



Computer
Science

CSC380: Principles of Data Science

Probability Primer

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Overview for today

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

Acknowledgement and thanks:

Materials Built on previous product by

- Jason Pacheco,
- Kwang-Sung Jun,
- Chicheng Zhang
- Xincheng yu

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Announcements

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• Reading

- Ch. 6 ([WJ: Watkins, J., "An Introduction to the Science of Statistics: From Theory to Implementation"](#))

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Outline

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

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Random Events and Probability

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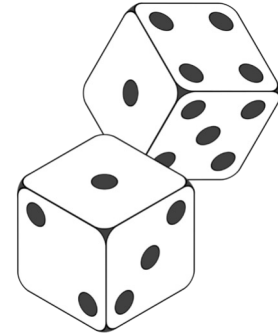
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Random Events and Probability

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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?

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Random Events and Probability

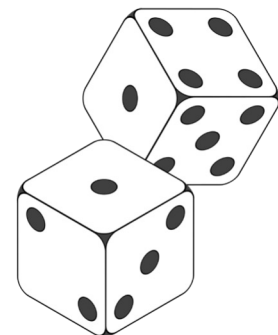
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Probability of a random event

\approx

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

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



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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
```

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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

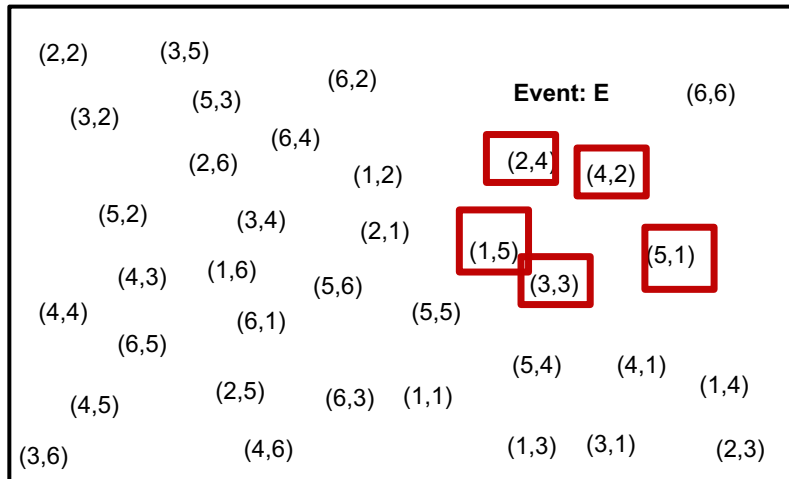
There seems to be a precise value that it will converge to.. what is it?

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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)

outcomes: 36

Probability that one specific
outcome (say (3,3) appears)

=> $1/36$

of **outcomes** that sum to 6:
=> 5

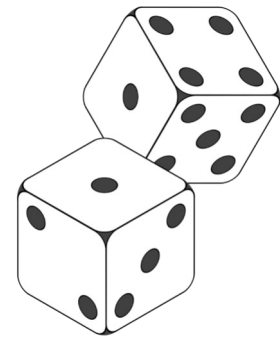
answer:
 $(1/36) * 5 = 0.13888..$

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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.



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Mathematics of Probability

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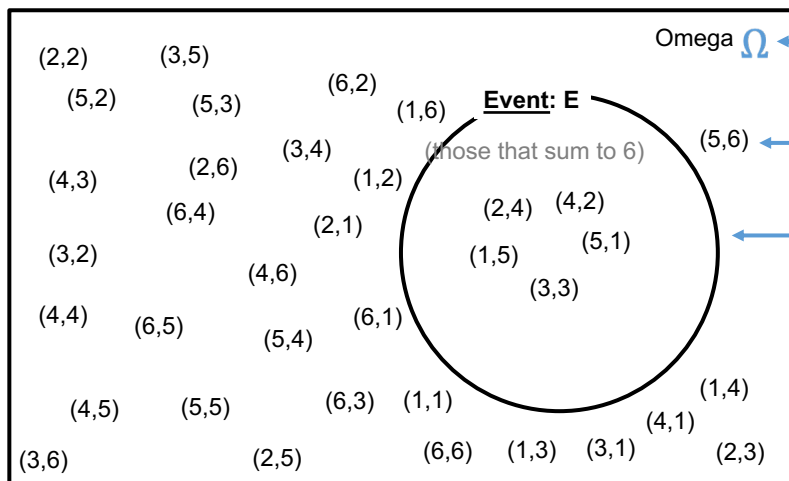
- **Probability** is a real-world phenomenon.
- But under what mathematical framework can we formulate **probability** so we can solve practical problems?
 - e.g., weather prediction, predicting the election outcome
- **Disclaimer:** not all mathematics correspond to real-world phenomenon (e.g., Banach–Tarski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

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Random Events and Probability

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



Omega Ω

Outcome space consists of all possible outcomes

Each **outcome** $\omega \in \Omega$.

Event is a collection of outcomes and a subset of the outcome space
