



Probability and Statistics

UG2, Core course, IIIT,H

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IIIT, Hyderabad

August 31, 2021

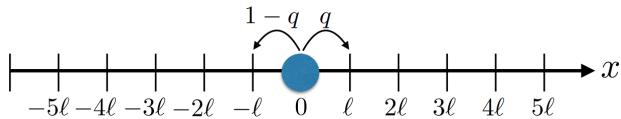
- ① Random Walks
- ② Conditional Probability, Bayes Theorem
- ③ Define Conditional Probability and Chain Rule

- ④ The Monty Hall Problem
- ⑤ Independence
- ⑥ Conditional Independence

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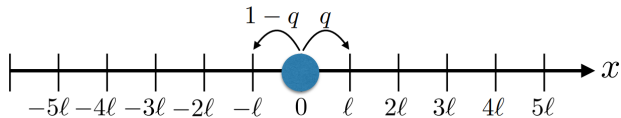
1 Random Walk, Probability

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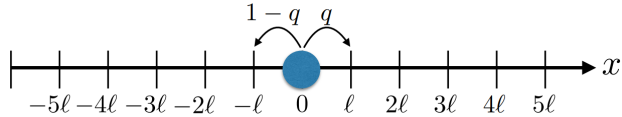
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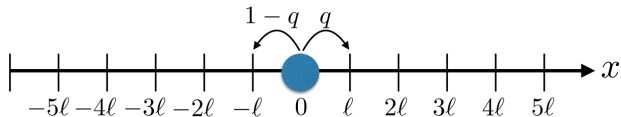
- Consider a person at $x = 0$, he can travel one step to the right or to the left

1 Random Walk, Probability

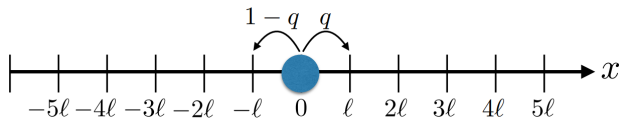
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- Consider a person at $x = 0$, he can travel one step to the right or to the left
 - He can travel one step to the right with probability q



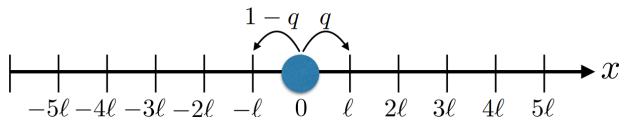
- Consider a person at $x = 0$, he can travel one step to the right or to the left
 - He can travel one step to the right with probability q
 - He can travel one step to the left with probability $(1 - q)$



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Simple Random Walk in 1D

A walk is called **simple random walk** in 1D if there is a **equal** probability of either going to right or going to the left. Above, we set $p = q = 1/2$



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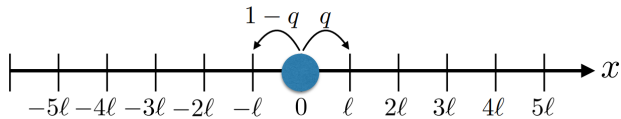
Question

What is the probability that the person after i th step is at $x = 0$?

1 Analysis of Simple Random Walk in 1D

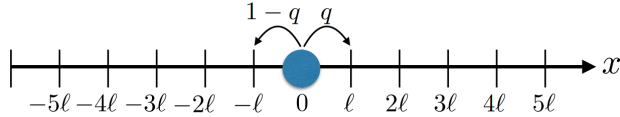
1 Analysis of Simple Random Walk in 1D

| 4

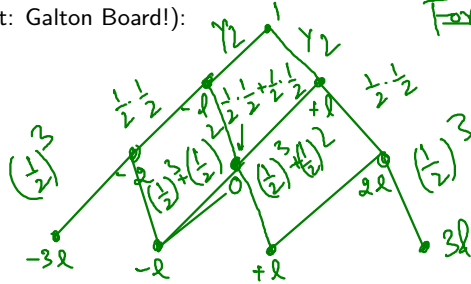


1 Analysis of Simple Random Walk in 1D

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Draw the choice tree (Hint: Galton Board!):



