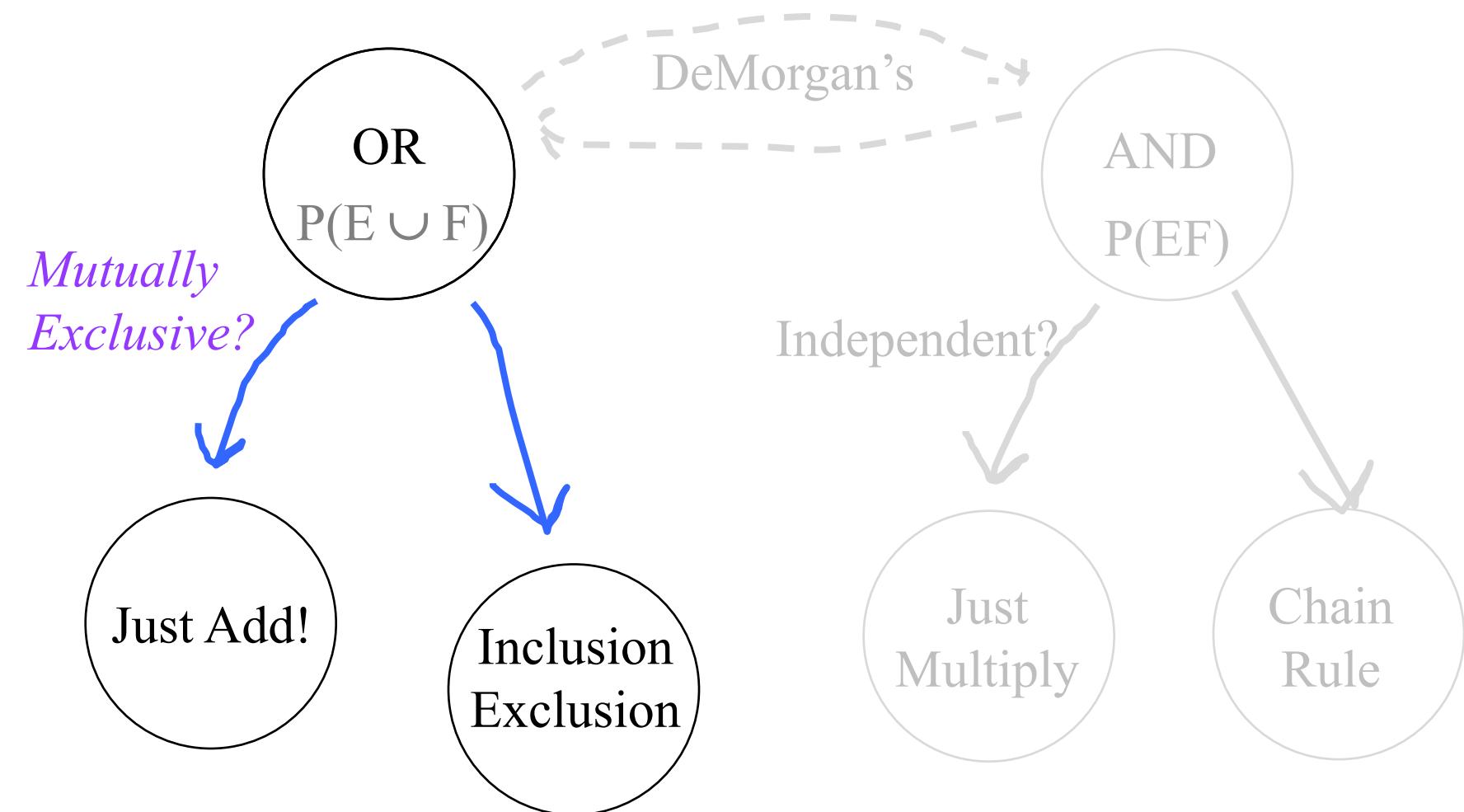


End Review

Today

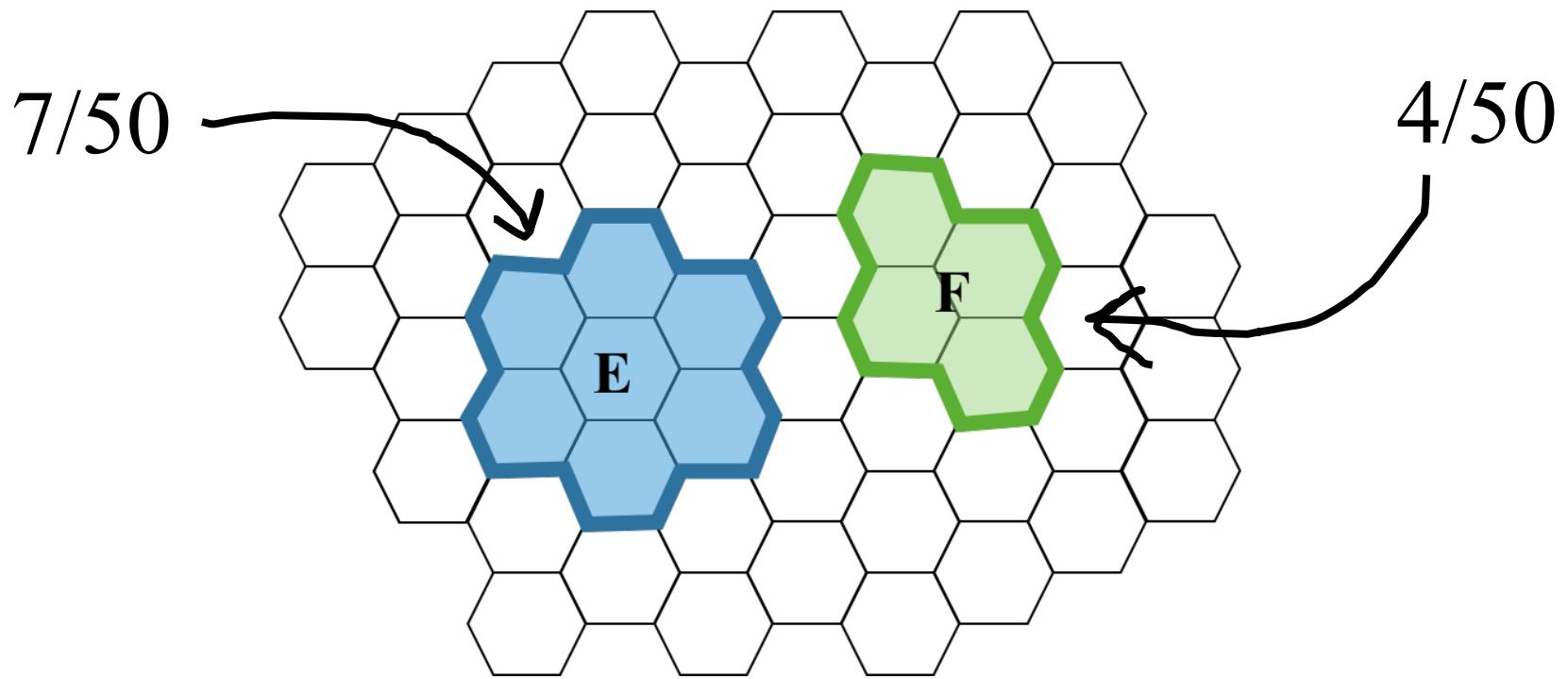


Today



Probability of “OR”

OR with Mutually Exclusive Events

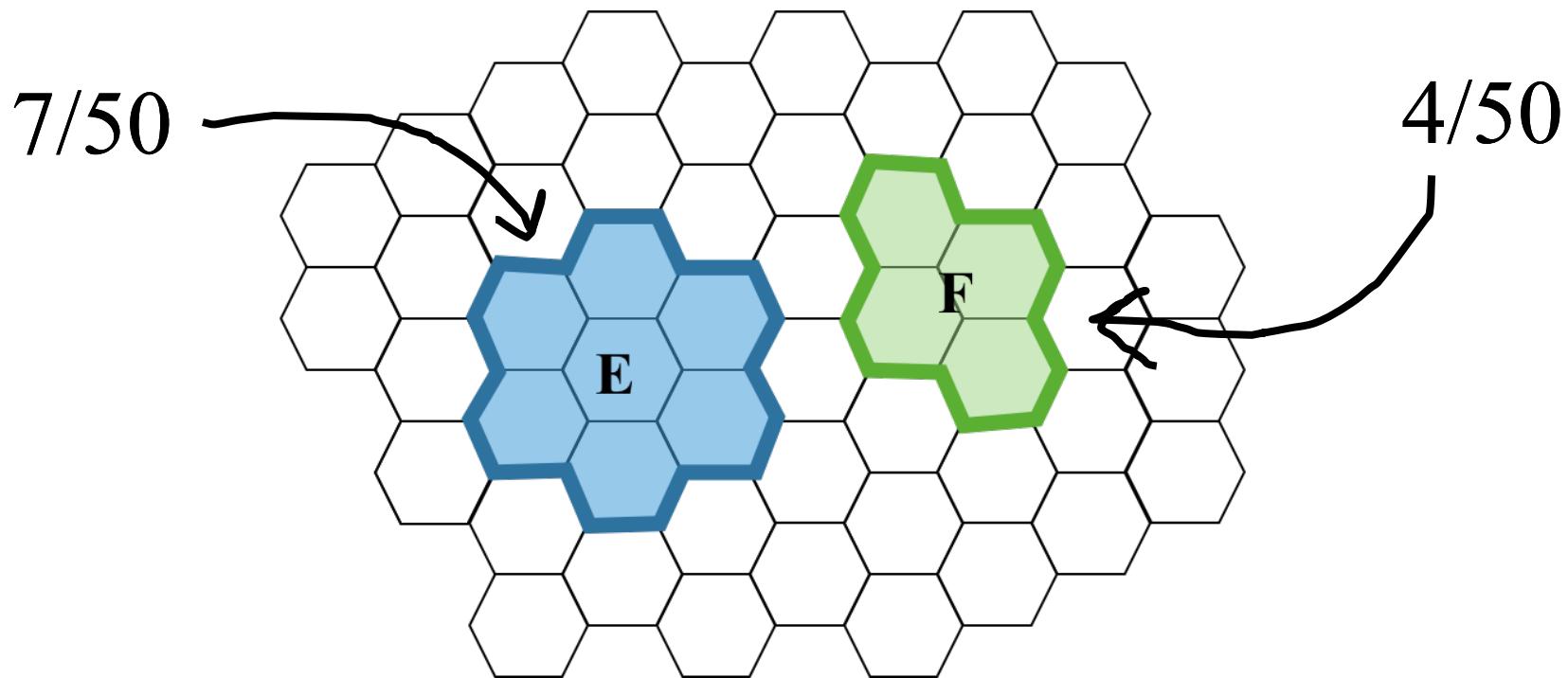


If events are mutually exclusive, probability of OR is simple:

