# o5: Independence

David Varodayan

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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

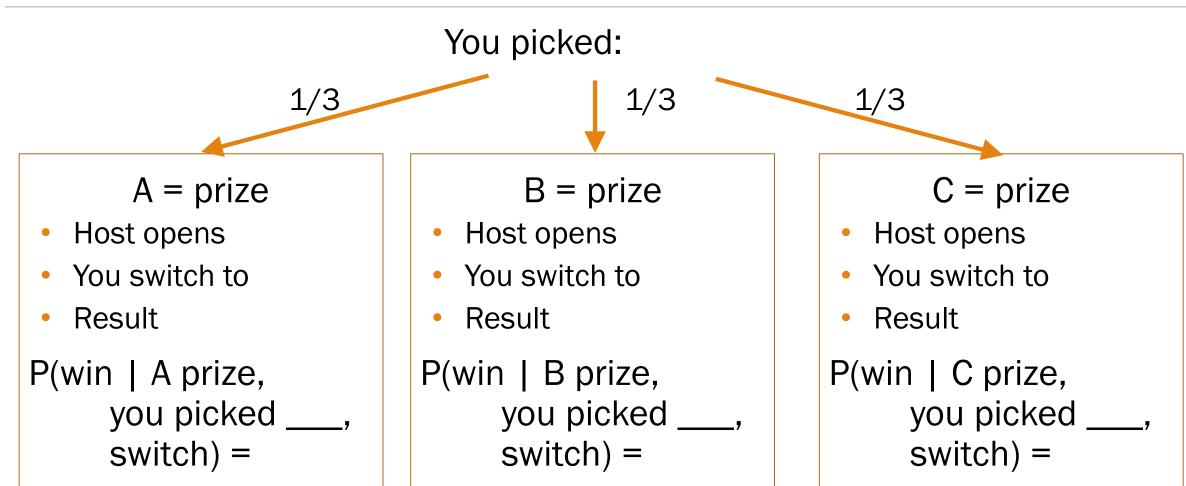
- 1. You choose a door
- 2. Host opens 1 of other 2 doors, revealing a goat
- 3. You are given an option to change to the other door.

Should you switch?



Doors A, B, C

# What happens if you switch



P(win | you picked \_\_\_\_, switched) =

# Monty Hall, 1000 envelope version

Start with 1000 envelopes (of which 1 is the prize).  $\frac{1}{1000}$  = P(envelope is prize) You choose 1 envelope. = P(other 999 envelopes have prize)  $\frac{333}{1000}$  = P(998 empty envelopes had prize) 2. I open 998 of remaining 999 + P(last other envelope has prize) (showing they are empty). = P(last other envelope has prize) P(you win without switching) =original # envelopes 3. Should you switch? P(you win with switching) = <u>original # envelopes - 1</u> original # envelopes Stanford University

#### This class going forward

Last week Equally likely events For most of this course

Not equally likely events

Р

 $\begin{pmatrix} E & given some evidence \\ has been observed \end{pmatrix}$ 

 $P(E \cap F) \qquad P(E \cup F)$ 

(counting, combinatorics)

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- 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