For a simple random walk

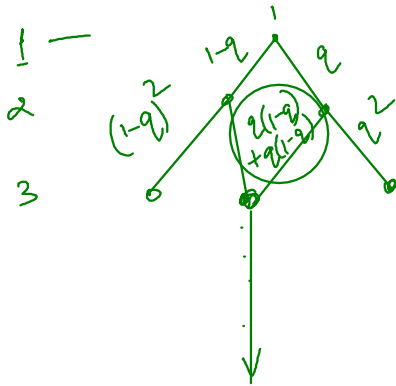
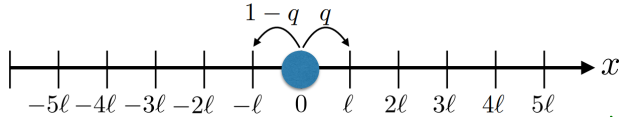
$$p = q = \frac{1}{2}$$

$$\frac{1}{2} \cdot \frac{1}{2}$$

1 Analysis of Biased Random Walk in 1D, Binomial Distribution

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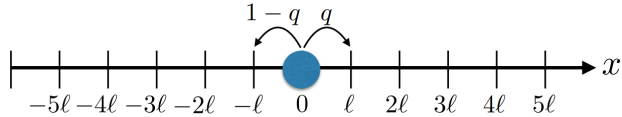


Q Write down the expression for prob. at i^{th} node of j^{th} level.

Q Prob to come back at 0 at i^{th} step.

1 Analysis of Biased Random Walk in 1D, Binomial Distribution

| 5



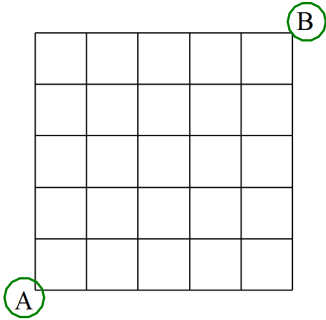
Draw the choice tree for **unbiased random walk**, derive binomial distribution:

1 Random Walkers on 2D Grid...

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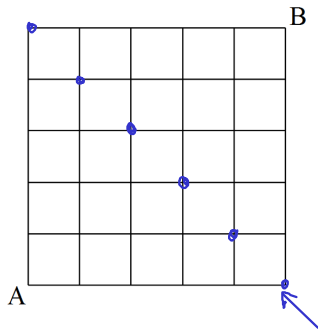
Alice moves = $\{ \text{right, up} \}$
Bob moves = $\{ \text{left, down} \}$

- Alice starts at point A, and each second, walks one edge right or up (if a point has two options, each direction has a 50



1 Random Walkers on 2D Grid...

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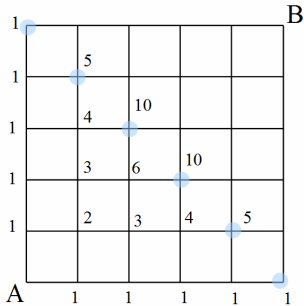
- Alice starts at point A, and each second, walks one edge right or up (if a point has two options, each direction has a 50)
- At the same time, Bob starts at point B, and each second he walks one edge left or down (if a point has two options, each direction has a 50)

• point where they can meet at same time (i.e., in this case 5 seconds).

1 Solution: Random Walkers on 2D Grid...

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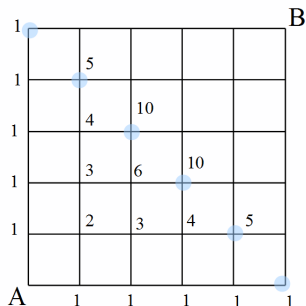
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- If Alice and Bob meet, they can meet only at the blue circles, which is after both have taken 5 steps

1 Solution: Random Walkers on 2D Grid...

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- If Alice and Bob meet, they can meet only at the blue circles, which is after both have taken 5 steps
- Number of ways they could have taken the 5 steps is 2^5 each, so combined, by product rule they can take $2^5 2^5 = 4^5 = 1024$ total paths!

$$\textcircled{A} = \{R, U\}$$

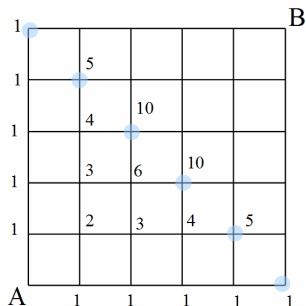
$$\underline{2} \quad \underline{2} \quad \underline{2} \quad \underline{2} \quad \underline{2} = 2^5$$

$$\textcircled{B} = \{L, D\}$$

$$\underline{2} \quad \underline{2} \quad \underline{2} \quad \underline{2} \quad \underline{2} = 2^5$$