 $E \subseteq \Omega$

Note: outcome = sample

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Random Events and Probability

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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 is even,

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

- The sum is greater than 12,

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

This is a well defined event. However it never occurs

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

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Axioms of Probability

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Random Events and Probability

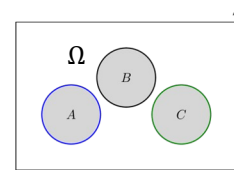
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But, what is probability, really?

(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 defined on a set Ω . \Rightarrow i.e., takes in an event, spits out a real value
- P must map every events (that is, every shape on Ω) 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



disjoint: intersection is empty

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

$$P(A \cup B \cup C) = P(A) + P(B) + P(C)$$

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Random Events and Probability

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

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

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

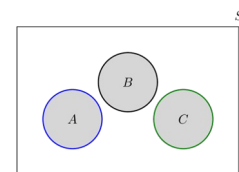
$$0 \leq \mathbb{P}(A) \leq 1$$

$$\mathbb{P}(A^c) = 1 - \mathbb{P}(A) \quad \text{E.g., } A = \text{getting 1, } B = \text{getting 3 or 5}$$

$$A \cap B = \emptyset \implies \mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B).$$

disjoint events

A^c , sometimes denoted. A , is the complement of A . All outcomes not in A



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

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Random Events and Probability

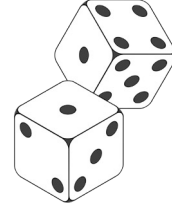
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Special case

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,

$$P((2,2) \cup (2,4) \cup \dots \cup (6,6)) = P((2,2)) + P((2,4)) + \dots + P((6,6))$$

9 Possible outcomes, each with equal probability of occurring

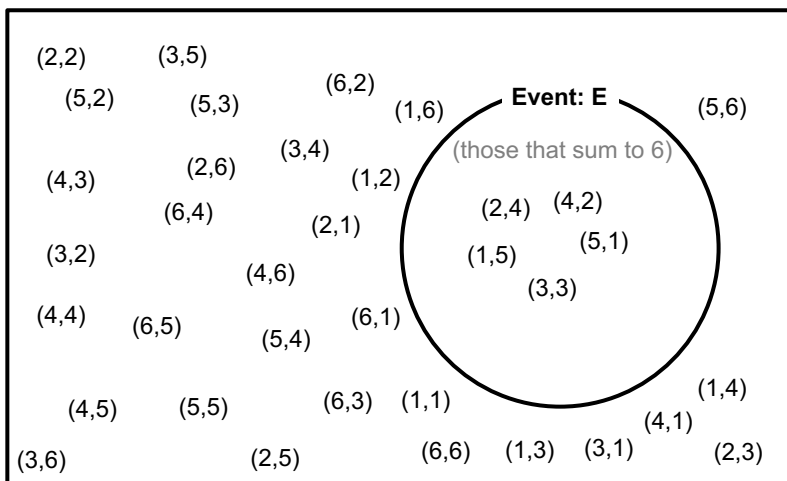
$$= \frac{1}{36} + \frac{1}{36} + \dots + \frac{1}{36} = \frac{9}{36}$$

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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..

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

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Set Theory

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Set Theory20

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:

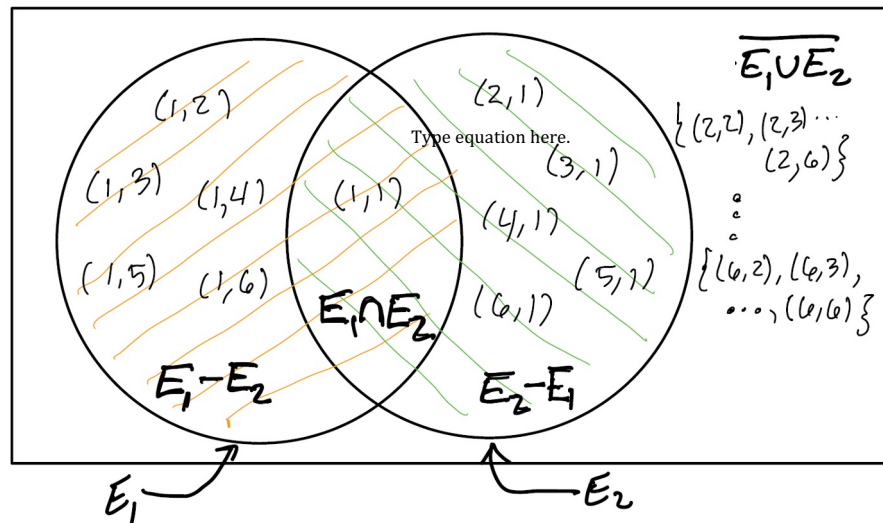
Operation	Value	Interpretation
Union (this OR this) $E_1 \cup E_2$	$\{(1, 1), (1, 2), \dots, (1, 6), (2, 1), \dots, (6, 1)\}$	Any die rolls 1
Intersection $E_1 \cap E_2$	$\{(1, 1)\}$	Both dice roll 1