$$P(E \cup F) = P(E) + P(F)$$



OR with Mutually Exclusive Events

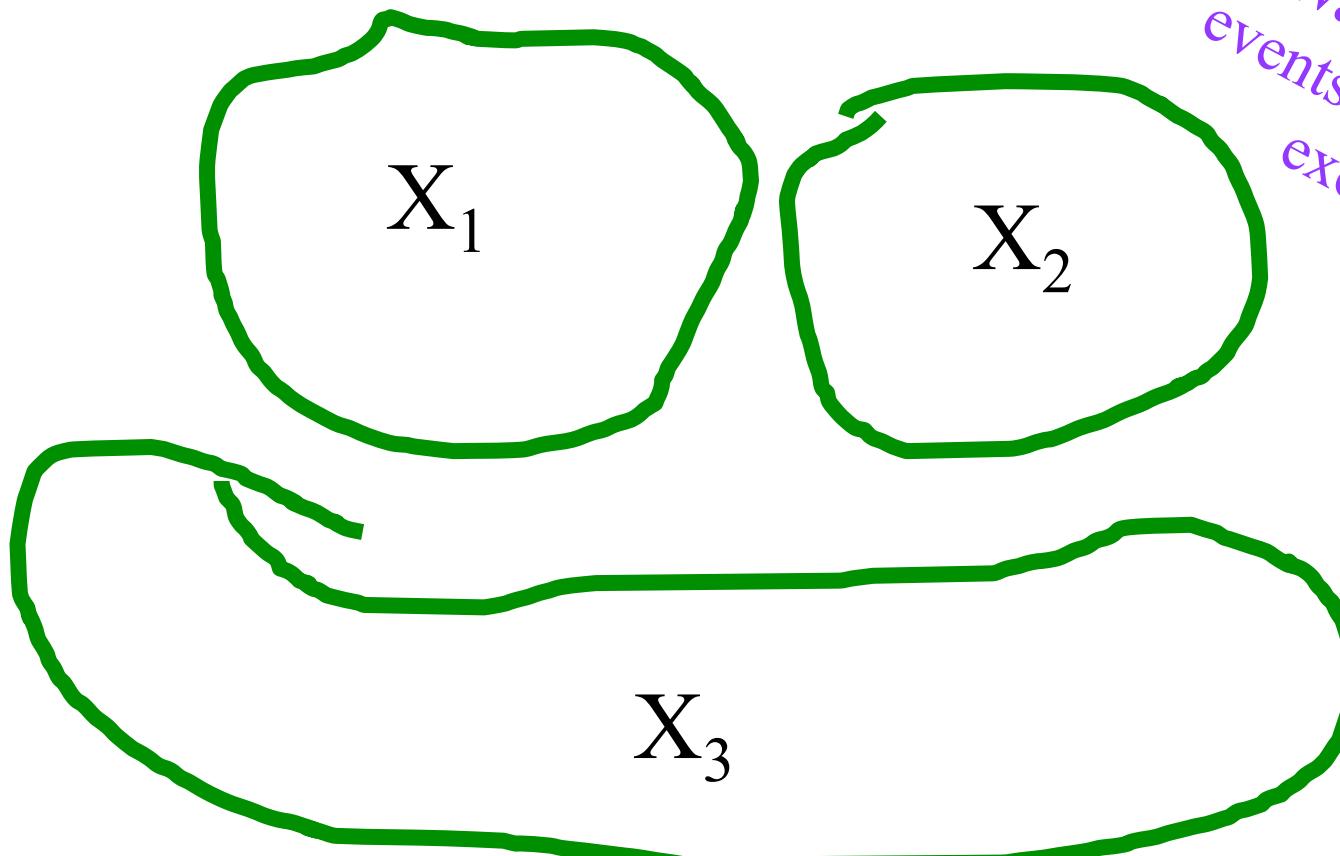


If events are mutually exclusive, probability of OR is simple:

$$P(E \cup F) = \frac{7}{50} + \frac{4}{50} = \frac{11}{50}$$



OR with Many Mutually Exclusive Events



*Wahoo! All my
events are mutually
exclusive*

$$P(X_1 \cup X_2 \cup \dots \cup X_n) = \sum_{i=1}^n P(X_i)$$



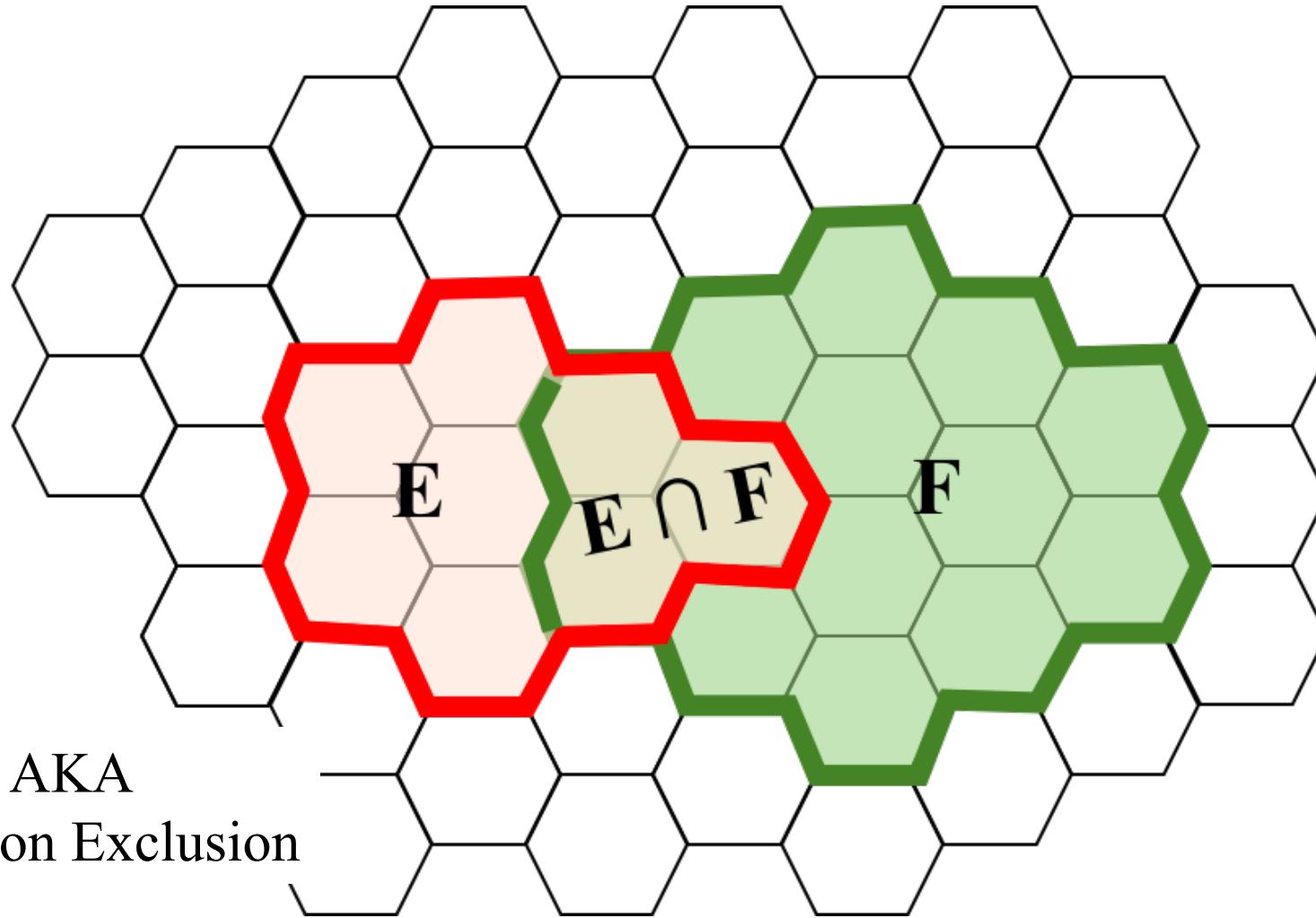


If events are *mutually exclusive* probability of OR is easy!



What about when they are not
Mutually exclusive?

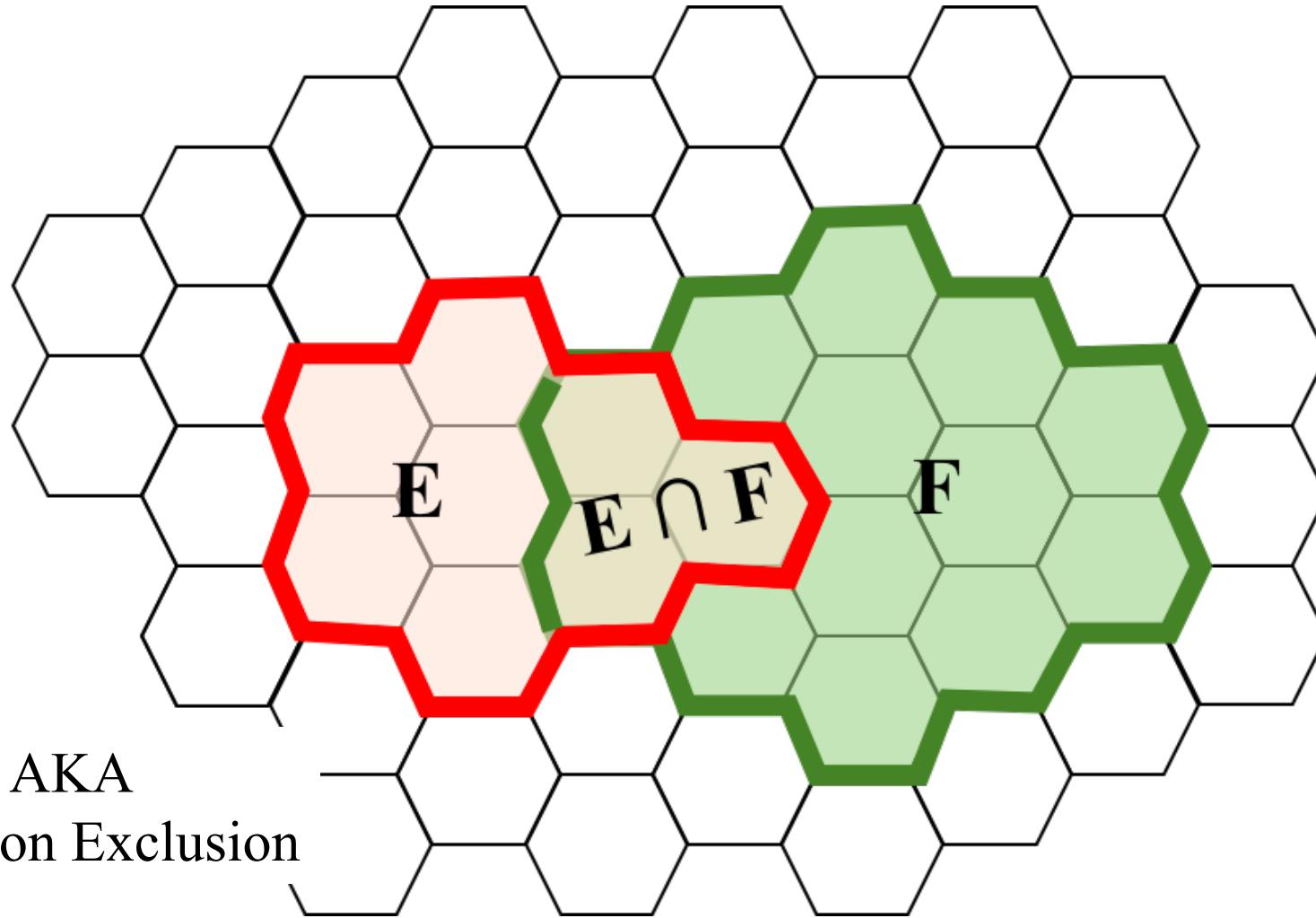
OR without Mutually Exclusivity



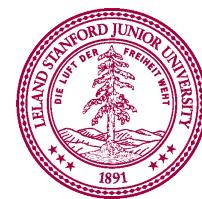
$$P(E \cup F) = P(E) + P(F) - P(EF)$$



OR without Mutually Exclusivity



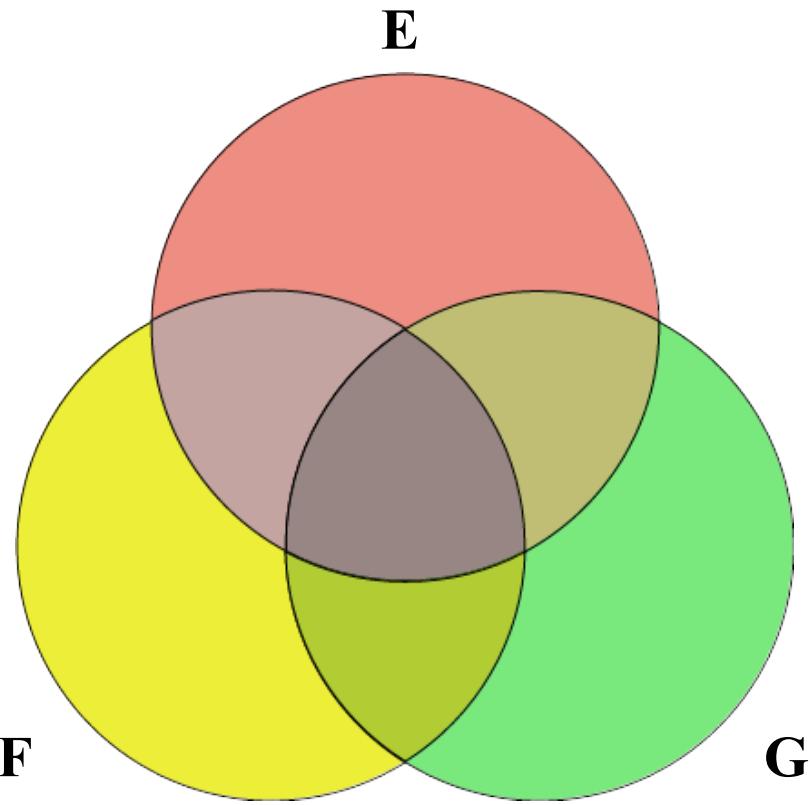
$$P(E \cup F) = \frac{8}{50} + \frac{14}{50} - \frac{3}{50} = \frac{19}{50}$$



More than two sets?

