



Computer  
Science

# CSC380: Principles of Data Science

## Probability Primer

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Acknowledgement and thanks: Materials Built on previous product by

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# Annoucements

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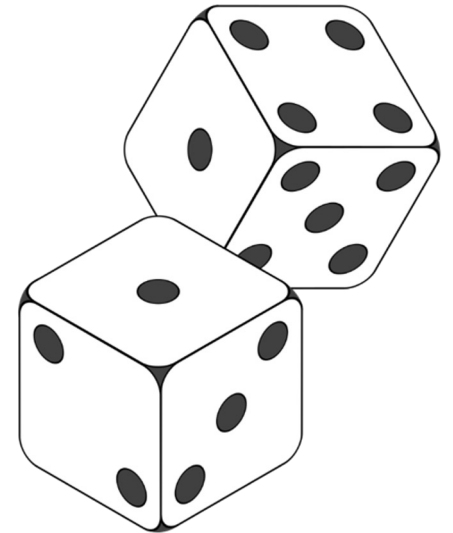
- Readings before next Tuesday
  - Ch. 6 (WJ: Watkins, J., “An Introduction to the Science of Statistics: From Theory to Implementation”)

- Random Events and Probability
- Axioms of Probability
- Set Theory
- Law of Total Probability

# Random Events and Probability

***Suppose we roll two fair dice...***

- ❑ What are the possible outcomes?
- ❑ What is the *probability* of rolling **even** numbers?
- ❑ What is the *probability* of having two numbers sum to 6?
- ❑ If one die rolls 1, then what is the probability of the second die also rolling 1?



***...this is a random process.***

How to mathematically formulate outcomes  
and compute these probabilities?

# Random Events and Probability

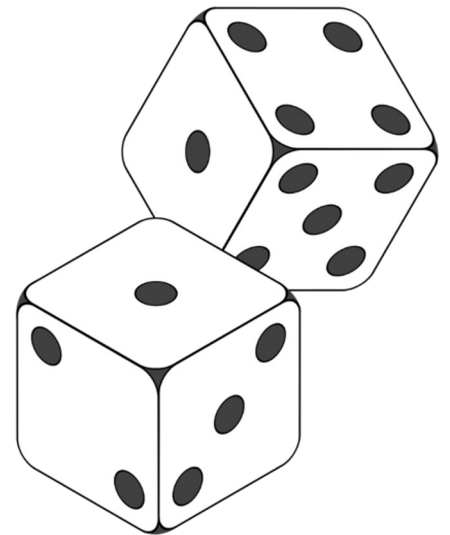
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Probability of a random event

$\approx$

Simulate the random process  $n$  times, the fraction of times this event happens

- How large should  $n$  be?
- Simulation results vary from trails?



# Background: Numpy in Python

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## Numpy: numerical computing package

```
import numpy as np
np.random.randint(1,1+6,size=10)
=> array([5, 4, 1, 1, 1, 5, 5, 2, 4, 6])
```

```
randint(low,high,size)
: generate `size` random numbers
in {low, low+1, ..., high-1}
```

## Numpy array

- Replaces python's list in numpy.
- More numerical functionality
- It's a 'vector' in mathematics.

```
a=np.array([1,2]); b=np.array([4,5])
a+b
=> np.array([5,7]) // elementwise addition
np.dot(a,b)
=> 14           // dot product
```

# Random Events and Probability

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*Consider: What is the probability of having two numbers sum to 6?*

```
import numpy as np
for n in [10,100,1_000,10_000,100_000]:
    res_dice1 = np.random.randint(1,6+1,size=n)
    res_dice2 = np.random.randint(1,6+1,size=n)
    res = [(res_dice1[i], res_dice2[i]) for i in range(len(res_dice1))]

    cnt = len(list(filter(lambda x: x[0] + x[1] == 6, res)))
    print("n=%6d, result: %.4f " % (n, cnt/n))
```

```
n=    10, result: 0.1000
n=   100, result: 0.1200
n=  1000, result: 0.1350
n= 10000, result: 0.1365
n=100000, result: 0.1388
n=1000000, result: 0.1385
```

```
n=    10, result: 0.1000
n=   100, result: 0.1900
n=  1000, result: 0.1540
n= 10000, result: 0.1366
n=100000, result: 0.1371
n=1000000, result: 0.1394
```

every time you run, you  
get a different result

however, the number  
seems to converge to  
0.138-0.139

There seems to be a precise value that it will converge to.. what is it?