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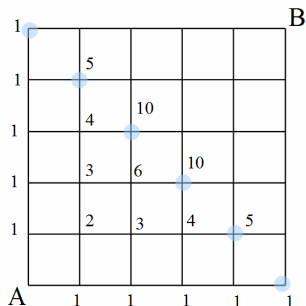
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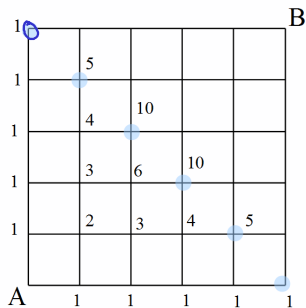
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 - Total ways Bob and Alice could meet at blue dots is square of binomial coefficients

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- If Alice and Bob meet, they can meet only at the blue circles, which is after both have taken 5 steps
- Number of ways they could have taken the 5 steps is 2^5 each, so combined, by product rule they can take $2^5 2^5 = 4^5 = 1024$ total paths!
- The number of ways Bob can reach the blue dots is given by binomial coefficients or Pascal's triangle! Same for Alice
 - Total ways Bob and Alice could meet at blue dots is square of binomial coefficients
- Hence, the total number of ways Bob and Alice could meet

$$1^2 + 5^2 + 10^2 + 10^2 + 5^2 + 1^2 = \underline{\underline{252}}$$

- The probability that they meet is

$$\frac{252}{1024} = 24.6\%$$

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2 Motivation for Conditional Probability with an Example ...

| 9

- **Experiment:** Throw two dice A and B simultaneously

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2 Motivation for Conditional Probability with an Example ...

- **Experiment:** Throw two dice A and B simultaneously
- **Event:** Odd number on first die
- **Question:** What is the probability of the event E ?

$$S = \left\{ \begin{array}{l} (1,1), (1,2), (1,3), \dots, (1,6) \\ (2,1), (2,2), \dots, (2,6) \\ \vdots \\ (6,1), (6,2), \dots, (6,6) \end{array} \right\} \quad |S| = 36$$

$$E = \left\{ \begin{array}{l} (1,1), (1,2), \dots, (1,6) \\ (3,1), (3,2), \dots, (3,6) \\ (5,1), (5,2), \dots, (5,6) \end{array} \right\}$$

$$|E| = 18$$
$$P(E) = \frac{18}{36} = \frac{1}{2}$$

- [illegible]

2 Motivation for Conditional Probability with an Example ...

$$P(A|B) = \underline{\hspace{2cm}}$$

- **Experiment:** Throw two dice A and B simultaneously
- **Event:** Odd number on first die
- **Question:** What is the probability of the event E ?

$$A|B = \left\{ (1,2), (1,4), (1,6), (3,2), (3,4), (3,6), (5,2), (5,4), (5,6) \right\}$$

- **Event:** Odd number on first die A , given that even shows on die B

- **Question:** What is the probability of the event E ?

$$P(E) = \frac{9}{18} = \left(\frac{1}{2} \right)$$

$$E = \left\{ \begin{array}{l} (1,1), (1,2), \dots, (1,6) \\ (3,1), (3,2), \dots, (3,6) \\ (5,1), (5,2), \dots, (5,6) \end{array} \right\}$$

6×3

$$B = \frac{\text{even shows on die } B}{\left\{ \begin{array}{l} (1,2), (2,2), (3,2), \dots, (6,2) \\ (1,4), (2,4), (3,4), \dots, (6,4) \\ (1,6), (2,6), (3,6), \dots, (6,6) \end{array} \right\}}$$

Since given that B happens, B becomes sample space!

2 Motivation for Conditional Probability with an Example with Dice...

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Question

Roll two 6-sided dice, yielding values D_1 and D_2 . Let E be the event $D_1 + D_2 = 4$. What is $P(E)$?

$$E : D_1 + D_2 = 4.$$

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

$$|S| = 36$$

$$E = \{ (1,3), (2,2), (3,1) \}$$

$$|E| = 3$$

$$Pr(E) = \frac{3}{36} = \frac{1}{12}$$

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$$P(E) =$$

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Question

Roll two 6-sided dice, yielding values D_1 and D_2 . Let E be the event $D_1 + D_2 = 4$. Let F be the event $D_1 = 2$. What is $P(E, \text{ given } F \text{ already observed})$?

• $|S| =$

• $E =$

$$E|F = \{(2, 2)\}$$

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

$P(E) =$

$F : D_1 = 2$

$$P(E|F)$$

$$P_1(E|F) = \frac{1}{6}$$

New sample space, because F is observed!

