

Dr. Denton Bobeldyk

CIS 365 Artificial Intelligence

Uncertainty in AI

Week in Review

Blackboard Check-in

Uncertainty in *AI*

- ❖ Methods to cope with uncertainty:
 - ❖ Fuzzy Logic
 - ❖ Probability Theory

Uncertainty in AI - Fuzzy Sets

- ❖ Crisp Sets:
 - ❖ Standing = $\{x \mid x \text{ is a student standing in your class}\}$
 - ❖ Sitting = $\{y \mid y \text{ is a student sitting in your class}\}$
- ❖ The intersection of the two above sets is an empty set. A student belongs to one or the other.

$$\textit{Standing} \cap \textit{Sitting} = \emptyset$$

Uncertainty in AI - Fuzzy Sets

Notation Note:

x such that x is a student standing in your class
used for set notation

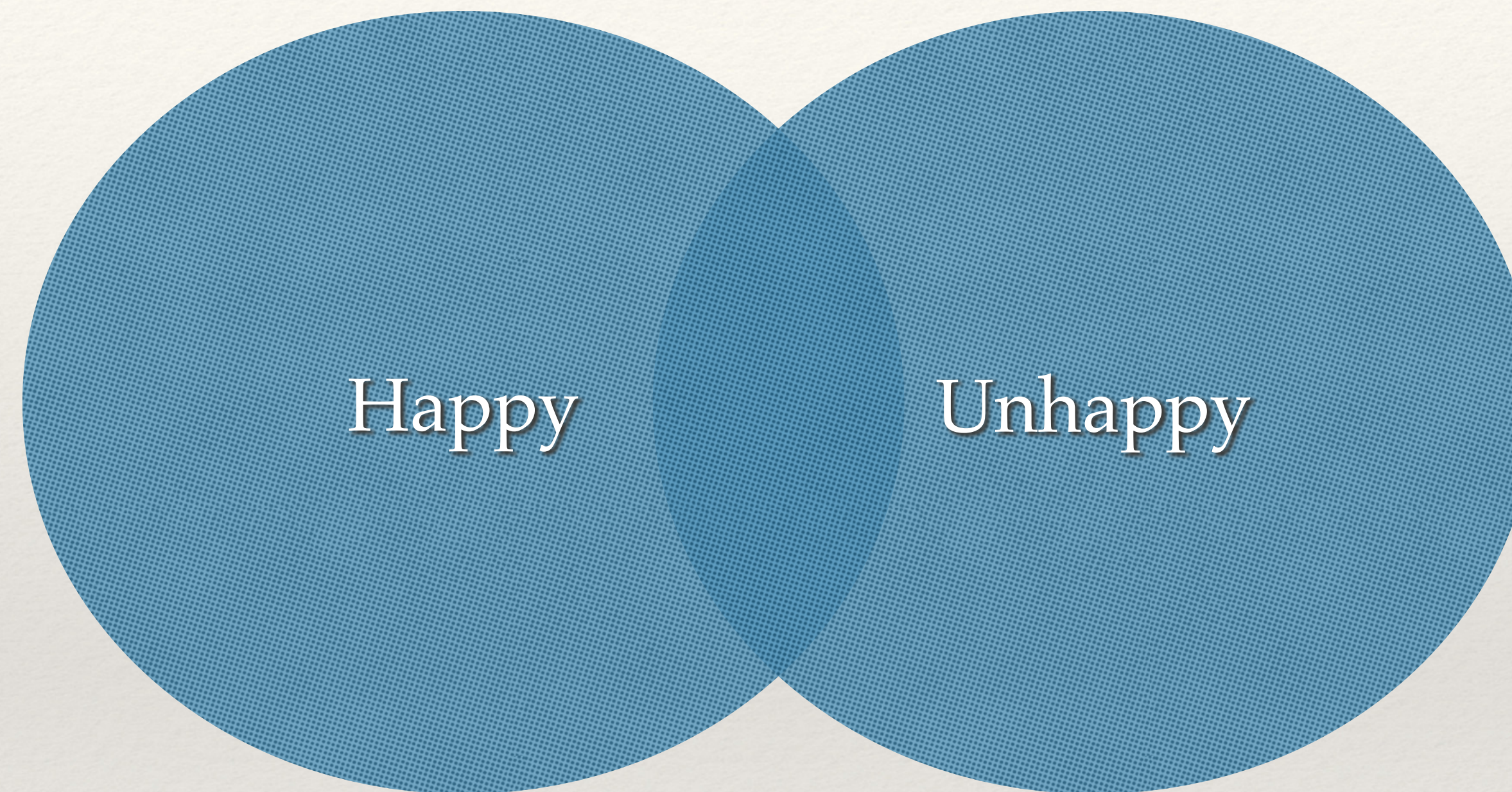
- ❖ Crisp Sets:
 - ❖ Standing = $\{x \mid x \text{ is a student standing in your class}\}$
 - ❖ Sitting = $\{y \mid y \text{ is a student sitting in your class}\}$
- ❖ The intersection of the two above sets is an empty set. A student belongs to one or the other.

$$\textit{Standing} \cap \textit{Sitting} = \emptyset$$

Fuzzy Sets

- ❖ Are you happy with your job?
- ❖ Are you unhappy with your job?