A 6x6 grid of coordinate pairs (x,y) is shown. The pairs are arranged in a pattern that suggests a sequence or a specific arrangement. The top right cell contains the text "Event: E".

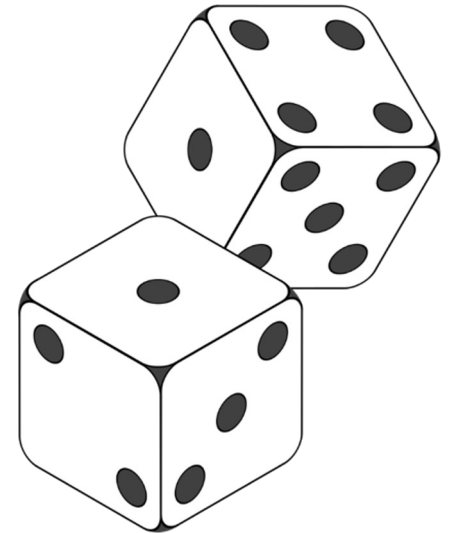
|       |       |       |       |          |       |
|-------|-------|-------|-------|----------|-------|
| (2,2) | (3,5) |       |       |          |       |
|       | (5,3) | (6,2) |       | Event: E | (6,6) |
| (3,2) |       | (6,4) |       |          |       |
|       | (2,6) |       | (1,2) | (2,4)    | (4,2) |
| (5,2) |       | (3,4) | (2,1) |          |       |
|       | (4,3) | (1,6) |       | (1,5)    | (5,1) |
| (4,4) |       | (5,6) |       | (3,3)    |       |
|       | (6,5) | (6,1) | (5,5) |          |       |
|       |       |       |       | (5,4)    | (4,1) |
| (4,5) | (2,5) | (6,3) | (1,1) |          | (1,4) |
| (3,6) |       | (4,6) |       | (1,3)    | (3,1) |
|       |       |       |       |          | (2,3) |

answer:  
 $(1/36) * 5 = 0.13888..$

# Random Events and Probability

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- **Theoretical probability** describes how likely an event is going to occur based on math.
- **Experimental probability** describes how frequently an event actually occurred in an experiment.