Two Dice

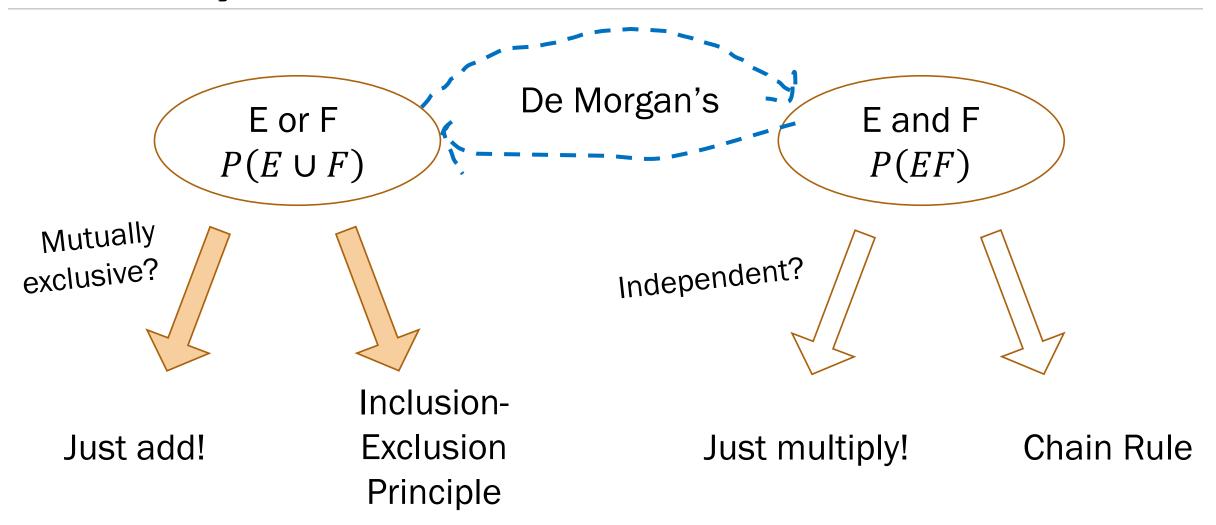
- 4. Roll a 5 on roll 2
- 5. Do not roll a 5 on one of the rolls or both

- A. P(F)
- $\mathsf{B.} \quad P(E \cup F)$
- C.  $P(E^C \cup F^C)$
- **D.** P(EF)
- $\mathsf{E.} \quad P(E^{C}F^{C})$





#### Probability of events



#### Inclusion-Exclusion

- P(student programs in Java) = 0.28
- P(student programs in Python) = 0.07
- P(student programs in Java and Python) = 0.05.

What is P(student does not program in (Java or Python))?

Define events
 & state goal

Want:  $P((E \cup F)^{C})$ 

2. Identify known<br/>probabilities3. Solve

 $P((E \cup F)^{C}) = 1 - P(E \cup F)$ = 1 - [P(E) + P(F) - P(E \circ F)] = 1 - [0.28 + 0.07 - 0.05] = 0.70

P(E)

P(F)

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



**Definition** of conditional probability:

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

The Chain Rule:

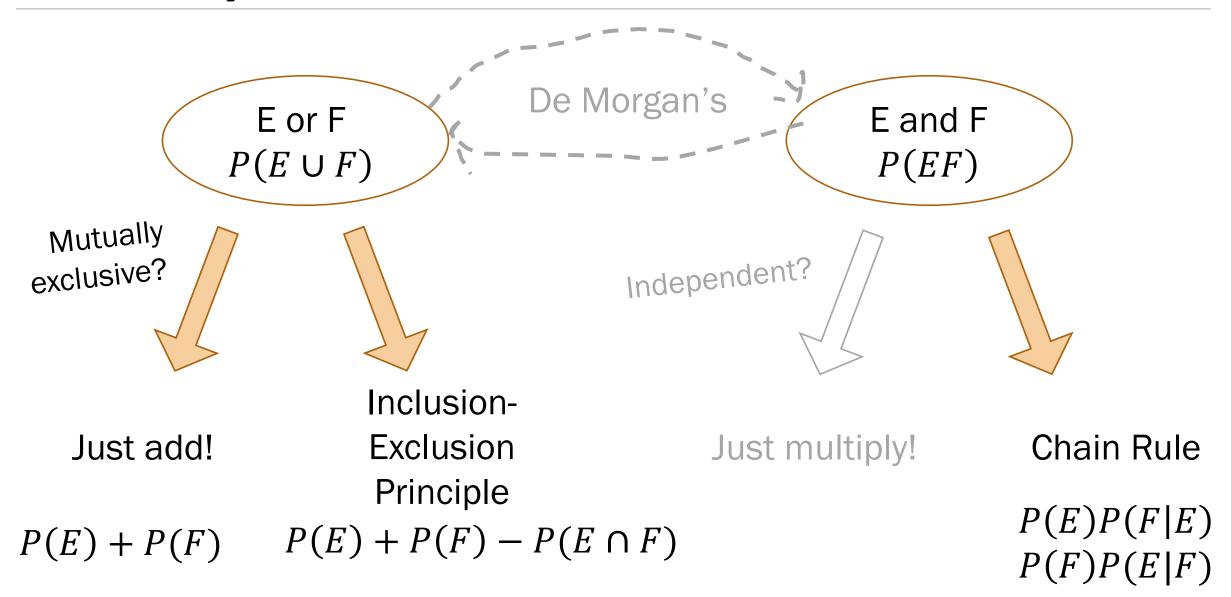
P(EF) = P(E|F)P(F)