Question

Roll two 6-sided dice, yielding values D_1 and D_2 . Let E be the event $D_1 + D_2 = 4$. What is $P(E)$?

- $|S| =$

- $E =$

$$P(E) =$$

$$P(E|F) =$$

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3 Define Conditional Probability...

Definition of Conditional Probability

The **conditional probability** of E given F is the probability that E occurs given that F has already occurred. This is known as conditioning on F .

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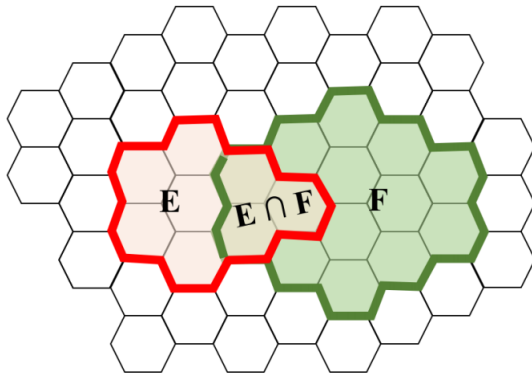
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- **Sample space** is **all** possible outcomes consistent with F (i.e., $S \cap F$)
- **Event** is **all** outcomes in E consistent with F (i.e., $E \cap F$)

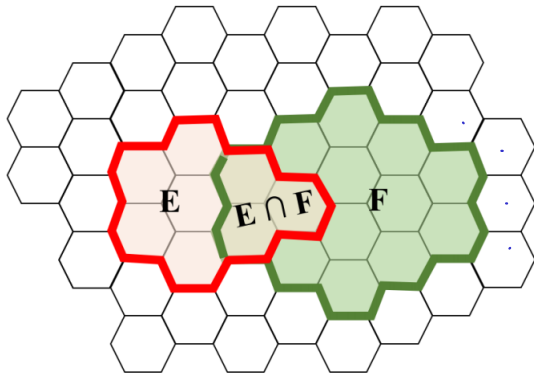


With equally likely outcomes:

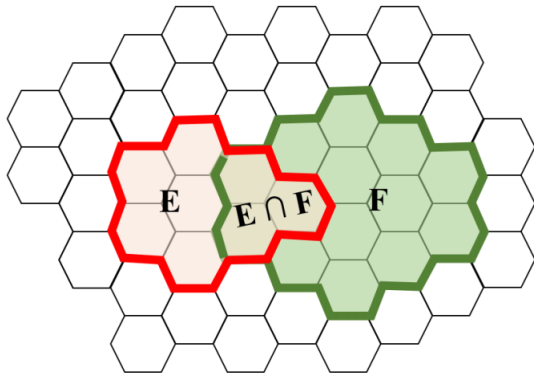
$$P(E|F) = \frac{|E \cap F|}{|S \cap F|} = \frac{|E \cap F|}{|F|}$$



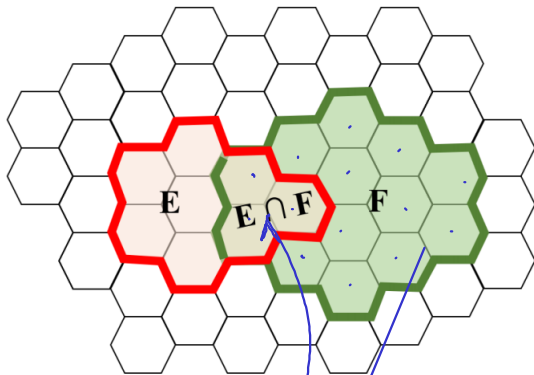
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- Question: What is $P(E | F)$? \triangleright



- **Question:** What is $P(E)$? Here $P(E) = \frac{8}{50} \approx 0.16$
- **Question:** What is $P(E | F)$? Here $P(E | F) = \frac{3}{14} \approx 0.21$

3 Probability of Receiving Spam Emails...

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Email Spam Conditional Probability Problem

24 emails are sent, 6 each to 4 users.

- 10 of the 24 emails are spam.
- All possible outcomes are equally likely.

Question-1

Let event E = user 1 receives 3 spam emails. What is $P(E)$?

3 Probability of Receiving Spam Emails...

Email Spam Conditional Probability Problem

24 emails are sent, 6 each to 4 users.