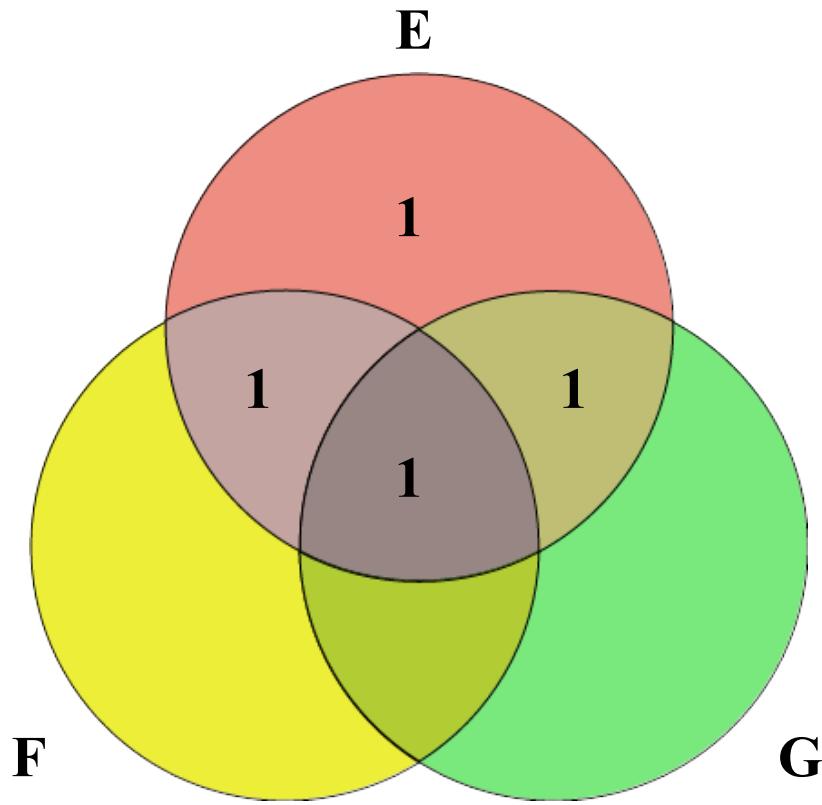
Inclusion Exclusion with Three Sets

$$P(E \cup F \cup G) =$$



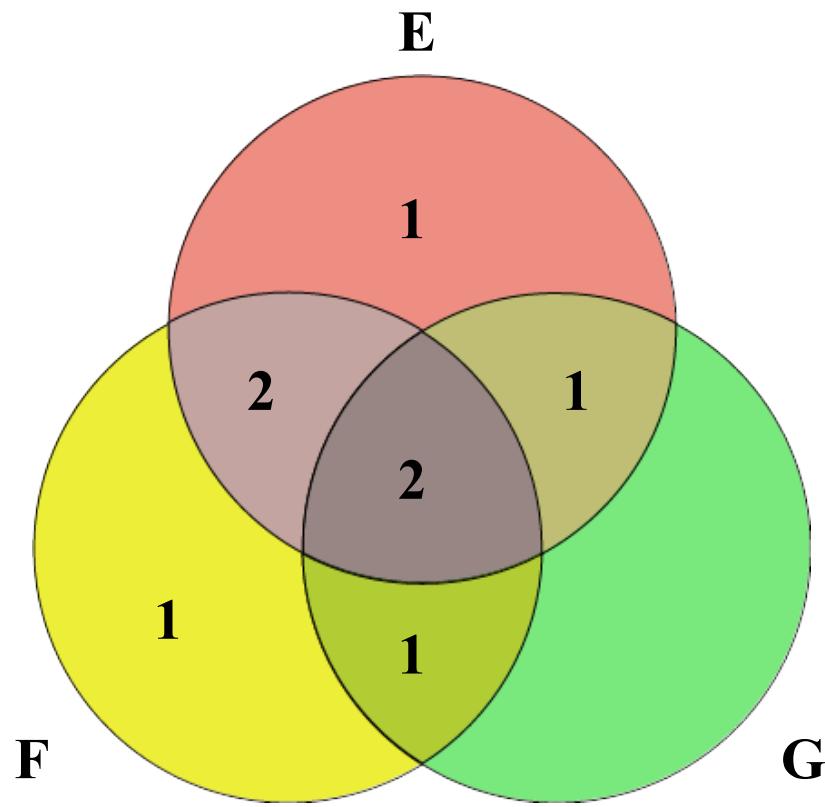
Inclusion Exclusion with Three Sets

$$P(E \cup F \cup G) = P(E)$$



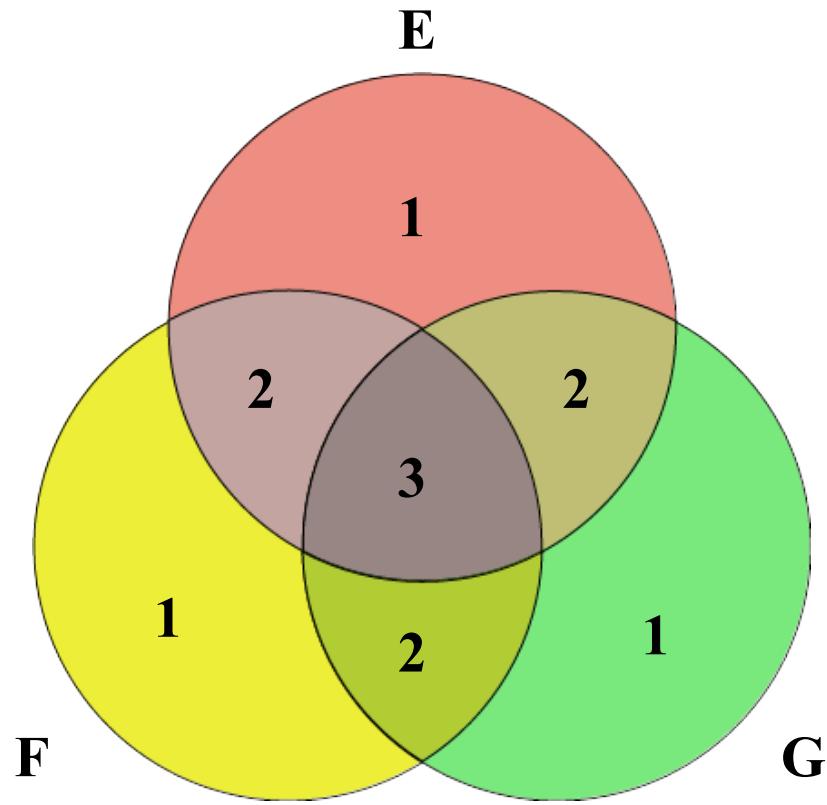
Inclusion Exclusion with Three Sets

$$P(E \cup F \cup G) = P(E) + P(F) + P(G) - P(E \cap F) - P(E \cap G) - P(F \cap G) + P(E \cap F \cap G)$$



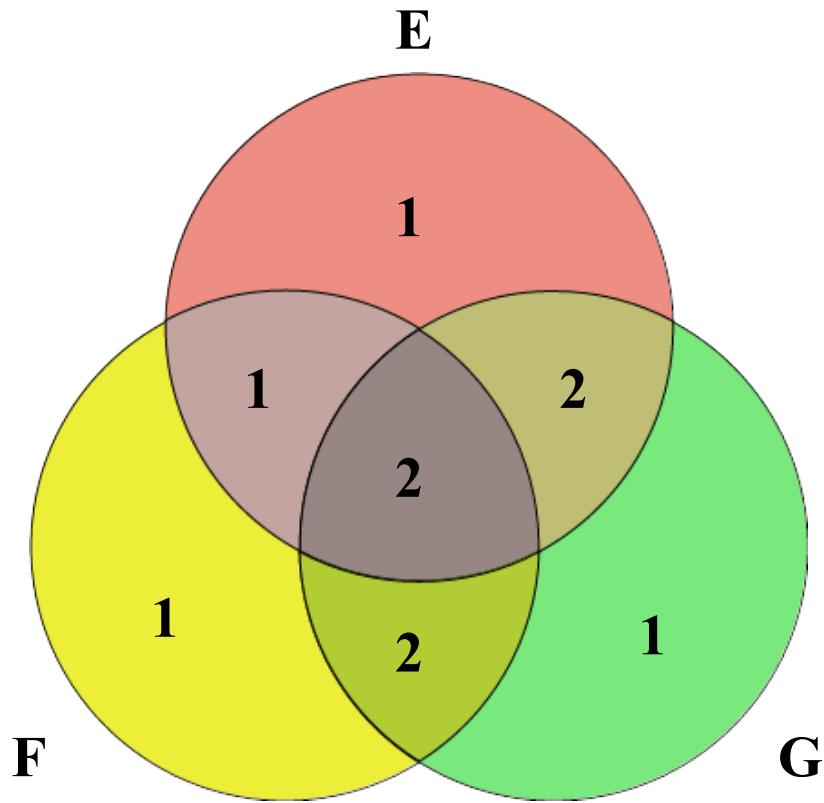
Inclusion Exclusion with Three Sets

$$P(E \cup F \cup G) = P(E) + P(F) + P(G)$$



Inclusion Exclusion with Three Sets

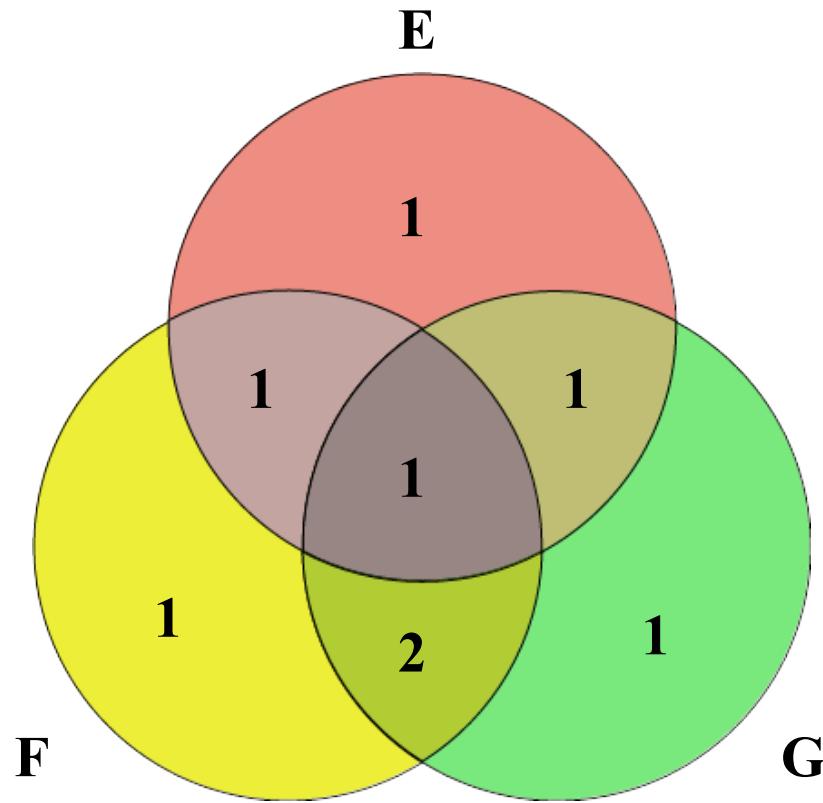
$$P(E \cup F \cup G) = P(E) + P(F) + P(G) - P(EF) - P(EG) - P(FG) + P(EFG)$$