#### 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



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#### ndependence

Independent trials

De Morgan's Laws

Conditional independence (if time)

#### CONGRESS, July 4, 1776. A DECLARATION the REPRESENTATIVES Ry of the UNITED STATES OF AMERICA. IN GENERAL CONGRESS ASSEMBLED.

HEN in the Courie of human Events, is becomes necellary for one People to diffolve the Political Bands which have conof the Earth, the feperate and equal Station to which the Laws of Nature and of Nature's God entitle them, a decent Respect to the Opinions of Markind requires that they fhould declare the Caufes which impel them to the feperation.

We hold theie Traths to be felf-evident, that all Men are created nen .- That to fecure thefe Rights, Governments are initiated among abfulute Rule into thefe Colonies Men, deriving their juft Powers from the Confent of the Governed, that whenever any Form of Government becomes defiretives of these Ends. altering fundamentally the Forms of our Governments it is the Right of the People to alree or to aboldh it, and to influtate new Government, laving its foundation on fach Principles, and organizing its with Power to legiflate for to in all Cafes whathever. Powers in fuch Form, as to them fhall frem most likely to effect their Sufety and Happinelis. Prudence, indeed, will dictate that Governments and waging War against us. long eftablished should not be changed for light and transfect Caules r and accordingly all Experience bath thewn, that Mankind are more dif- defroyed the Lives of our People. polied to luffer, while Lyds are fufferable, than to sight themielves by

For quartering large Bodies of armed Troops among us : For protecting them, by a mock Trial, from Punifhment for any Murnected them with another, and to allume among the Powers deri which they flould commit on the Inhabitants of thefe States : For cutting off our Trade with all Parts of the World : For impoting Taxes on us without our Confent : For depriving us, in many Cafes, of the Benefits of Trial by Jory : For transporting us beyond Seas to be tried for pretended Offeners

For abolishing the free fyllem of English Laws in a neighbouring Province. equal, that they are endowed by their Greator with certain unal-enable effablishing therein an arbitrary Government, and enlarging infloondation, for Rights, that among thele are Life, Liberty, and the Putfult of Happi- as to render is at once an Example and ht Inftrument for introducing the fame

For taking away nor Charters, abolifhing our molt valuable Laws, and

For fulpending our own Legislatures, and declaring themfelves invefied

He has abdicated Government here, by declaring us out of his Protection

He has plundered our Seas, ravaged our Coaffs, burnt our. Towns, and

Halt at this Time, transporting large &e nies of foreign Mercenaries to comof thing the Forms to which they are accultomed. But when a long- pleat the Works of Death, Defolation and Tytanny, already began with

Two events *E* and *F* are defined as <u>independent</u> if: P(EF) = P(E)P(F)

#### Otherwise *E* and *F* are called <u>dependent</u> events.

An equivalent definition:

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

#### Intuition through proof

Statement:

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

Proof:

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

Definition of conditional probability

Independence of *E* and *F* 

Knowing that F happened <u>does not</u> <u>change</u> our belief that E happened. Stanford University 14

Independent P(EF) = P(E)P(F)

events E and F P(E|F) = P(E)