Fuzzy Sets



It is possible to be Happy and Unhappy about your job at the same time

Uncertainty in AI

- ❖ What about concepts like ‘tallness’? Is someone that is 5 foot, someone that would be considered tall?

Uncertainty in AI

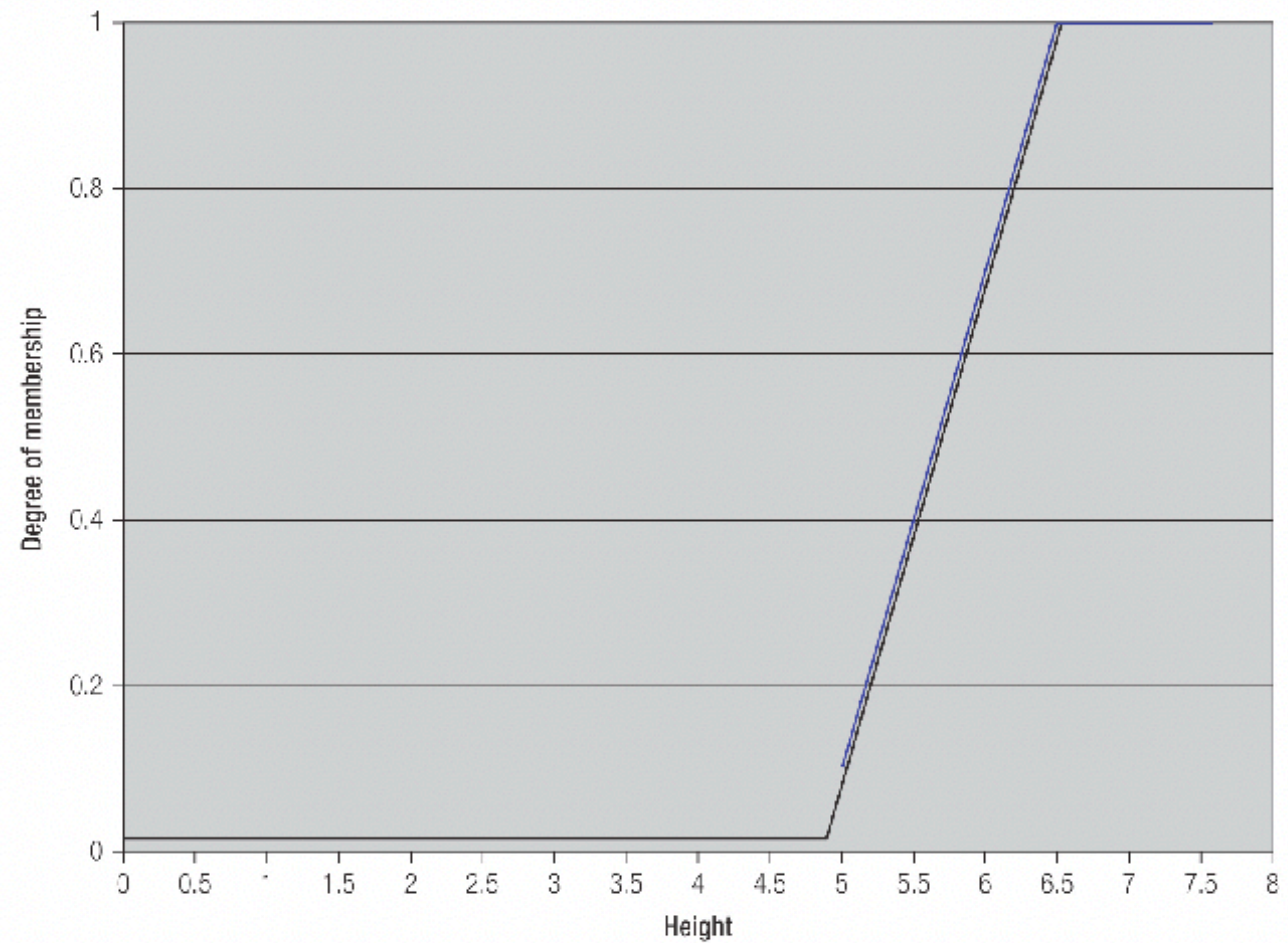


Figure 8.2

Membership function for the set of tall people.