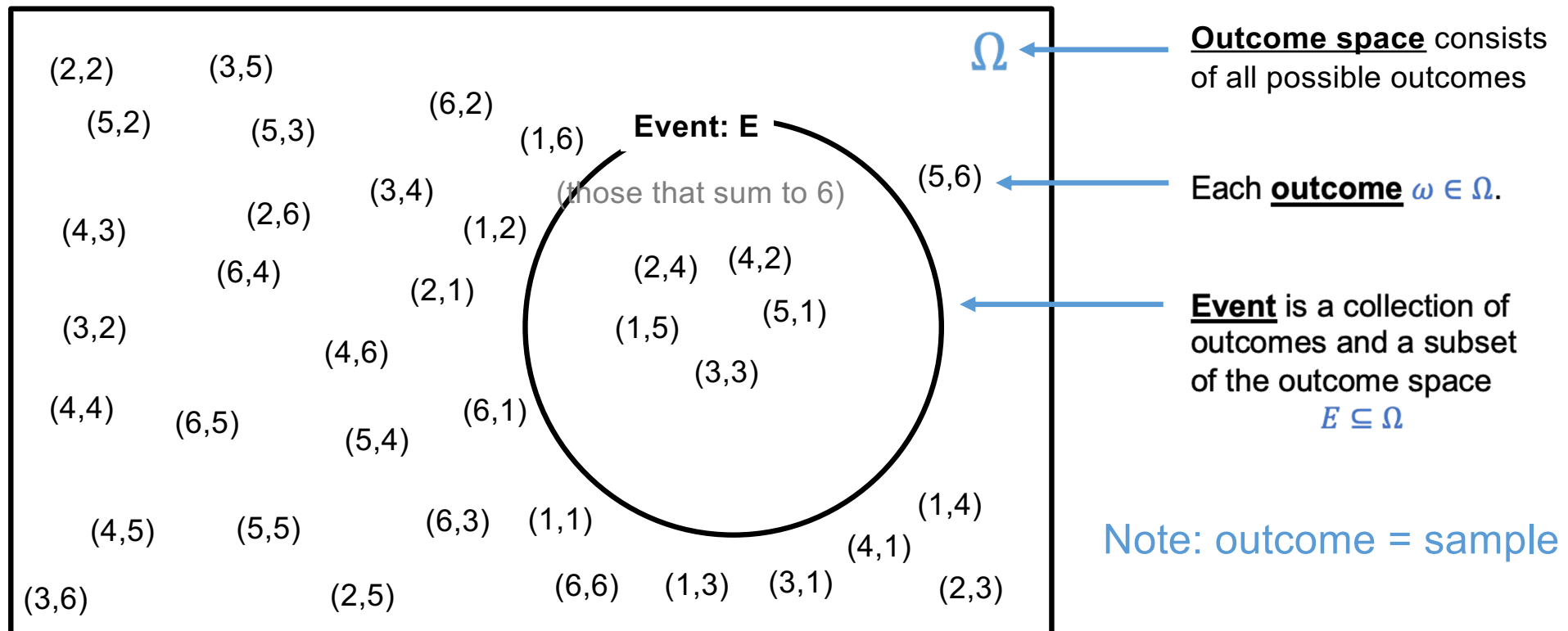


- **Probability** is a real-world phenomenon.
- But under what mathematical framework can we formulate **probability** so we can solve practical problems?
  - e.g., weather prediction, predicting the election outcome
- **Disclaimer**: not all mathematics correspond to real-world phenomenon (e.g., Banach–Tarski paradox). Fortunately, we will not talk about this in our lecture 😊

# Random Events and Probability

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*Consider: What is the probability of having two numbers sum to 6?*



## Some examples of events...

- Both even numbers

Q: how many such pairs?

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$$E^{\text{even}} = \{(2, 2), (2, 4), \dots, (6, 4), (6, 6)\}$$

- The sum of both dice is even,

$$E^{\text{sum even}} = \{(1, 1), (1, 3), (1, 5), \dots, (2, 2), (2, 4), \dots\}$$

- The sum is greater than 12,

$$E^{\text{sum} > 12} = \emptyset$$

We can talk about  
impossible outcomes

- The product is even (How many events????)

# Axioms of Probability

# Random Events and Probability

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But, what is probability, really?

(e.g., can explain the probability of seeing an event when throwing two dice)

Mathematicians have found a set of conditions that 'makes sense'.

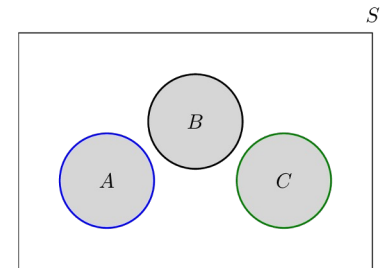
- Probability is a map  $P$  on a set  $\Omega$ .  
 $\Rightarrow$  i.e., takes in an event, spits out a real value
- $P$  must map events to a real value in interval  $[0,1]$ .
- $P$  is a (valid) **probability distribution** if it satisfies the following **axioms of probability**,

1. For any event  $E$ ,  $P(E) \geq 0$

2.  $P(\Omega) = 1$

3. For any sequence of disjoint events  $E_1, E_2, E_3, \dots$

$$P\left(\bigcup_{i \geq 1} E_i\right) = \sum_{i \geq 1} P(E_i)$$



**disjoint:** intersection is empty

# Random Events and Probability

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- Many properties follows (i.e., can be proved mathematically)

$$\mathbb{P}(\emptyset) = 0$$

$$A \subset B \implies \mathbb{P}(A) \leq \mathbb{P}(B) \quad \text{E.g., throw a die. } A = \text{getting 1, } B = \text{getting an odd number}$$