- 10 of the 24 emails are spam.
- All possible outcomes are equally likely.

Handwritten notes and calculations:

6 6 6 6

3 spam emails

$P(E) = ?$

$\frac{\binom{10}{3}}{\binom{24}{6}}$

$\frac{\binom{10}{3} \binom{14}{3}}{\binom{24}{6}}$

Question-1

Let event E = user 1 receives 3 spam emails. What is $P(E)$?

Question-2

Let event F = user 2 receives 6 spam emails. What is $P(E|F)$?

Conditional Probability Implies Chain Rule...

$$\left\{ P(E|F) = \frac{P(E \cap F)}{P(F)} \right\} \Rightarrow \underbrace{P(E \cap F)} = P(F)P(E|F)$$

These hold even when outcomes are not equally likely!

Conditional Probability Implies Chain Rule...

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

These hold even when outcomes are not equally likely!

Law of Total Probability (Theorem)

$$P(E) = P(E | F)P(F) + P(E|F^c)P(F^c)$$

3 Law of Total Probability...

Recall
Axiom 3

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

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Conditional Probability Implies Chain Rule...

$$P(E|F) = \frac{P(E \cap F)}{P(F)} \Rightarrow \underline{P(E \cap F) = P(F)P(E|F)}$$

These hold even when outcomes are not equally likely!

Law of Total Probability (Theorem)

$$P(E) = \underline{P(E|F)P(F)} + \underline{P(E|F^c)P(F^c)}$$

Proof:

$$E = \underbrace{(E \cap F)}_{\text{mutually exclusive}} \cup \underbrace{(E \cap F^c)}_{\text{mutually exclusive}}$$

From Axiom 3.

$$P(E) = P(E \cap F) + P(E \cap F^c) = \underline{P(E|F)P(F)} + \underline{P(E|F^c)P(F^c)}$$

Using these

3 Compute $P(E)$ from $P(E|F)$ Using Probability Tree...

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Problem

Flips a fair coin.

3 Compute $P(E)$ from $P(E|F)$ Using Probability Tree...

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Flips a fair coin.

- If heads: roll a fair 6-sided die.

3 Compute $P(E)$ from $P(E|F)$ Using Probability Tree...

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Problem

Flips a fair coin.

- If heads: roll a fair 6-sided die.
- Else: roll a fair 3-sided die.

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| 16

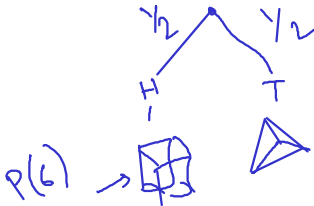
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Flips a fair coin.

- If heads: roll a fair 6-sided die.
- Else: roll a fair 3-sided die.

You win if you roll a 6. What is $P(\text{winning})$?

Solution using probability tree:



$$\frac{1}{2} \cdot \frac{1}{6} = \frac{1}{12}$$

3 Compute $P(E)$ from $P(E|F)$ Using Total Probability...

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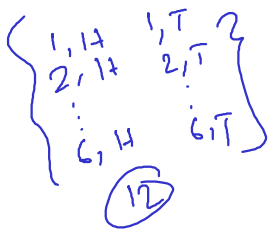
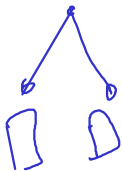
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Solution using total probability:



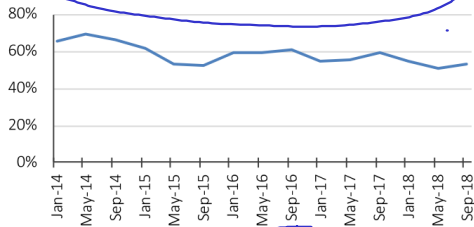
$E, F = \{H, T\}$
 $E = \{6\}$

$$P(E) = P(E|F)P(F) + P(E|F^c)P(F^c)$$
$$= \frac{2}{12} \cdot \frac{1}{2} + 0$$
$$= \frac{1}{12}$$

3 Bayes Theorem. Why?

Bayesian Inference
Machine Learning

Spam volume as percentage of total email traffic worldwide



$$P(A|B) \leftrightarrow P(B|A)$$

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Data

Features → Labels

Samples



Cat

dog

Known in a supervised ML.

$$P(\text{features} | \text{label})$$

Prediction Problem

$$P(\text{Features} | \text{label})$$

$$P(\text{label} | \text{feature})$$

want to know!