Probability Theory

Probability Theory

- ❖ Sample Space for an experiment S is the set of all possible outcomes

Probability Theory

- ❖ Sample space of flipping one coin = $\{H, T\}$
- ❖ Sample space of flipping two coins = $\{(H,H), (H,T), (T, H), (T,T)\}$

Probability Theory

- ❖ Sample space of flipping one coin = {H, T}
- ❖ Probability of Heads = 0.5
 - ❖ $P(H) = \{\text{H}, T\} = 0.5$
- ❖ Probability of Tails = 0.5
 - ❖ $P(T) = \{H, \text{T}\} = 0.5$

Probability Theory

- ❖ Sample space of flipping two coins = $\{(H,H), (H,T), (T, H), (T,T)\}$
- ❖ Probability of outcomes:
 - ❖ $P(H,H) = \{(H,H), (H,T), (T, H), (T,T)\} = 1/4 = 0.25$
 - ❖ $P(H, T) = \{(H,H), (H,T), (T, H), (T,T)\} = 1/4 = 0.25$
 - ❖ $P(T, H) = \{(H,H), (H,T), (T, H), (T,T)\} = 1/4 = 0.25$
 - ❖ $P(T, T) = \{(H,H), (H,T), (T, H), (T,T)\} = 1/4 = 0.25$

Probability Theory

- ❖ Rolling a single die
 - ❖ What's the sample set?
 - ❖ What's the probability that each event will occur?

Probability Theory

- ❖ Rolling a single die
 - ❖ What's the sample set? $\{1, 2, 3, 4, 5, 6\}$
 - ❖ What's the probability that each event will occur?
 - ❖ $P(1) = \{1, 2, 3, 4, 5, 6\} = 1 / 6 = 16.67\%$
 - ❖ $P(2) = \{1, 2, 3, 4, 5, 6\} = 1 / 6 = 16.67\%$
 - ❖ ...

Probability Theory

- ❖ Rolling 2 dice
 - ❖ What's the sample set?
 - ❖ What's the probability that each event will occur?

Probability Theory

- ❖ Rolling 2 dice
 - ❖ What's the sample set?
 - ❖ $\{(1,1), (1,2), (1,3), (1,4), (1,5), (1,6), (2,1), (2,2), (2,3), (2,4), (2,5), (2,6),$
 - ❖ $(3,1), (3,2), (3,3), (3,4), (3,5), (3,6), (4,1), (4,2), (4,3), (4,4), (4,5), (4,6),$
 - ❖ $(5,1), (5,2), (5,3), (5,4), (5,5), (5,6), (6,1), (6,2), (6,3), (6,4), (6,5), (6,6)\}$
 - ❖ What's the probability that each event will occur?

Probability Theory

❖ Rolling 2 dice:

❖ $P(1,1) =$

❖ $\{(1,1), (1,2), (1,3), (1,4), (1,5), (1,6), (2,1), (2,2), (2,3), (2,4), (2,5), (2,6),$

❖ $(3,1), (3,2), (3,3), (3,4), (3,5), (3,6), (4,1), (4,2), (4,3), (4,4), (4,5), (4,6),$

❖ $(5,1), (5,2), (5,3), (5,4), (5,5), (5,6), (6,1), (6,2), (6,3), (6,4), (6,5), (6,6)\}$

❖ $P(1,1) = P(\text{Event} / \text{Number of items in sample space}) = 1 / 36$

Probability Theory

- ❖ Rolling a 2 dice:
 - ❖ $P(\text{Sum of two dice} = 4)$
 - ❖ $\{(1,1), (1,2), (1,3), (1,4), (1,5), (1,6), (2,1), (2,2), (2,3), (2,4), (2,5), (2,6),$
 - ❖ $(3,1), (3,2), (3,3), (3,4), (3,5), (3,6), (4,1), (4,2), (4,3), (4,4), (4,5), (4,6),$
 - ❖ $(5,1), (5,2), (5,3), (5,4), (5,5), (5,6), (6,1), (6,2), (6,3), (6,4), (6,5), (6,6)\}$
 - ❖ $P(\text{Sum of two dice} = 4) = ??$