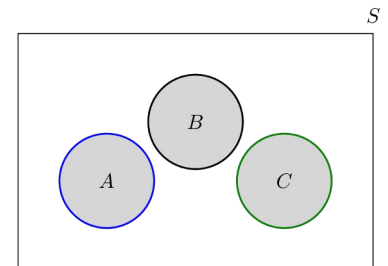
$$0 \leq \mathbb{P}(A) \leq 1$$

$$\mathbb{P}(A^c) = 1 - \mathbb{P}(A)$$

$$A \cap B = \emptyset \implies \mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B). \quad \text{E.g., } A = \text{getting 1, } B = \text{getting 3 or 5}$$

$A^c$

the complement of A. All outcomes not in A



(I recommend that you maintain your own version of cheat sheet!)



# Random Events and Probability

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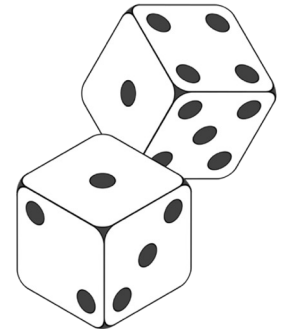
## Special case

Assume each outcome is equally likely, and sample space is finite, then the probability of event is:

$$P(E) = \frac{|E|}{|\Omega|}$$

Number of elements  
in event set

Number of possible  
outcomes (36)



This is called uniform probability distribution

Q: What axiom we are using?  
=> Axiom 3

**(Fair) Dice Example:** Probability that we roll even numbers,

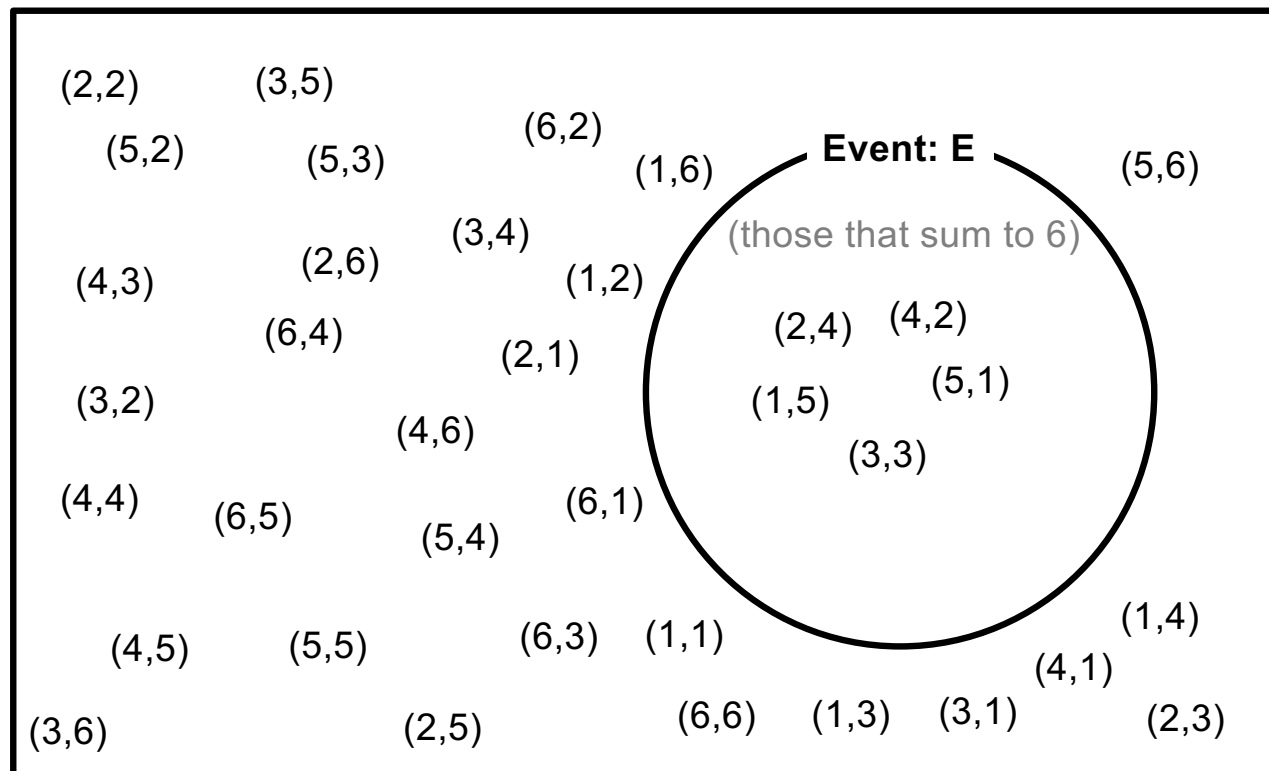
$$\begin{aligned} P((2, 2) \cup (2, 4) \cup \dots \cup (6, 6)) &= P((2, 2)) + P((2, 4)) + \dots + P((6, 6)) \\ &= \frac{1}{36} + \frac{1}{36} + \dots + \frac{1}{36} = \frac{9}{36} \end{aligned}$$