Inclusion Exclusion with Three Sets

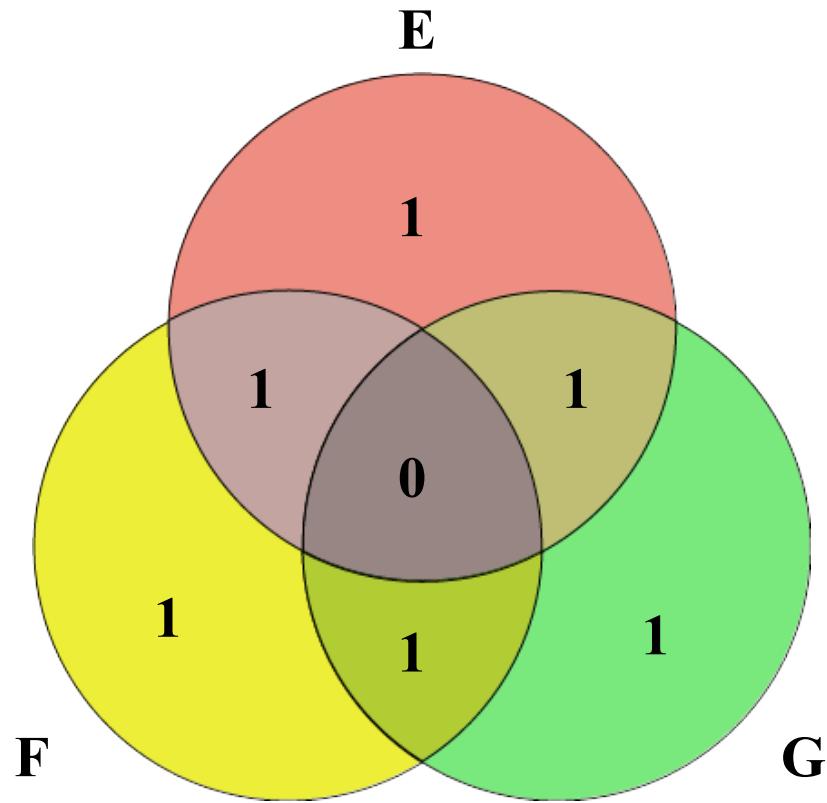
$$P(E \cup F \cup G) = P(E) + P(F) + P(G)$$

$$-P(EF) - P(EG)$$



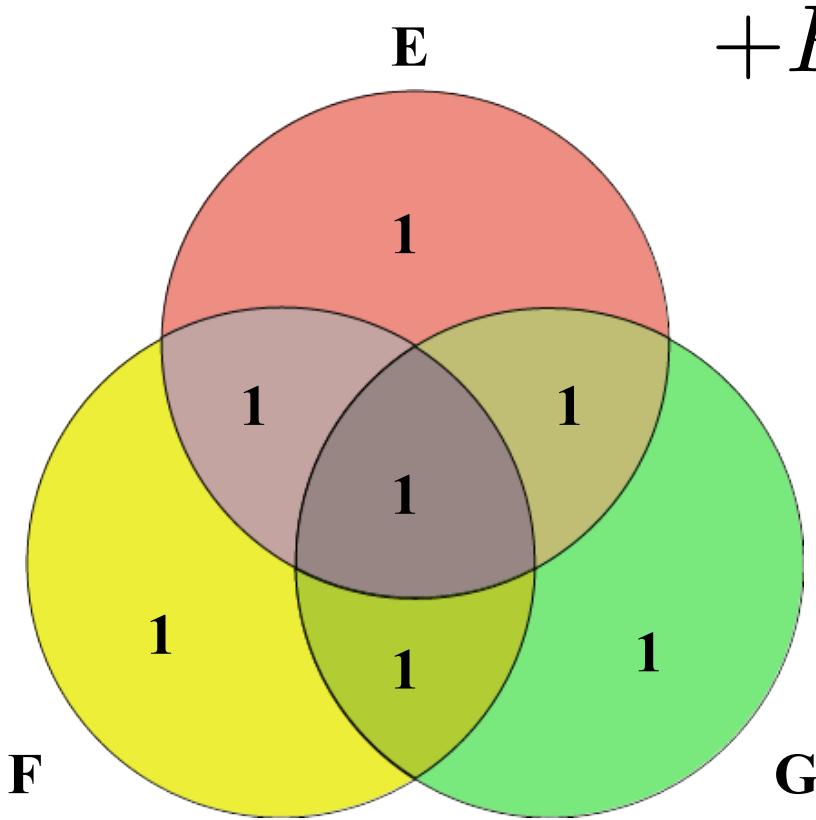
Inclusion Exclusion with Three Sets

$$\begin{aligned} P(E \cup F \cup G) &= P(E) + P(F) + P(G) \\ &\quad - P(EF) - P(EG) - P(FG) \end{aligned}$$



Inclusion Exclusion with Three Sets

$$\begin{aligned} P(E \cup F \cup G) &= P(E) + P(F) + P(G) \\ &\quad - P(EF) - P(EG) - P(FG) \\ &\quad + P(EFG) \end{aligned}$$



General Inclusion Exclusion

$$P(E_1 \cup E_2 \cup \dots \cup E_n) = \sum_{r=1}^n (-1)^{r+1} Y_r$$

* Where Y_r is the sum, for all combinations of r events, of the probability of the union those events.

Y_1 = Sum of all events on their own

$$\sum_i P(E_i)$$

Y_2 = Sum of all pairs of events

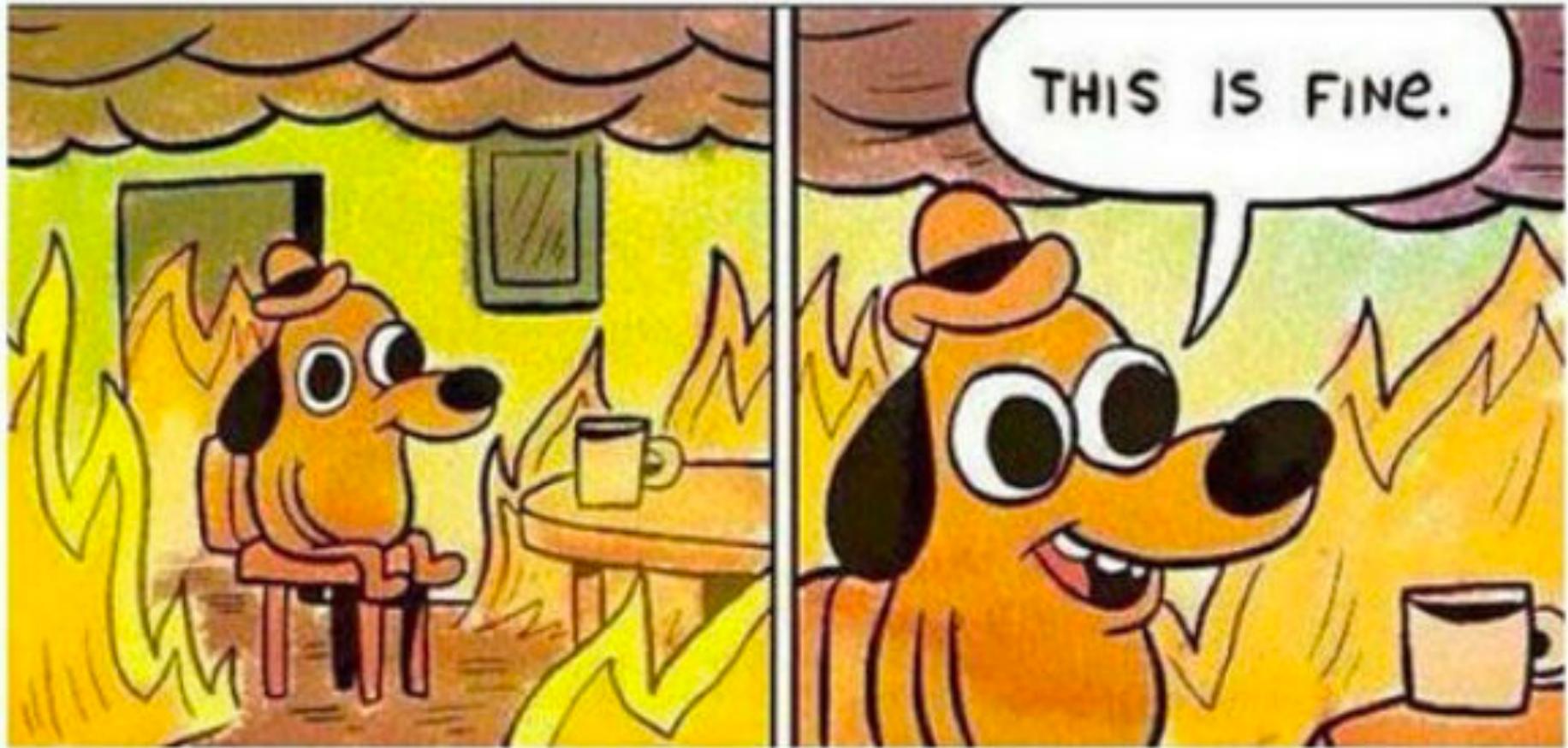
$$\sum_{i,j} P(E_i \cap E_j) \quad \text{s.t. } i \neq j$$

Y_3 = Sum of all triples of events

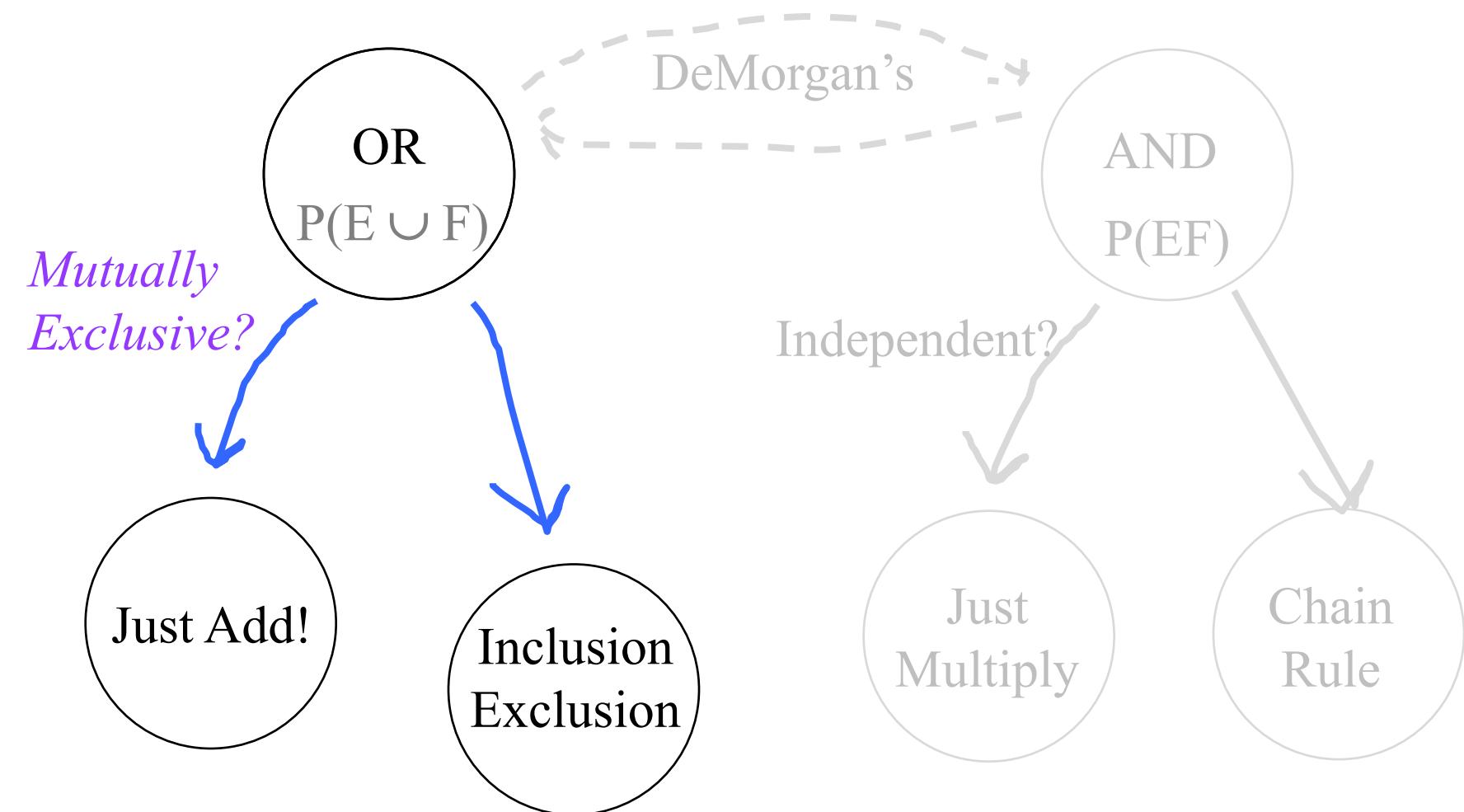
$$\sum_{i,j,k} P(E_i \cap E_j \cap E_k) \quad \text{s.t. } i \neq j, j \neq k, i \neq k$$