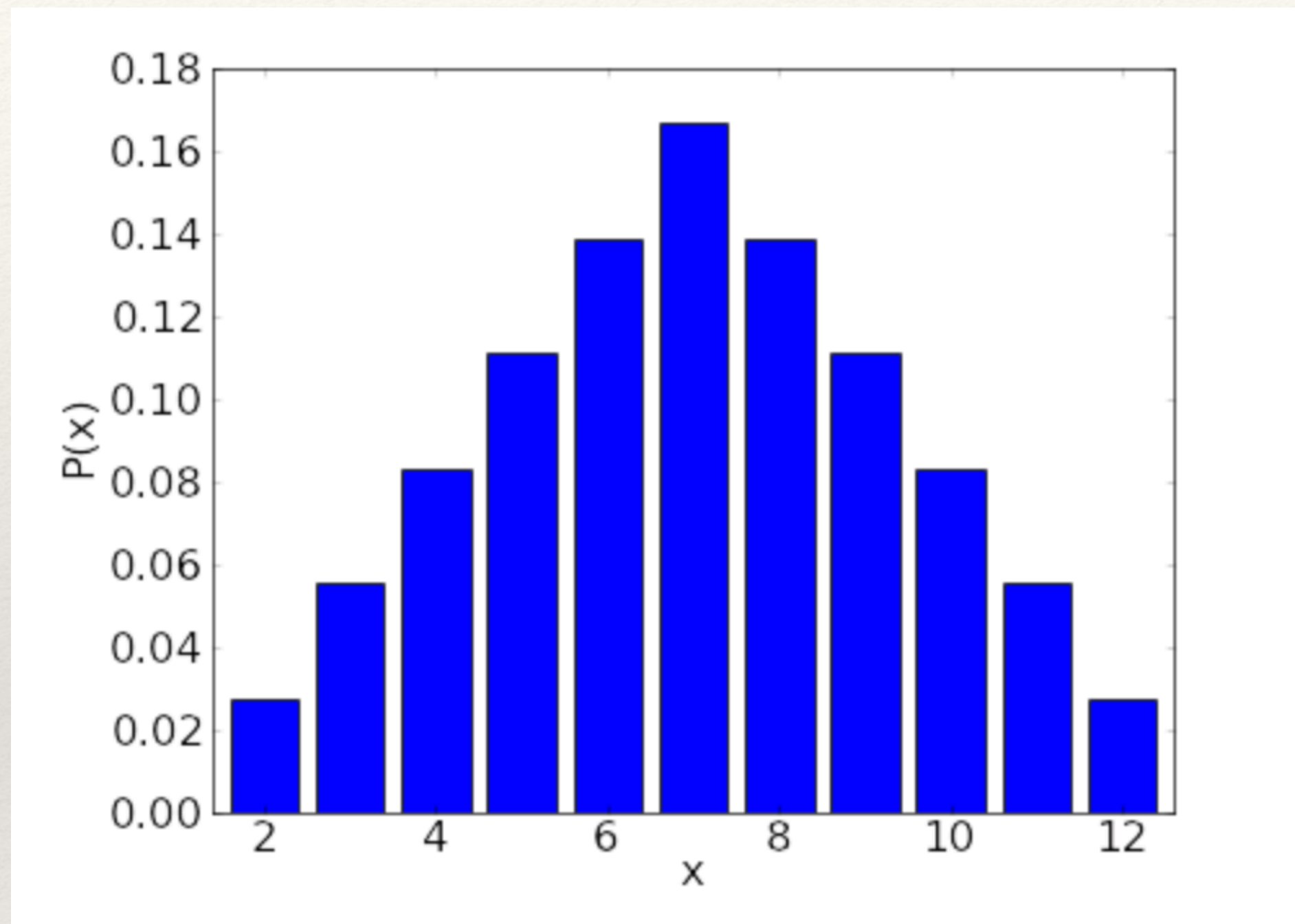
Probability Theory

- ❖ Rolling a 2 dice:
 - ❖ $P(\text{Sum of two dice} = 4)$
 - ❖ $\{(1,1), (1,2), (1,3), (1,4), (1,5), (1,6), (2,1), (2,2), (2,3), (2,4), (2,5), (2,6),$
 - ❖ $(3,1), (3,2), (3,3), (3,4), (3,5), (3,6), (4,1), (4,2), (4,3), (4,4), (4,5), (4,6),$
 - ❖ $(5,1), (5,2), (5,3), (5,4), (5,5), (5,6), (6,1), (6,2), (6,3), (6,4), (6,5), (6,6)\}$
 - ❖ $P(\text{Sum of two dice} = 4) = 3 / 36 = 8.3\%$