9 Possible outcomes, each with  
equal probability of occurring

# Random Events and Probability

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*Consider: What is the probability of having two numbers sum to 6?*



Each outcome is equally likely.  
by the **independence**  
(will learn this concept later)  
=> 1/36

# of outcomes that sum to 6:  
=> 5

answer:  
(1/36) \* 5 = 0.13888..

$$P(E) = \frac{|E|}{|\Omega|}$$

# Set Theory

# Set Theory

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**Two dice example:** Suppose

$E_1$  : First die equals 1

$E_2$  : Second die equals 1

$$E_1 = \{(1, 1), (1, 2), \dots, (1, 6)\}$$

$$E_2 = \{(1, 1), (2, 1), \dots, (6, 1)\}$$

**Operators on events:**

| Operation                 | Value  | Interpretation             |
|---------------------------|--|----------------------------|
| $E_1 \cup E_2$            | $\{(1, 1), (1, 2), \dots, (1, 6), (2, 1), \dots, (6, 1)\}$ | Any die rolls 1            |
| $E_1 \cap E_2$            | $\{(1, 1)\}$   | Both dice roll 1           |
| $E_1 \setminus E_2$       | $\{(1, 2), (1, 3), (1, 4), (1, 5), (1, 6)\}$               | Only the first die rolls 1 |
| $\overline{E_1 \cup E_2}$ | $\{(2, 2), (2, 3), \dots, (2, 6), (3, 2), \dots, (6, 6)\}$ | No die rolls 1             |