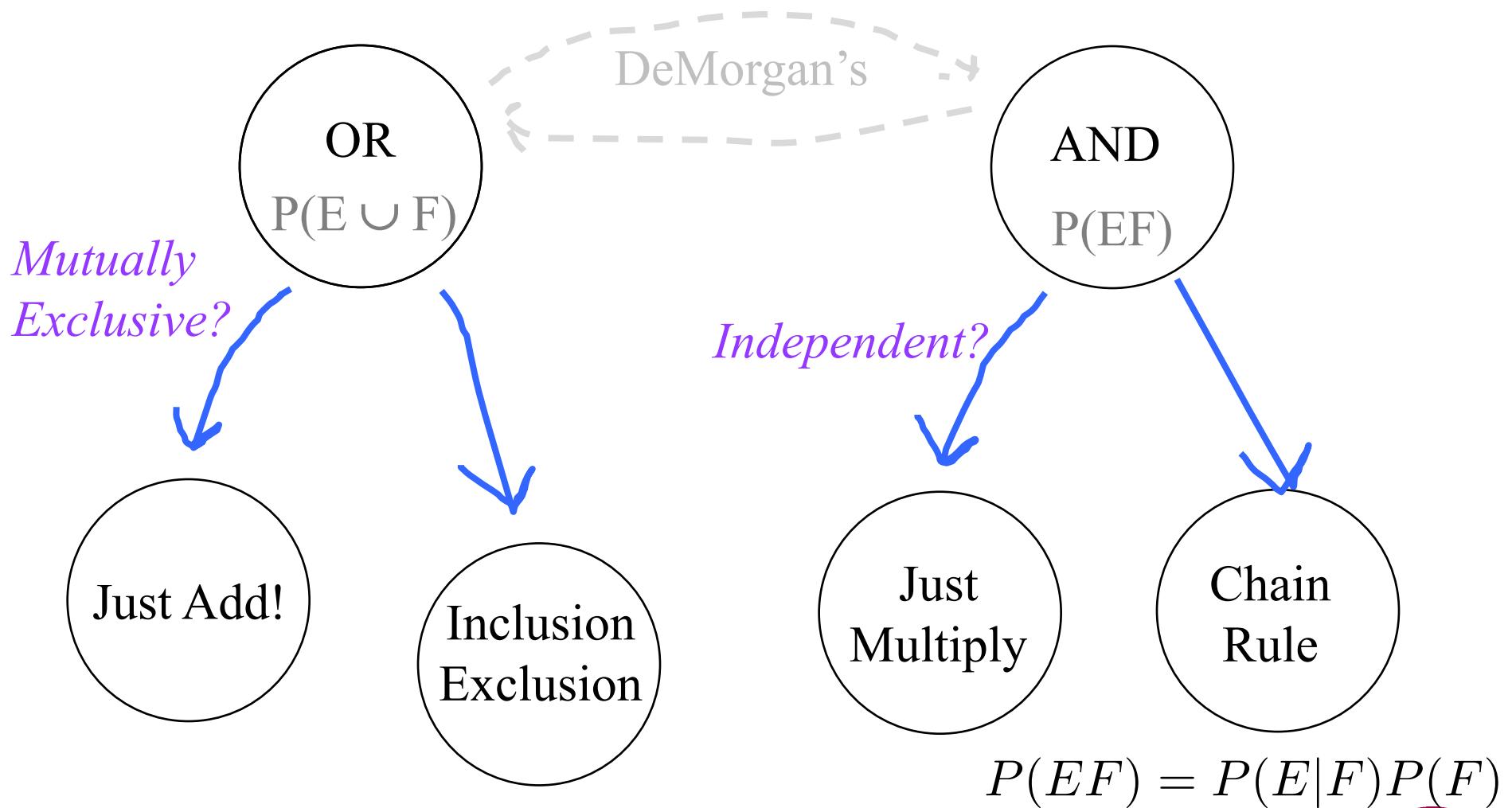
General Inclusion Exclusion



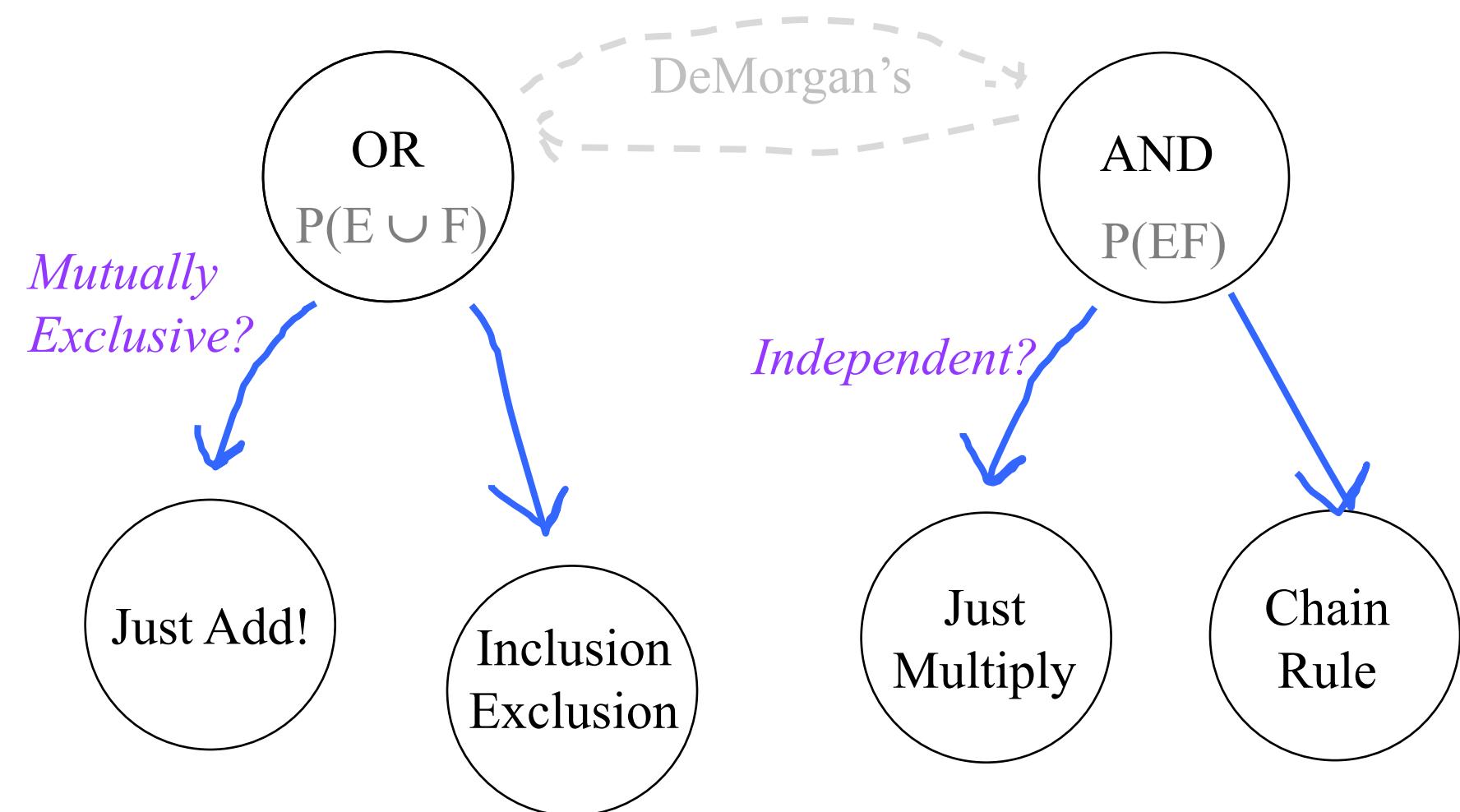
Today



Today



Today



Probability of “AND”

We the People

insure domestic Tranquility, provide for the common defense, and our Posterity, do ordain and establish this Constitution.

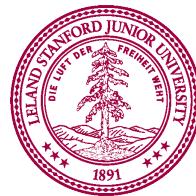
Section 1. All legislative Powers herein granted shall be vested in a Congress of the United States, which shall consist of a Senate and a House of Representatives.

Independence

Two events A and B are called independent if:

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

Otherwise, they are called dependent events





If events are *independent*
probability of AND is easy!

*You will need to use this “trick” with high probability



Intuition through proofs

Let A and B be independent

$$P(A|B) = \frac{P(AB)}{P(B)}$$

Definition of
conditional probability

$$= \frac{P(A)P(B)}{P(B)}$$

Since A and B are
independent

$$= P(A)$$

Taking the bus to
cancel city

Knowing that event B happened, doesn't change
our belief that A will happen.



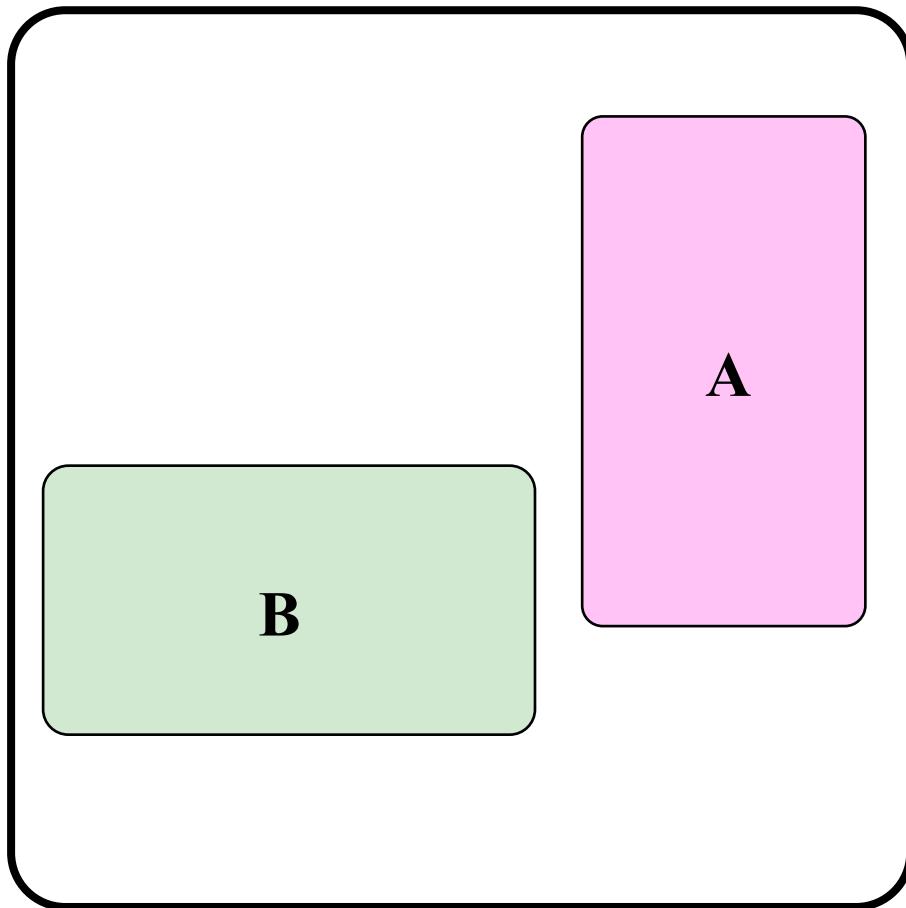
Dice, Our Misunderstood Friends

- Roll two 6-sided dice, yielding values D_1 and D_2
 - Let E be event: $D_1 = 1$
 - Let F be event: $D_2 = 1$
- What is $P(E)$, $P(F)$, and $P(EF)$?
 - $P(E) = 1/6$, $P(F) = 1/6$, $P(EF) = 1/36$
 - $P(EF) = P(E) P(F)$ \rightarrow E and F independent
- Let G be event: $D_1 + D_2 = 5$ $\{(1, 4), (2, 3), (3, 2), (4, 1)\}$
- What is $P(E)$, $P(G)$, and $P(EG)$?
 - $P(E) = 1/6$, $P(G) = 4/36 = 1/9$, $P(EG) = 1/36$
 - $P(EG) \neq P(E) P(G)$ \rightarrow E and G dependent



What does independence look like?