Probability Theory

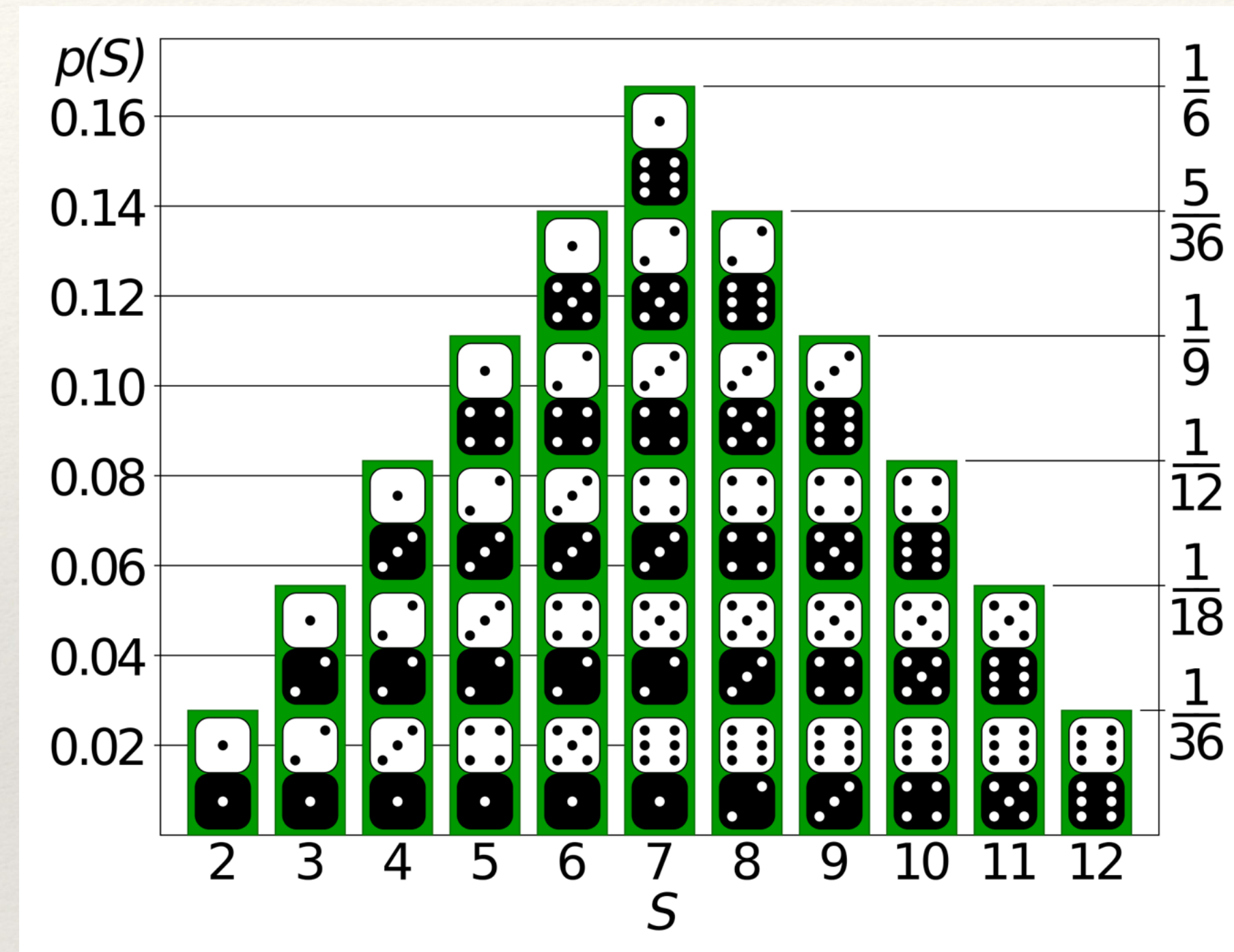
- ❖ Probability Distribution for the sum of two dice is?
- ❖ If we know the probability distribution, we can determine:
 - ❖ What sum has the highest probability?
 - ❖ What sum has the lowest probability?
 - ❖ Cumulative distribution (i.e., what's the probability of rolling **less** than 8)

Probability Distribution



https://mathinsight.org/image/two_dice_distribution

Probability Distribution



<https://math.stackexchange.com/questions/1204396/why-is-the-sum-of-the-rolls-of-two-dices-a-binomial-distribution-what-is-define>

Probability Distribution

Three basic axioms of probability:

1. For any event E : $P(E) \geq 0$
2. $P(S) = 1$ where S is the sample space of an experiment
3. If the events E_1 and E_2 are mutually exclusive then $P(E_1 \cap E_2) = P(E_1) + P(E_2)$

Class Exercise

- ❖ Probability Distribution for 1d6 and 2d6 have been covered
- ❖ What is the sample space for:
 - ❖ Rolling 1d4
 - ❖ Rolling 2d4
- ❖ What is the probability distribution:
 - ❖ Rolling 1d4
 - ❖ Rolling 2d4
 - ❖ Rolling 3d4

Bayesian Classifier

- ❖ Probability distribution
 - ❖ Rolling 1d4
 - ❖ Rolling 2d4
 - ❖ Rolling 3d4
- ❖ Given the following numbers, what is the probability that it came from each of the above distributions?
 - ❖ 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12

Bayesian Classifier

- ❖ Distributions: 1d4, 2d4, 3d4
- ❖ Given the following numbers, what is the probability that it came from each of the above distributions?
 - ❖ 1:
 - ❖ Probability of 1 Given 1d4 = ?
 - ❖ Probability of 1 Given 2d4 = ?
 - ❖ Probability of 1 Given 3d4 = ?

Bayesian Classifier

- ❖ Distributions: $1d4$, $2d4$, $3d4$
- ❖ Given the following numbers, what is the probability that it came from each of the above distributions?
 - ❖ 2:
 - ❖ Probability of 2 Given $1d4 = ?$
 - ❖ Probability of 2 Given $2d4 = ?$
 - ❖ Probability of 2 Given $3d4 = ?$

Bayesian Classifier

- ❖ Distributions: 1d4, 2d4, 3d4
- ❖ Given the following numbers, what is the probability that it came from each of the above distributions?
 - ❖ Complete for each of the following numbers: 1 - 12