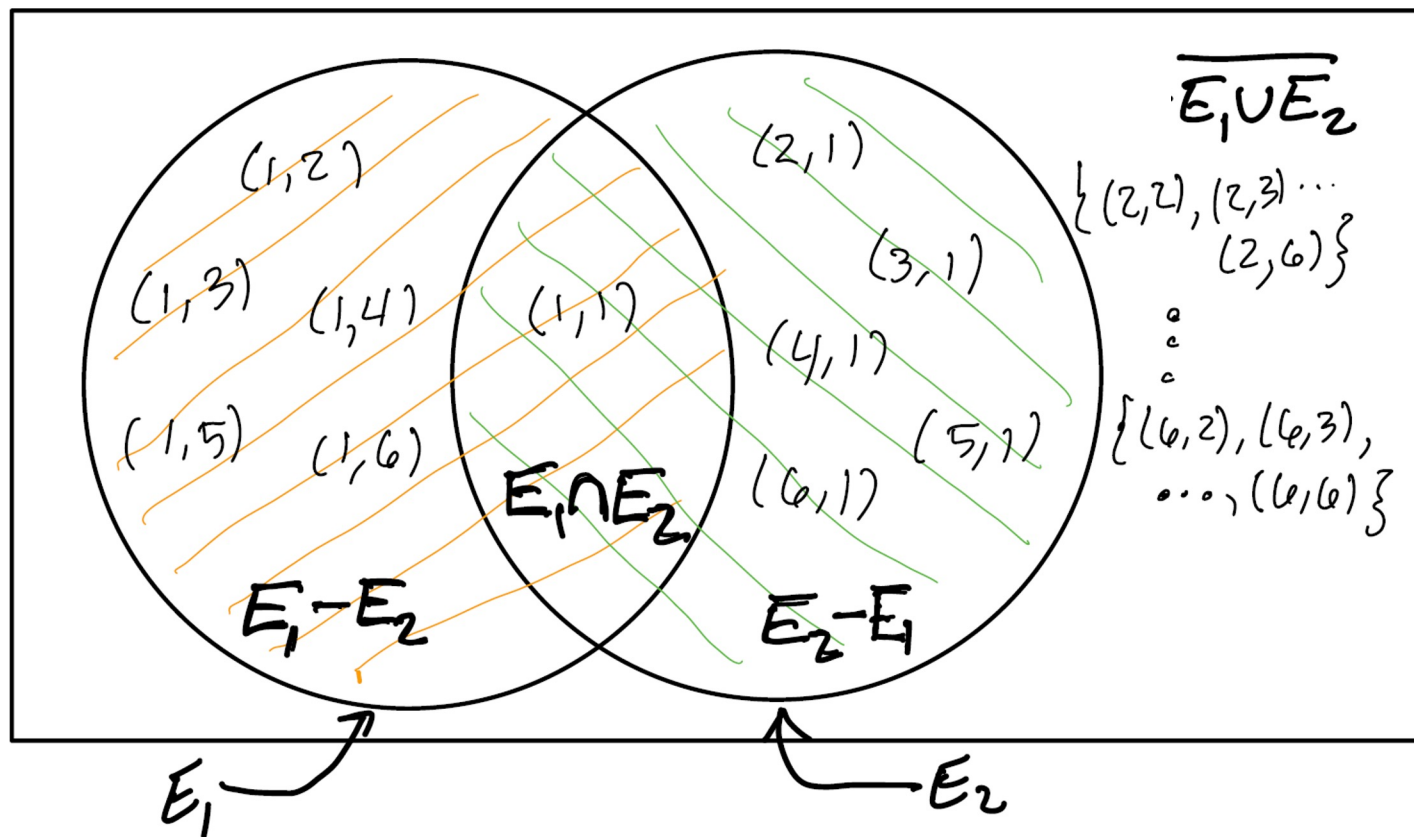
(=  $E_1 - E_2 := E_1 \cap E_2^c$ )

(=  $(E_1 \cup E_2)^c$ )

# Set Theory

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*Can interpret these operations as a Venn diagram...*



# Set Theory

## More results

- $\neg(\bigcup_n A_n) = \bigcap_n \neg A_n$ ,  $\neg(\bigcap_n A_n) = \bigcup_n \neg A_n$  DEMORGAN

Special case:  $\neg(A \cup B) = \neg A \cap \neg B$

Notation:  $\neg A := A^c$

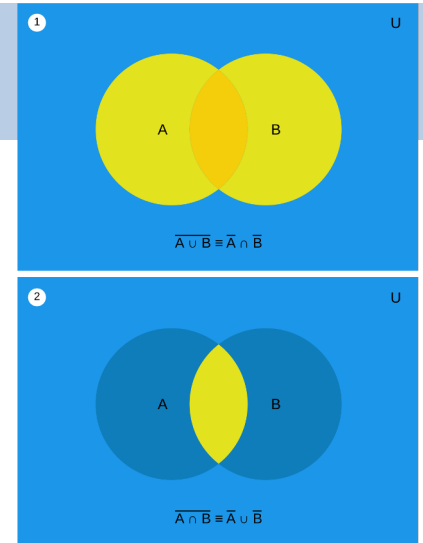
- $A \cap (B \cup C) = (A \cap B) \cup (A \cap C)$  and  $A \cup (B \cap C) = (A \cup B) \cap (A \cup C)$ .

$$A \cap (\bigcup_i B_i) = \bigcup_i (A \cap B_i), \quad A \cup (\bigcap_i B_i) = \bigcap_i (A \cup B_i)$$