Independence?

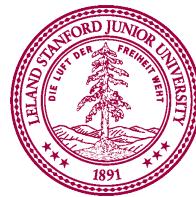


Independence Definition 1:

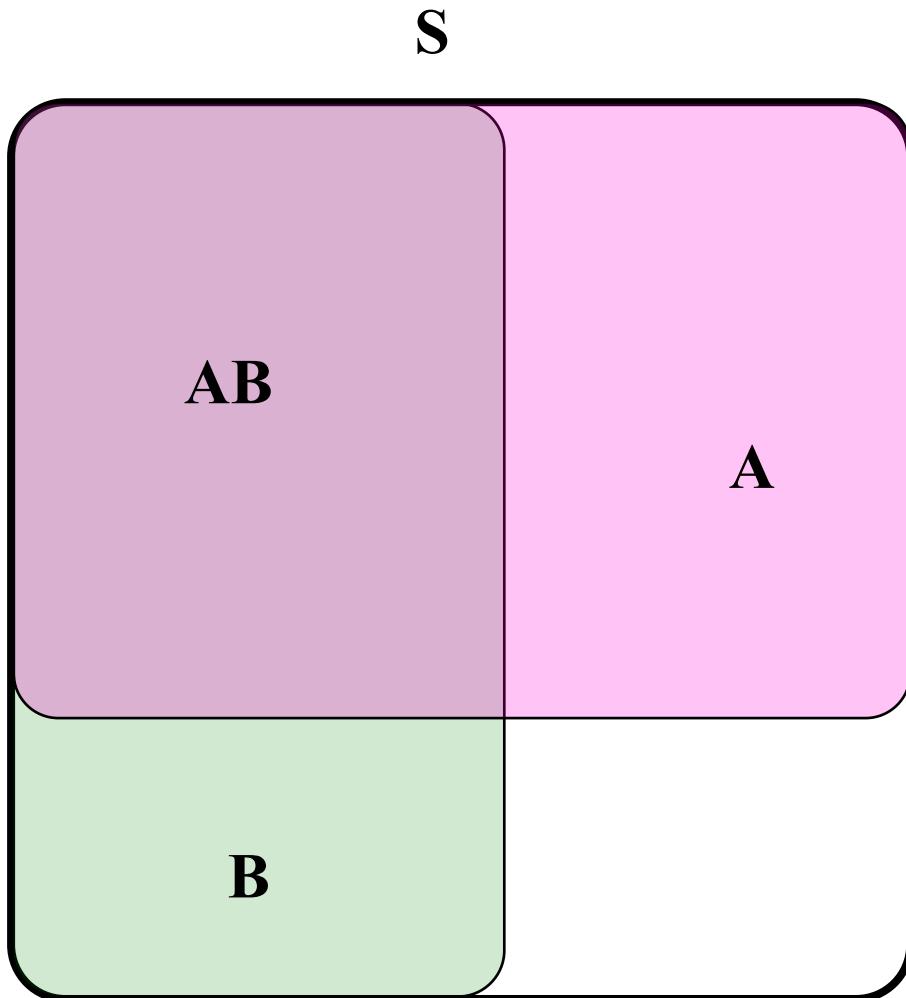
$$P(AB) = P(A)P(B)$$

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

0



Independence



Independence Definition 1:

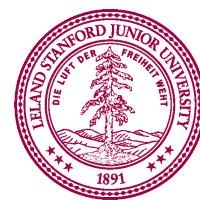
$$P(AB) = P(A)P(B)$$

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

Independence Definition 2:

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

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



More Intuition through proofs:

Independence

Given independent events A and B, prove that A and B^C are independent

We want to show that $P(AB^C) = P(A)P(B^C)$

$$\begin{aligned} P(AB^C) &= P(A) - P(AB) && \text{By Intersection Rule} \\ &= P(A) - P(A)P(B) && \text{By independence} \\ &= P(A)[1 - P(B)] && \text{Factoring} \\ &= P(A)P(B^C) && \text{Since } P(B) + P(B^C) = 1 \end{aligned}$$

So if A and B are independent A and B^C are also independent

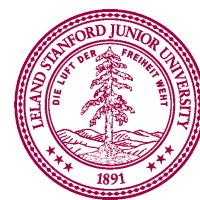


Generalization

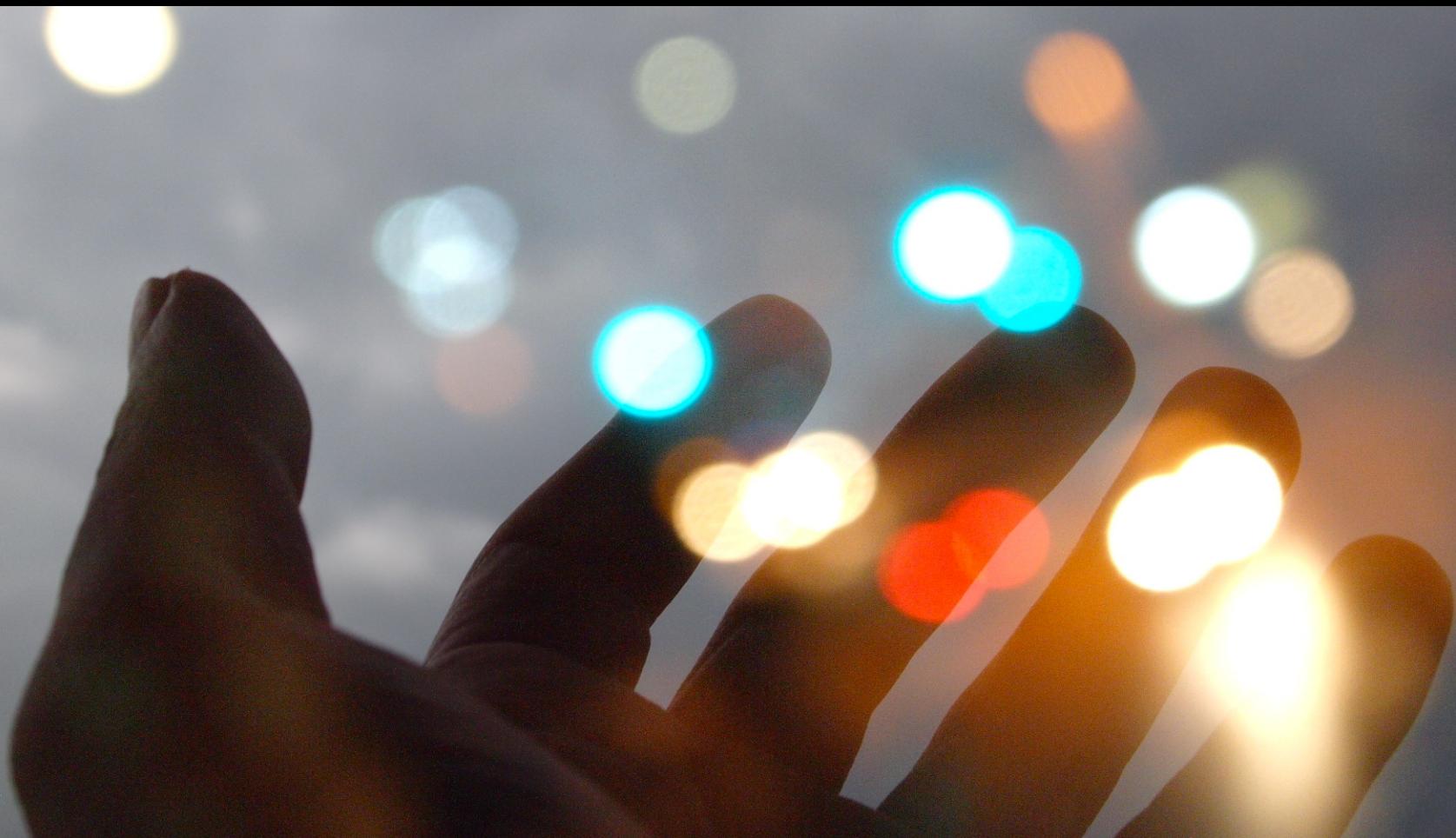


Generalized Independence

- General definition of Independence:
Events E_1, E_2, \dots, E_n are independent if for every subset with r elements (where $r \leq n$) it holds that:
$$P(E_1, E_2, E_3, \dots, E_r) = P(E_1)P(E_2)P(E_3)\dots P(E_r)$$
- Example: outcomes of n separate flips of a coin are all independent of one another
 - Each flip in this case is called a “trial” of the experiment



Math > Intuition



Two Dice

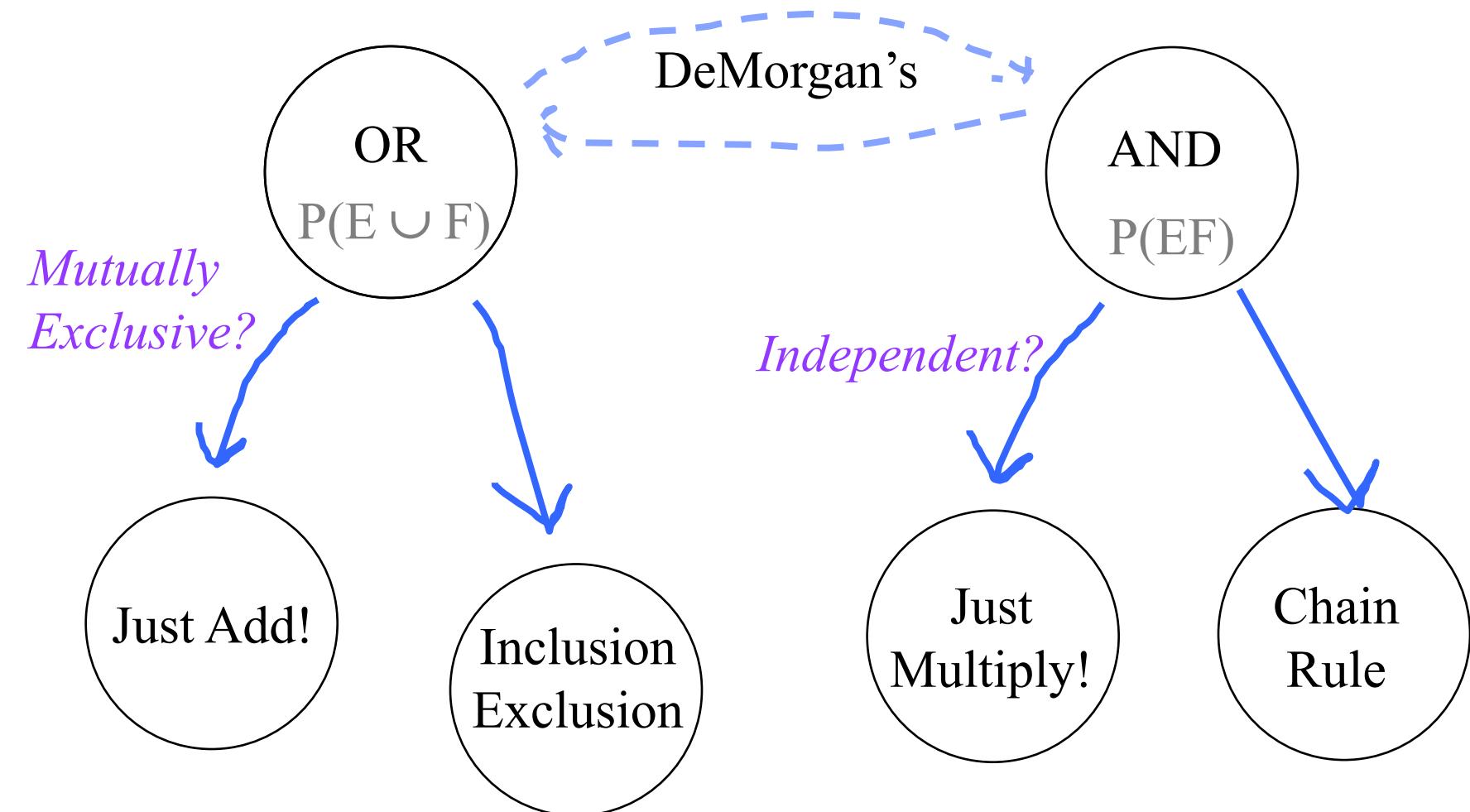
- Roll two 6-sided dice, yielding values D_1 and D_2
 - Let E be event: $D_1 = 1$
 - Let F be event: $D_2 = 6$
 - Are E and F independent? Yes!
- Let G be event: $D_1 + D_2 = 7$
 - Are E and G independent? Yes!
 - $P(E) = 1/6, P(G) = 1/6, P(E \cap G) = 1/36$ [roll (1, 6)]
 - Are F and G independent? Yes!
 - $P(F) = 1/6, P(G) = 1/6, P(F \cap G) = 1/36$ [roll (1, 6)]
 - Are E, F and G independent? No!
 - $P(E \cap F \cap G) = 1/36 \neq 1/216 = (1/6)(1/6)(1/6)$