Completed Answer Next Slide

Probability Distribution

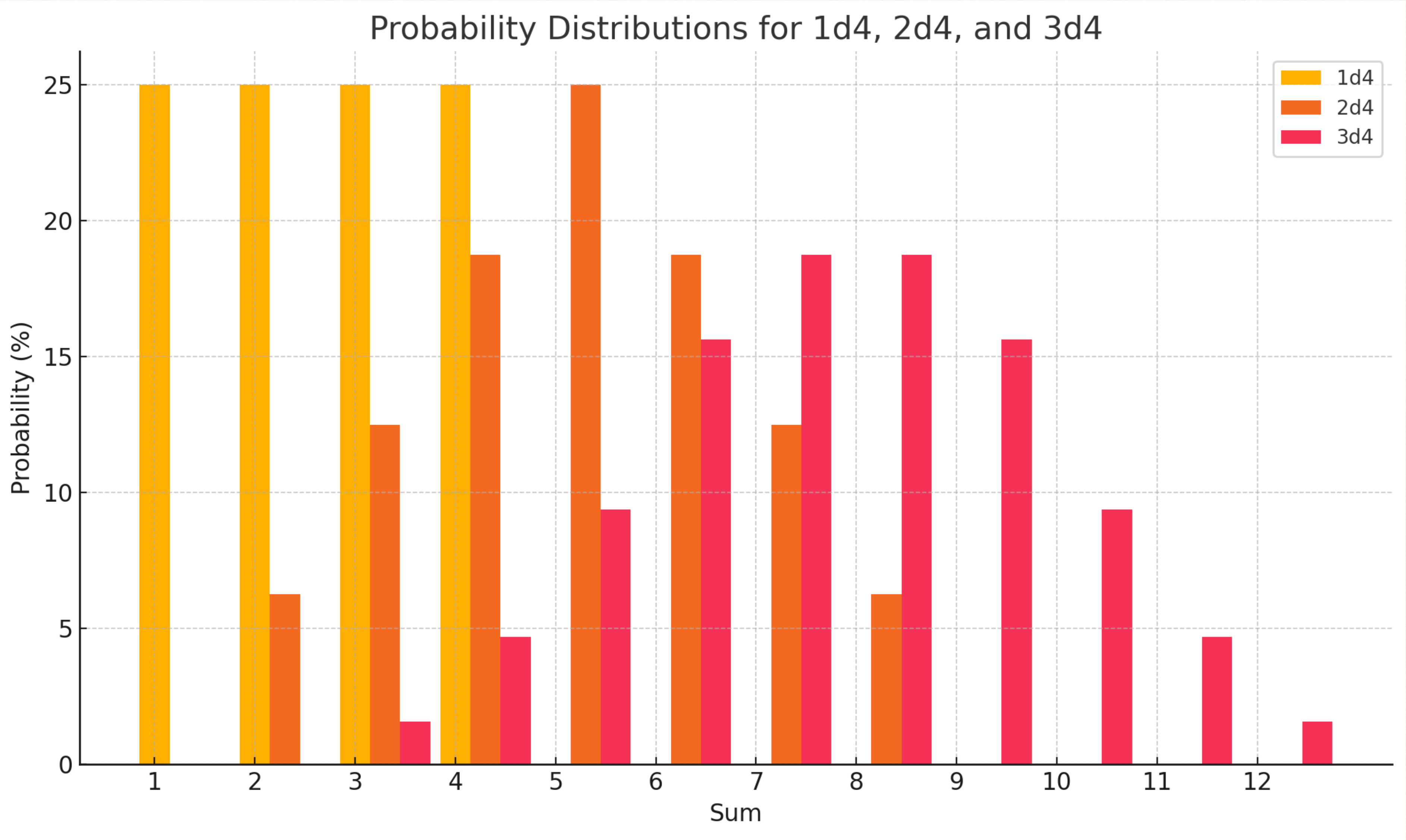
	1d4	2d4	3d4
1	25	0	0
2	25	6.25	0
3	25	12.5	1.6
4	25	18.75	4.7
5	0	25	9.4
6	0	18.75	15.6
7	0	12.5	18.8
8	0	6.25	18.8
9	0	0	15.6
10	0	0	9.4
11	0	0	4.7
12	0	0	1.6

Which distribution did it come from?

- ❖ Can we determine if given an outcomes what distribution it came from?

Bayesian Classifier - Which Distribution Generated the outcome?

Outcome	1d4	2d4	3d4
1	25	0	0
2	25	6.25	0
3	25	12.5	1.6
4	25	18.75	4.7
5	0	25	9.4
6	0	18.75	15.6
7	0	12.5	18.8
8	0	6.25	18.8
9	0	0	15.6
10	0	0	9.4
11	0	0	4.7
12	0	0	1.6