- $B = \Omega \cap B = (A \cup \neg A) \cap B = (A \cap B) \cup (\neg A \cap B)$

// by distributive law

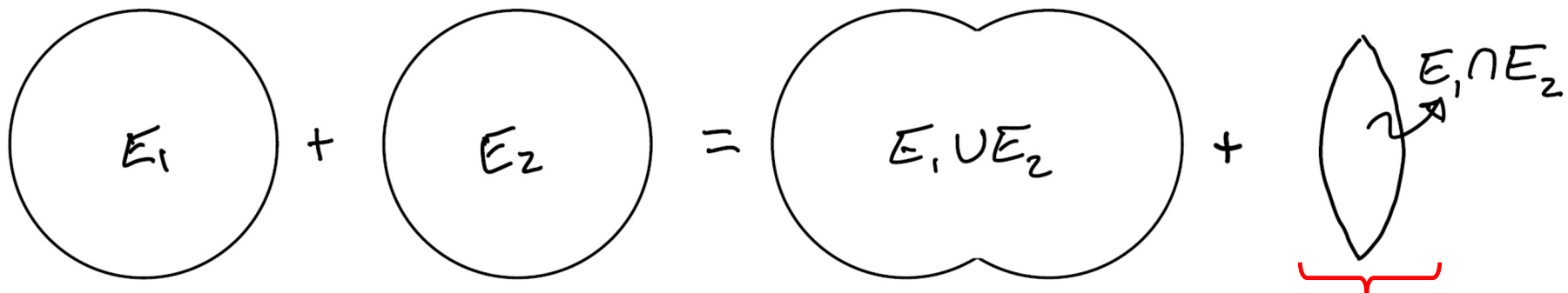
**TIP**: always draw a picture to visualize these identities!



**Lemma: (inclusion-exclusion rule)** For any two events  $E_1$  and  $E_2$ ,

$$P(E_1 \cup E_2) = P(E_1) + P(E_2) - P(E_1 \cap E_2)$$

**Graphical Proof:**



Subtract from both sides

## Alternative Proof

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**Lemma:** For any two events  $E_1$  and  $E_2$ ,

$$P(E_1 \cup E_2) = P(E_1) + P(E_2) - P(E_1 \cap E_2)$$

**Alternative proof:**



$$\begin{aligned} A &= E_1 - (E_1 \cap E_2) \\ B &= E_1 \cap E_2 \\ C &= E_2 - (E_1 \cap E_2). \end{aligned}$$

$$\begin{aligned} P(E_1 \cup E_2) &= P(A \cup B \cup C) \\ &= P(A) + P(B) + P(C) \\ &= P(A) + P(B) + P(B) + P(C) - P(B) \\ &= P(A \cup B) + P(B \cup C) - P(B) \end{aligned}$$

(by axiom 3)

(by axiom 3)



## Exercise:

Quiz candidate

- Consider rolling two fair dice
- $E_1$ : two dice sum to 6
- $E_2$ : second die is even
- Compute the numerical value of  $P(E_1 \cup E_2)$ . Hint: Use inclusion-exclusion rule.

$$P(E_1) = 5/36$$

$$P(E_2) = 18/36$$

$$P(E_1 \cap E_2) = 2/36$$

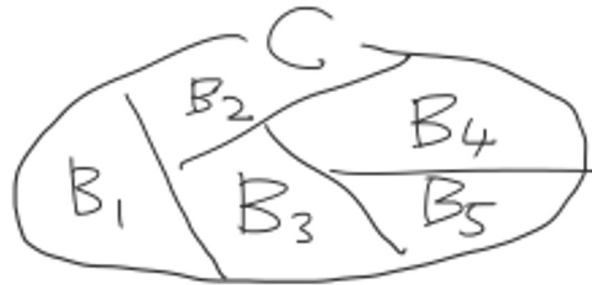
answer: 21/36

# Law of Total Probability

# Random Events and Probability

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**[Def]** The set of events  $\{B_i\}_{i=1}^n$  **partitions** outcome space  $C \Leftrightarrow \cup_i B_i = C$  and  $B_1, B_2, \dots$  are disjoint.