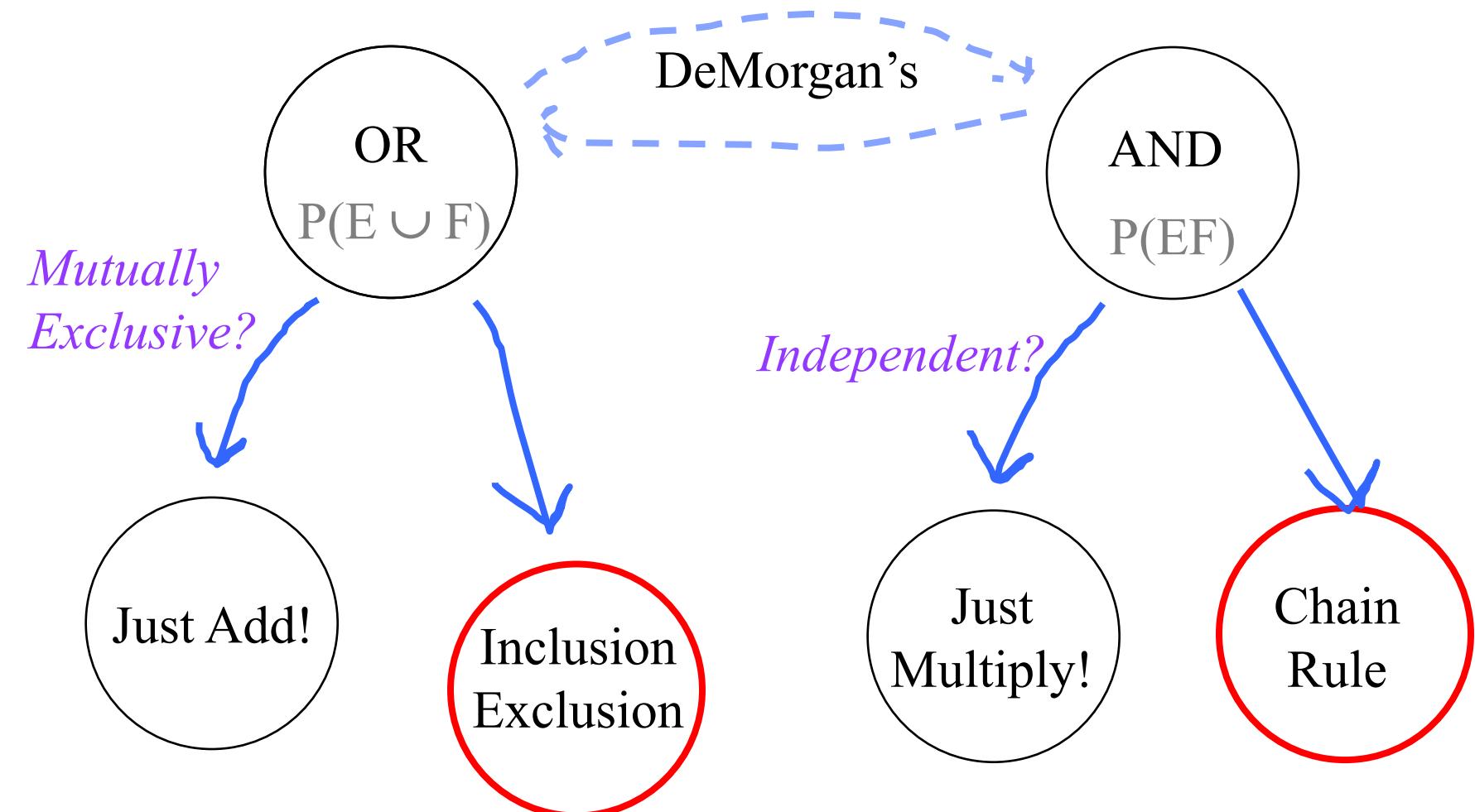
New Ability



Today



Today



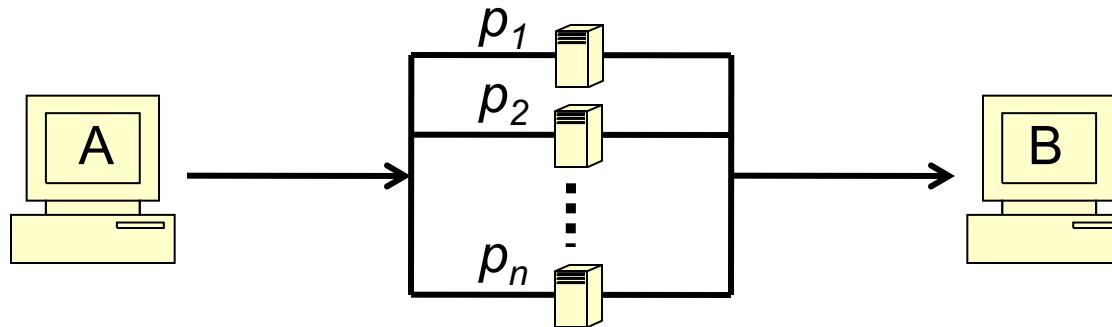


Use the two properties
(mutual exclusion and
independence)



Sending a Message Through Network

- 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 $P(E)$?
- Solution:
 - $$\begin{aligned} P(E) &= 1 - P(\text{all routers fail}) \\ &= 1 - (1 - p_1)(1 - p_2)\dots(1 - p_n) \\ &= 1 - \prod_{i=1}^n (1 - p_i) \end{aligned}$$



Coin Flips

- Say a coin comes up heads with probability p
 - Each coin flip is an independent trial
- $P(n \text{ heads on } n \text{ coin flips}) = p^n$
- $P(n \text{ tails on } n \text{ coin flips}) = (1 - p)^n$
- $P(\text{first } k \text{ heads, then } n - k \text{ tails}) = p^k (1 - p)^{n-k}$
- $P(\text{exactly } k \text{ heads on } n \text{ coin flips}) = ?$



Explain...

P(exactly k heads on n coin flips)?

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

Think of the flips as ordered:

Ordering 1: T, H, H, T, T, T....

The coin flips are independent!

Ordering 2: H, T, H, T, T, T....

And so on...

$$P(F_i) = p^k (1-p)^{n-k}$$

Let's make each ordering with k heads an event... F_i

P(exactly k heads on n coin flips) = P(any one of the events)

P(exactly k heads on n coin flips) = P(F_1 or F_2 or F_3 ...)

Those events are mutually exclusive!



Hash Tables

- m strings are hashed (unequally) into a hash table with n buckets
 - Each string hashed is an independent trial, with probability p_i of getting hashed to bucket i
 - $E = \text{at least one string hashed to first bucket}$
 - What is $P(E)$?
- Solution



To the white board!

Yet More Hash Tables

- m strings are hashed (unequally) into a hash table with n buckets
 - Each string hashed is an independent trial, with probability p_i of getting hashed to bucket i
 - $E = \text{At least 1 of}$ buckets 1 to k has ≥ 1 string hashed to it
- Solution
 - $F_i = \text{at least one string hashed into } i\text{-th bucket}$
 - $P(E) = P(F_1 \cup F_2 \cup \dots \cup F_k) = 1 - P((F_1 \cup F_2 \cup \dots \cup F_k)^c)$
 $= 1 - P(F_1^c F_2^c \dots F_k^c)$ (DeMorgan's Law)
 - $P(F_1^c F_2^c \dots F_k^c) = P(\text{no strings hashed to buckets 1 to } k)$
 $= (1 - p_1 - p_2 - \dots - p_k)^m$
 - $P(E) = 1 - (1 - p_1 - p_2 - \dots - p_k)^m$



No, Really, More Hash Tables

- m strings are hashed (unequally) into a hash table with n buckets
 - Each string hashed is an independent trial, with probability p_i of getting hashed to bucket i
 - $E = \text{Each of}$ buckets 1 to k has ≥ 1 string hashed to it
 - Solution
 - $F_i = \text{at least one string hashed into } i\text{-th bucket}$
 - $P(E) = P(F_1 F_2 \dots F_k) = 1 - P((F_1 F_2 \dots F_k)^c)$
 $= 1 - P(F_1^c \cup F_2^c \cup \dots \cup F_k^c)$ (DeMorgan's Law)
 $= 1 - P\left(\bigcup_{i=1}^k F_i^c\right) = 1 - \sum_{r=1}^k (-1)^{(r+1)} \sum_{i_1 < \dots < i_r} P(F_{i_1}^c F_{i_2}^c \dots F_{i_r}^c)$
- where $P(F_{i_1}^c F_{i_2}^c \dots F_{i_r}^c) = (1 - p_{i_1} - p_{i_2} - \dots - p_{i_r})^m$



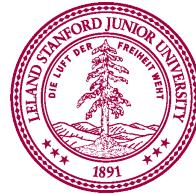
Phew!

Conditional Paradigm

- Recall:

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

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



Conditional Paradigm

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

$$P(A \cap B | E) = P(B | A \cap E)$$

$$P(A \cap B | E) = P(A | B \cap E) P(B | E)$$

- Can think of E as “everything you already know”
- Formally, $P(\bullet | E)$ satisfies 3 axioms of probability



BAE's Theorem?

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

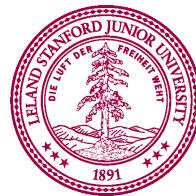


Conditional Independence

- Two events E and F are called conditionally independent given G, if

$$P(E \cap F | G) = P(E | G) P(F | G)$$

Or, equivalently: $P(E | F \cap G) = P(E | G)$



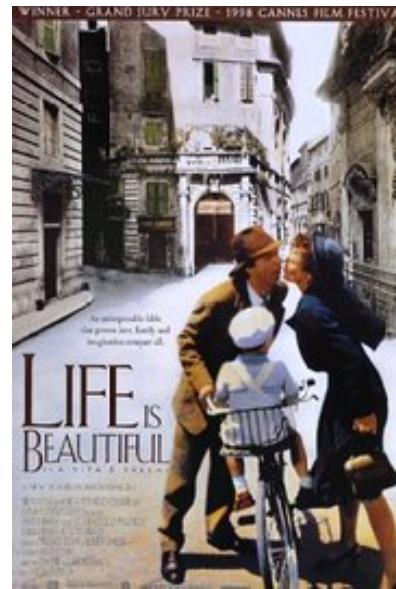
NETFLIX

And Learn

Netflix and Learn

What is the probability
that a user will watch
Life is Beautiful?

$$P(E)$$



$$P(E) = \lim_{n \rightarrow \infty} \frac{n(E)}{n} \approx \frac{\#\text{people who watched movie}}{\#\text{people on Netflix}}$$