Please note the X axis is slightly misaligned

D4 Distributions

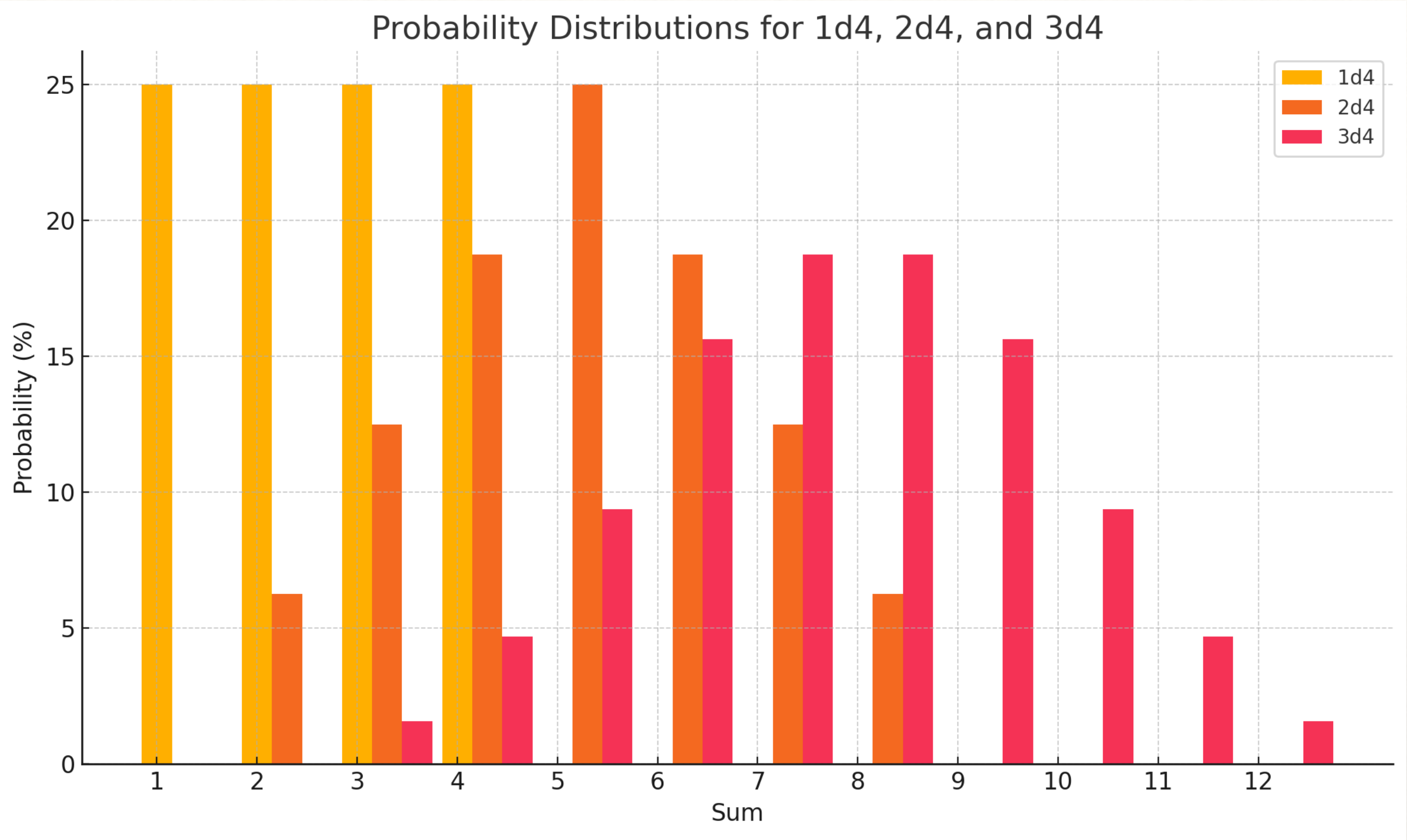
- ❖ Does a formula exist, if we know the number of d4 we are using, what the probability distribution is?

D4 Distributions

- ❖ If a formula did exist, we could easily calculate the probability of an outcome given the number of d4's that were used.
- ❖ For example, calculateD4Distribution(X, numberOfD4s)
 - ❖ calculateD4Distribution(1, 1) = .25
 - ❖ calculateD4Distribution(1,2) = 0
 - ❖ calculateD4Distribution(2,2) = 6.25