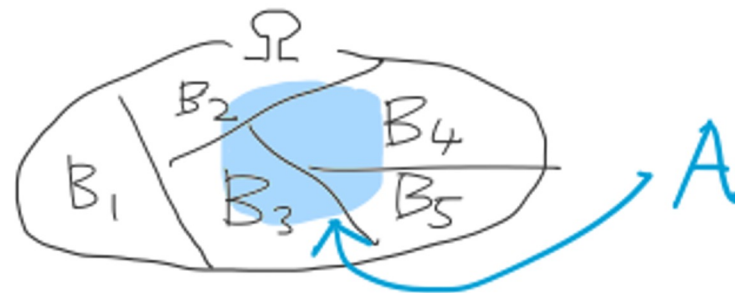


Q: Why is this true?

A: [Axiom 3!](#)

$$A = A \cap \Omega = A \cap (\cup_i B_i) = \cup_i (A \cap B_i)$$

Now,  $\{A \cap B_i\}_{i=1}^n$  partitions  $A$



# Random Events and Probability

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**Law of total probability:** Let  $A$  be an event. For any events  $B_1, B_2, \dots$  that partitions  $\Omega$ , we have

$$P(A) = \sum_i P(A \cap B_i)$$

**Example** Roll two fair dice. Let  $X$  be the outcome of the first die. Let  $Y$  be the sum of both dice. What is the probability that both dice sum to 6 (i.e.,  $Y=6$ )?

quiz candidate

$$p(Y = 6) = \sum_{x=1}^6 p(Y = 6, X = x)$$

$$P(A, B) := P(A \cap B)$$

$$\begin{aligned} &= p(Y = 6, X = 1) + p(Y = 6, X = 2) + \dots + p(Y = 6, X = 6) \\ &= \frac{1}{36} + \frac{1}{36} + \frac{1}{36} + \frac{1}{36} + \frac{1}{36} + 0 = \frac{5}{36} \end{aligned}$$

## Summary So Far

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- Most of the rules we learned is basically set theory + axiom 3

So, here is a generic workflow for computing  $P(A)$ .

1. Use set theory and slice and dice  $A$  into a manageable partition of  $A$  where  $P(\text{each piece of partition})$  is easy to compute.
2. Apply Axiom 3.

# Conditional Probability

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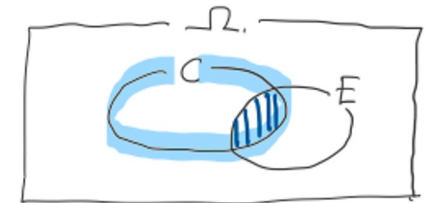
- Two fair dice example:

- Suppose I roll two dice secretly and tell you that one of the dice is 2. C
- In this situation, find the probability of two dice summing to 6. E

```
import numpy as np
for n in [10,100,1000,10_000,100_000, 1_000_000]:
    res_dice1 = np.random.randint(6,size=n) + 1
    res_dice2 = np.random.randint(6,size=n) + 1
    res = [(res_dice1[i], res_dice2[i]) for i in range(len(res_dice1))]
```

```
conditioned = list(filter(lambda x: x[0] == 2 or x[1] == 2, res))
n_eff = len(conditioned)
```

```
cnt = len(list(filter(lambda x: x[0] + x[1] == 6, conditioned)))
print("n=%9d, n_eff=%9d, result: %.4f " % (n, n_eff, cnt/n_eff))
```



compare:  
without conditioning,  
it was 0.138..

|    |                 |                        |
|----|-----------------|------------------------|
| n= | 10, n_eff=      | 4, result: 0.0000      |
| n= | 100, n_eff=     | 32, result: 0.2500     |
| n= | 1000, n_eff=    | 300, result: 0.1733    |
| n= | 10000, n_eff=   | 3002, result: 0.1742   |
| n= | 100000, n_eff=  | 30590, result: 0.1823  |
| n= | 1000000, n_eff= | 305616, result: 0.1818 |

|    |                 |                        |
|----|-----------------|------------------------|
| n= | 10, n_eff=      | 3, result: 0.3333      |
| n= | 100, n_eff=     | 32, result: 0.0625     |
| n= | 1000, n_eff=    | 343, result: 0.2245    |
| n= | 10000, n_eff=   | 3062, result: 0.1897   |
| n= | 100000, n_eff=  | 30651, result: 0.1811  |
| n= | 1000000, n_eff= | 305580, result: 0.1808 |