Set Minus $E_1 \setminus E_2$	$\{(1, 2), (1, 3), (1, 4), (1, 5), (1, 6)\}$	Only the first die rolls 1
<div>$\overline{E_1 \cup E_2}$</div> <div>$(= (E_1 \cup E_2)^c)$</div>	$\{(2, 2), (2, 3), \dots, (2, 6), (3, 2), \dots, (6, 6)\}$	No die rolls 1

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Set Theory

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Can interpret these operations as a Venn diagram...



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Set Theory

More results

$$\left(\bigcup_{i=1}^n A_i \right)^c = \bigcap_{i=1}^n (A_i^c) \quad \text{De Morgan}$$

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 $A \cup (B \cap C) = (A \cup B) \cap (A \cup C)$. // distributive law
 $A \cap (\cup_i B_i) = \cup_i (A \cap B_i), \quad A \cup (\cap_i B_i) = \cap_i (A \cup B_i)$
- $B = \Omega \cap B = (A \cup \neg A) \cap B = (A \cap B) \cup (\neg A \cap B)$ // by distributive law

TIP: always draw a picture to visualize these identities!

For more, see https://math.libretexts.org/Courses/Mount_Royal_University/MATH_1150%3A_Mathematical_Reasoning/2%3A3A_Basic_Concepts_of_Sets/2.3%3A_Properties_of_Sets

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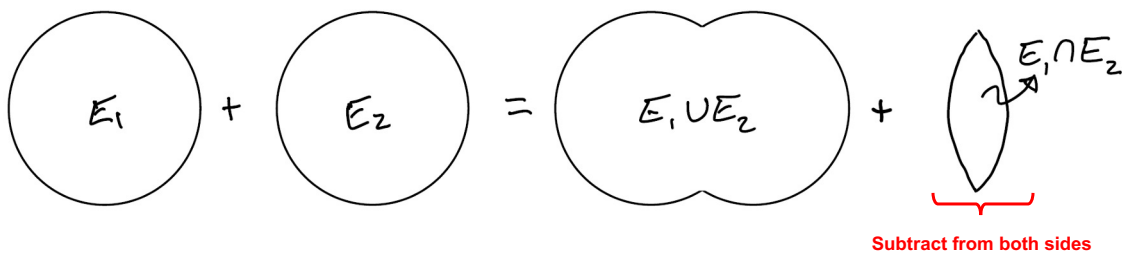
Random Events and Probability

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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:



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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:

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



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

(by axiom 3)

(by axiom 3)

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Random Events and Probability

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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

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Law of Total Probability

Pay attention – we will use it numerous times

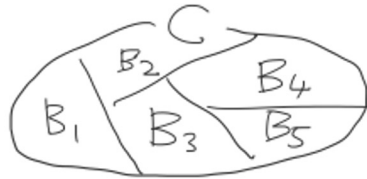
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Random Events and Probability

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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.



Claim:

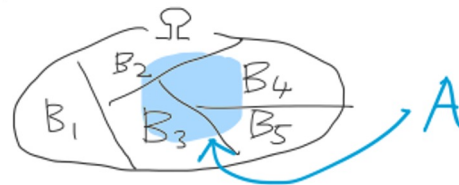
$$P(C) = \sum P(C \cap B_i)$$

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



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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 Ω , 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

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

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

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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.

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Conditional Probability

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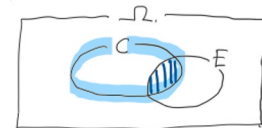
- Two fair dice example:

- 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
```

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