#### Dice, our misunderstood friends

- 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:  $D_1 + D_2 = 5$

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

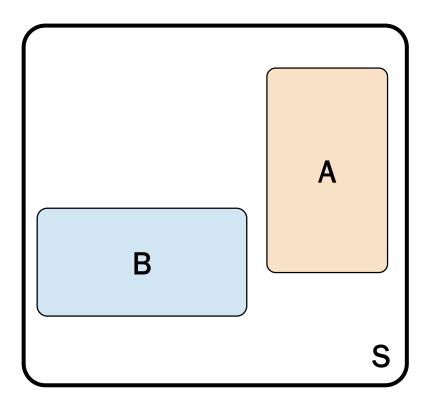
**1.** Are *E* and *F* independent?

2. Are *E* and *G* independent?

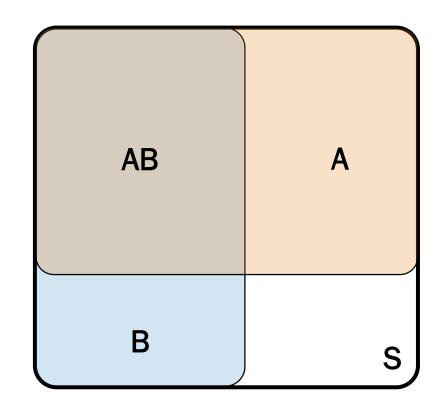


Independent events E and F P(EF) = P(E)P(F)P(E|F) = P(E)

# Independence?



Independent	P(EF) = P(E)P(F)
events $E$ and $F$	P(E F) = P(E)



Independent events E and F 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:

$$P(EF^{C}) = P(E) - P(EF)$$
  
=  $P(E) - P(E)P(F)$   
=  $P(E)[1 - P(F)]$   
=  $P(E)P(F^{C})$ 

E and  $F^{C}$  are independent

Intersection Independence of *E* and *F* Factoring Complement Definition of independence

Knowing that F <u>didn't happen</u> does not change our belief that E happened.

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Independence

Independent trials

De Morgan's Laws

Conditional independence (if time)

#### Generalizing independence

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

*n* events  $E_1, E_2, \dots, E_n$  are

$$P(EFG) = P(E)P(F)P(G), \text{ and}$$

$$P(EF) = P(E)P(F), \text{ and}$$

$$P(EG) = P(E)P(G), \text{ and}$$

$$P(FG) = P(F)P(G)$$
for  $r = 1, ..., n$ :
for every subset  $E_1, E_2, ..., E_r$ :
$$P(E_1, E_2, ..., E_r) = P(E_1)P(E_2) \cdots P(E_r)$$

#### Independent trials:

independent if:

Outcomes of n separate flips of a coin are all independent of one another. Each flip in this case is a <u>trial</u> 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*:  $D_1 + D_2 = 7$



 $G = \{(1,6), (2,5), (3,4), (4,3), (5,2), (6,1)\}$ 

independent?

independent?

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

P(E) = 1/6P(F) = 1/6P(EF) = 1/36

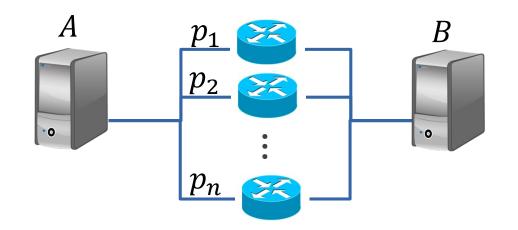
> Pairwise independence is not sufficient to prove independence of >2 events! Stanford University 20

#### Network reliability

Consider the following parallel network:

- *n* independent routers, each with probability  $p_i$  of functioning (where  $1 \le i \le n$ )
- 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(first k heads, then n k tails)
  - 4. P(exactly k heads on n coin flips)

#### Announcements

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Starts:

Late signups/changes: by end of day Solutions:

today end of week

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

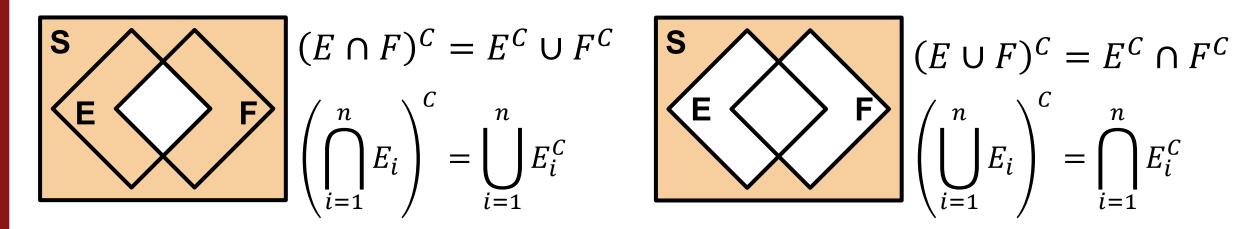
Independence

Independent trials

De Morgan's Laws

Conditional independence

## De Morgan's Laws



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

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#### 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<sub>i</sub> of getting hashed into bucket i.
- **1**. E =bucket 1 has  $\geq 1$  string hashed into it.

What is P(E)?

Define  $S_i = \text{string } i \text{ is}$ hashed into bucket 1  $S_i^C = \text{string } i \text{ is } \underline{\text{not}}$ 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<sub>i</sub> of getting hashed into bucket i.
- **1.** E = bucket 1 has  $\geq$  1 string hashed into it.

What is P(E)?

$$P(E) = P(S_1 \cup S_2 \cup \cdots \cup S_m)$$
  
=  $1 - P((S_1 \cup S_2 \cup \cdots \cup S_m)^C)$   
=  $1 - P(S_1^C S_2^C \cdots S_m^C)$   
=  $1 - P(S_1^C) P(S_2^C) \cdots P(S_m^C)$ 

Define  $S_i = \text{string } i \text{ is}$ hashed into bucket 1  $S_i^C = \text{string } i \text{ is } \underline{\text{not}}$ hashed into bucket 1 Complement  $P(S_i) = p_1$  $P(S_i^C) = 1 - p_1$  $S_i \text{ independent trials}$ 

#### 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<sub>i</sub> of getting hashed into bucket i.

Define

1. E = bucket 1 has  $\geq 1$  string hashed into it. 2. E = at least 1 of buckets 1 to k has  $\geq 1$  string hashed into it.

```
What is P(E)?

P(E) = P(S_1 \cup S_2 \cup \dots \cup S_m)
= 1 - P((S_1 \cup S_2 \cup \dots \cup S_m)^C)
= 1 - P(S_1^C S_2^C \cdots S_m^C)
= 1 - P(S_1^C) P(S_2^C) \cdots P(S_m^C)
```

 $S_i$  = string *i* is hashed into bucket 1  $S_i^C$  = string *i* is <u>not</u> 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  $\geq$  1 string hashed into it.2. E = at least 1 of buckets 1 to k has  $\geq$  1 string hashed into it.3. E = each of buckets 1 to k has  $\geq$  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  $\ge 1$  string hashed into it. 2. E = at least 1 of buckets 1 to k has  $\ge 1$  string hashed into it. 3. E = each of buckets 1 to k has  $\ge 1$  string hashed into it. What is P(E)?

$$P(E) = P(F_1F_2 \cdots F_k)$$
  

$$= 1 - P((F_1F_2 \cdots F_k)^C)$$
 Complement  

$$= 1 - P(F_1^C \cup F_2^C \cup \cdots \cup F_k^C)$$
 De Morgan's Law  

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

Independence

Independent trials

De Morgan's Laws



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# **Conditional Paradigm**

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

Axiom 1 Corollary 1 (complement) Commutativity Chain Rule  $0 \le P(A|E) \le 1$   $P(A|E) = 1 - P(A^{C}|E)$  P(AB|E) = P(BA|E) P(AB|E) = P(B|E)P(A|BE) $P(A|BE) = \frac{P(B|AE)P(A|E)}{P(B|E)}$ 

Bayes' Theorem

Independence relationships can change with conditioning.