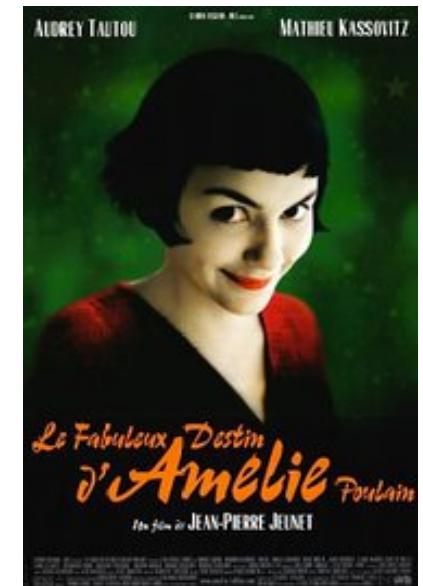
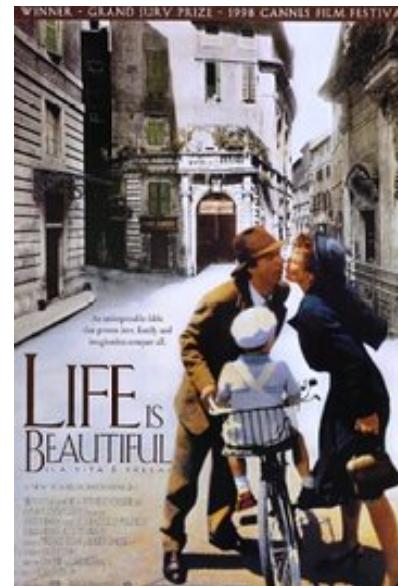
$$P(E) = 10,234,231 / 50,923,123 = 0.20$$



Netflix and Learn

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

$$P(E|F)$$



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

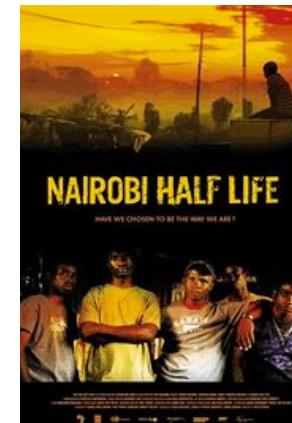
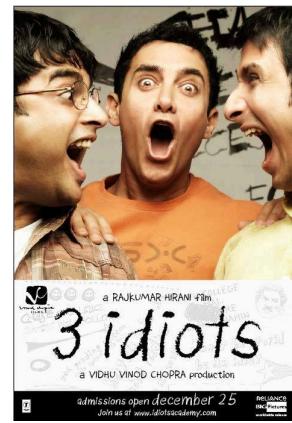
$$P(E|F) = 0.42$$



Conditioned on liking a set of movies?

Netflix and Learn

Each event corresponds to liking a particular movie



E_1

E_2

E_3

E_4

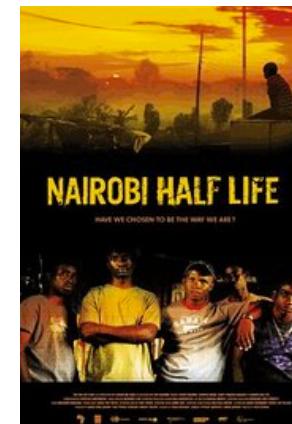
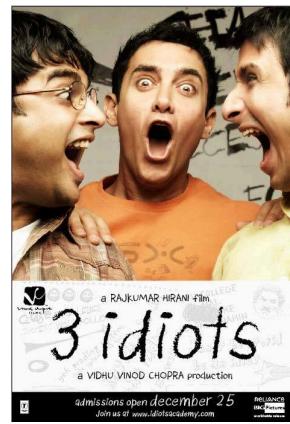
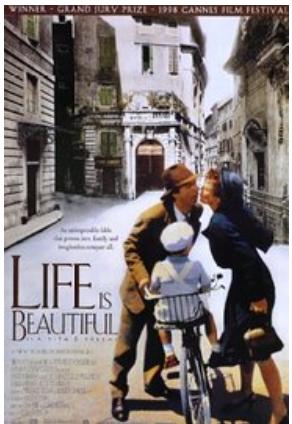
$$P(E_4 | E_1, E_2, E_3) ?$$



Is E_4 independent of E_1, E_2, E_3 ?

Netflix and Learn

Is E_4 independent of E_1, E_2, E_3 ?



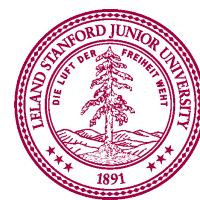
E_1

E_2

E_3

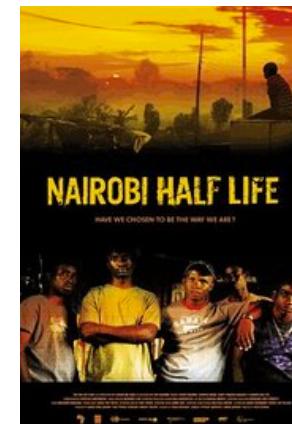
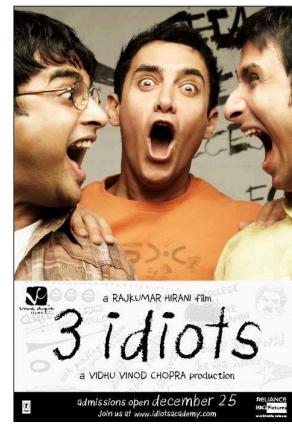
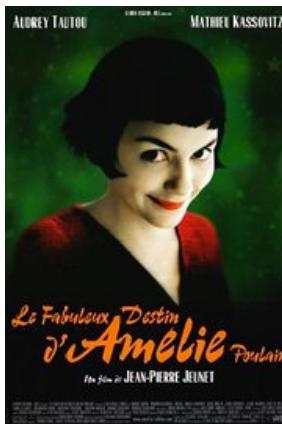
E_4

$$P(E_4 | E_1, E_2, E_3) \stackrel{?}{=} P(E_4)$$



Netflix and Learn

Is E_4 independent of E_1, E_2, E_3 ?



E_1

E_2

E_3

E_4

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



Netflix and Learn

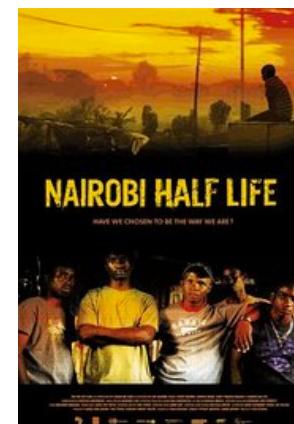
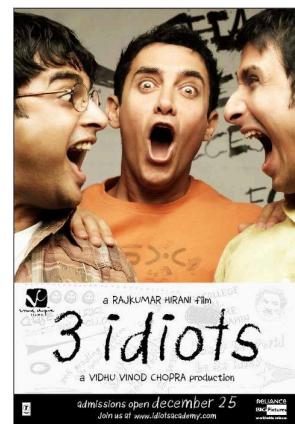
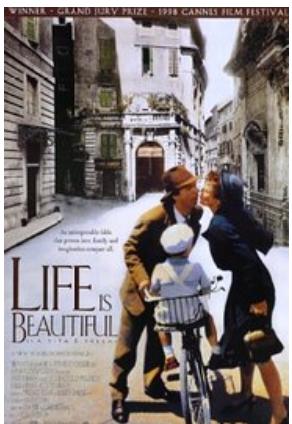
- What is the probability that a user watched four particular movies?
 - There are 13,000 titles on Netflix
 - The user watches 30 random titles
 - E = movies watched include the given four.
- Solution:

$$P(E) = \frac{\binom{4}{4} \binom{12996}{24}}{\binom{13000}{30}} = 10^{-11}$$

Watch those four *Choose 24 movies
not in the set*
*Choose 30 movies
from netflix*



Netflix and Learn



E_1

E_2

E_3

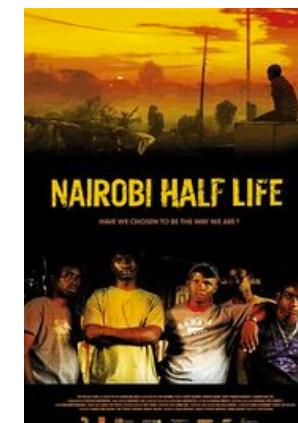
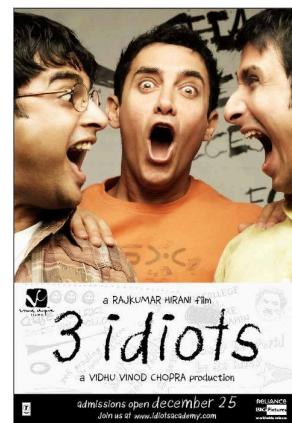
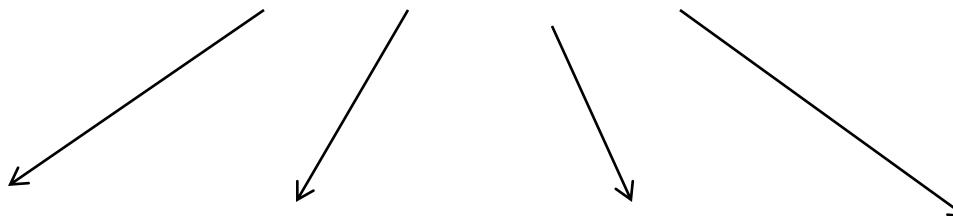
E_4



Netflix and Learn

K_1

Like foreign emotional comedies



E_1

E_2

E_3

E_4

Assume E_1 , E_2 , E_3 and E_4 are conditionally independent given K_1



Netflix and Learn

K_1

Like foreign emotional comedies



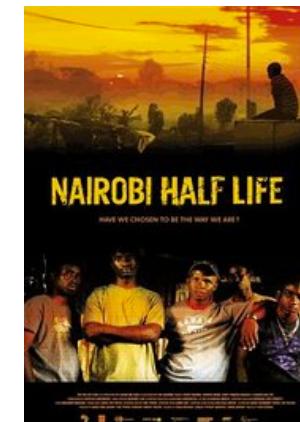
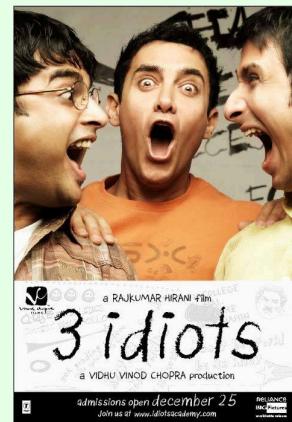
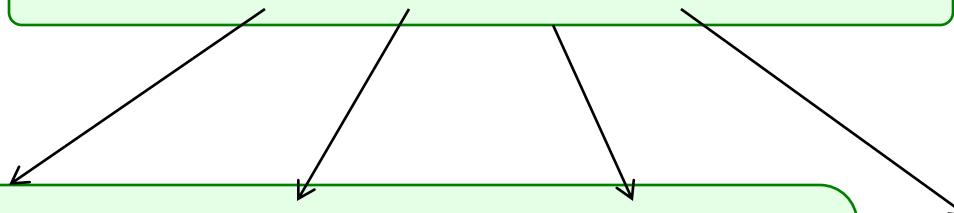
Assume E_1, E_2, E_3 and E_4 are conditionally independent given K_1



Netflix and Learn

K_1

Like foreign emotional comedies



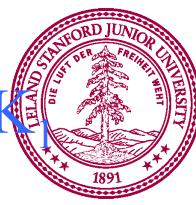
E_1

E_2

E_3

E_4

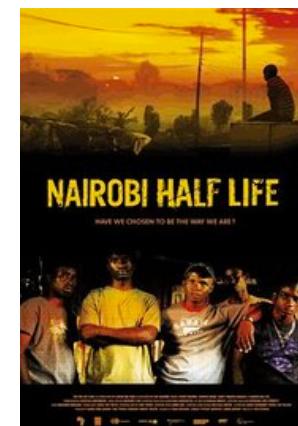
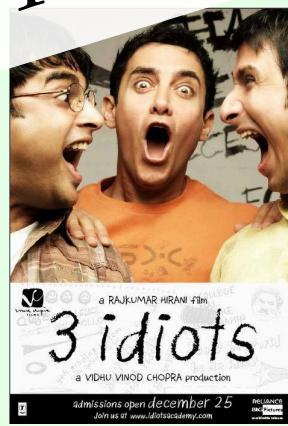
Assume E_1, E_2, E_3 and E_4 are conditionally independent given K_1



Netflix and Learn

K_1

Like foreign emotional comedies



E_1

E_2

E_3

E_4

Assume E_1, E_2, E_3 and E_4 are conditionally independent given K_1



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

Big Deal

“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*”





G₁

G₂

G₃

G₄

G₅

T

Discovered Pattern