A and Bdoes NOT alwaysA and Bindependentmeangiven E.

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Independent events E and F P(EF) = P(E)P(F)P(E|F) = P(E)

#### Two events *A* and *B* are defined as <u>conditionally independent given *E*</u> 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?

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

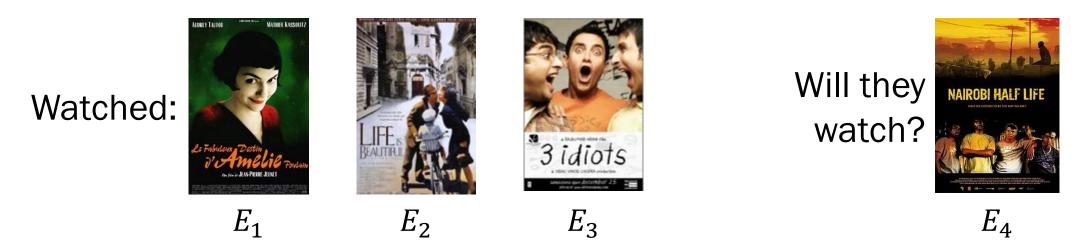
Let *E* be the event that a user watches the given movie. Let *F* be the event that the same user watches Amelie.

			RAJELUMAR HIRAN STIR BALAL DOCRA PRODUCTION WIDDU VINOD CHOPRA Production WIDDU VINOD CHOPRA Production WIDDU VINOD CHOPRA Production	
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!		Stanford University 35



#### Review

#### Netflix and Condition



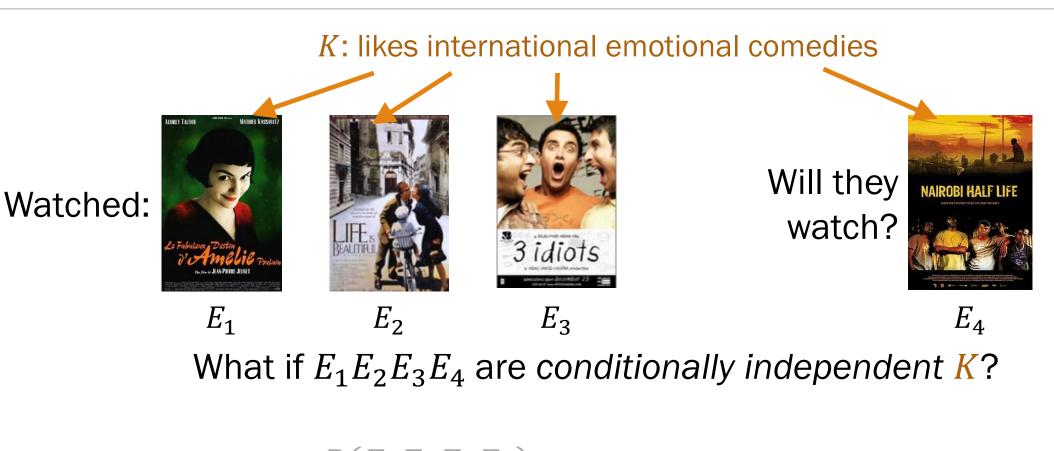
What if  $E_1E_2E_3E_4$  are not independent? (e.g., all international emotional comedies)

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

$$\frac{P(E_1E_2E_3E_4)}{P(E_1E_2E_3)}$$

$$\frac{P(E_1E_2E_3)}{P(E_1E_2E_3)}$$

#### Netflix and Condition



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

$$P(E_4|E_1E_2E_3K) = P(E_4|K)$$

#### 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



 $E_1E_2E_3E_4$  are dependent

#### $E_1E_2E_3E_4$ are conditionally independent given K

Dependent events can become conditionally independent. Stanford University 39