The formula would return the values to the right

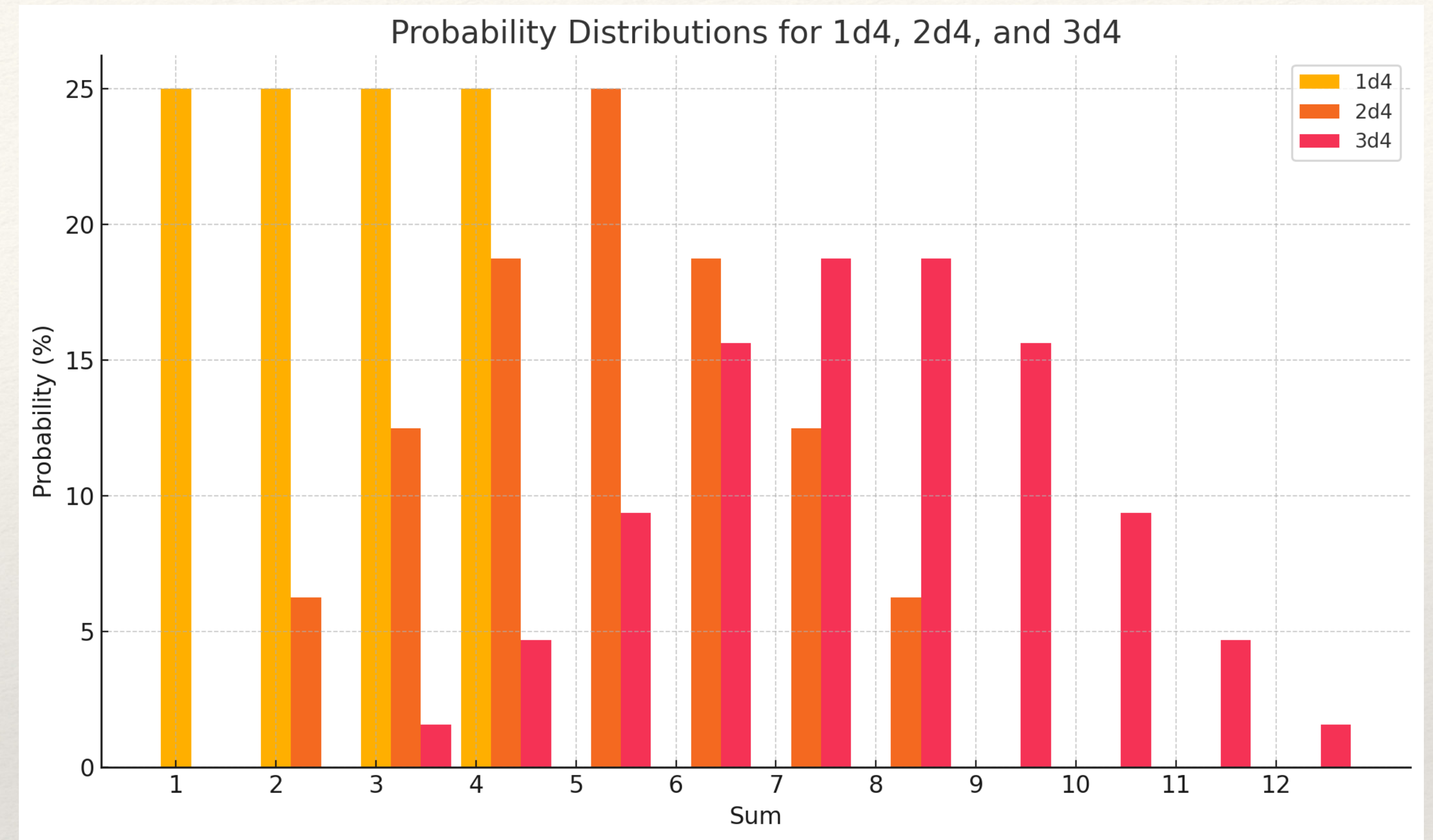
- ❖ `calculateD4Distribution(1, 1) = .25`
- ❖ `calcuatD4Distribution(1,2) = 0`
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Please note the X axis is slightly misaligned

The formula would return the values to the right

- ❖ `calculateD4Distribution(1, 1) = .25`
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- ❖ `calcuatateD4Distribution(2,2) = 6.25`



Please note the X axis is slightly misaligned

No Parametric distribution like this exists!!

We could however write a program that statically returns a value based on a static mapping

Gaussian Distribution

- ❖ No formula exists to determine the parametric distribution based on the number of d4's, however parametric distributions do exist, and one of those is the Gaussian distribution.
- ❖ Gaussian distribution is determined by the mean (average sample) and standard deviation (how spread out the values are)

Gaussian Distribution

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

x is the variable

σ is the standard deviation for the distribution

μ is the mean for the distribution

π is a pre-defined constant

e is a pre-defined constant

Gaussian Distribution

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

Can relatively easily convert the above into a function that programmatically determines the probability of x occurring given a gaussian distribution defined by 2 parameters: μ, σ

Gaussian Distribution

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

calculateDistribution(x, average, standardDeviation)

Gaussian Distribution

What if we have more than one feature?
Pack all of the features into a 'feature vector'

Please note the X axis is slightly misaligned

Gaussian Distribution

What if we have more than one feature?
Pack all of the features into a 'feature vector'

For example:
Height of an athlete

Just One feature!!

More than one feature would be:
Height, Weight, Shoe size

Three Features!!

Gaussian Distribution

What if we have more than one feature?
Pack all of the features into a 'feature vector'

For example:
Height of an athlete

Data Representation

[70]

More than one feature would be:
Height, Weight, Shoe size

[70,200,13]

Multivariate Gaussian Distribution

$$f(\mathbf{x}) = \frac{1}{\sqrt{(2\pi)^k |\boldsymbol{\Sigma}|}} \exp \left(-\frac{1}{2} (\mathbf{x} - \boldsymbol{\mu})^\top \boldsymbol{\Sigma}^{-1} (\mathbf{x} - \boldsymbol{\mu}) \right)$$

Please note the X axis is slightly misaligned


```
import numpy as np
```

```
def multivariate_gaussian(x, mu, covarianceMatrix):
```

```
    k = len(mu)
```

```
    covarianceMatrix_det = np.linalg.det(covarianceMatrix)
```

```
    covarianceMatrix_inv = np.linalg.inv(covarianceMatrix)
```

```
    norm_factor = 1.0 / np.sqrt((2 * np.pi) ** k * covarianceMatrix_det)
```

```
    diff = x - mu
```

```
    exponent = -0.5 * np.dot(np.dot(diff.T, covarianceMatrix_inv), diff)
```

```
    return norm_factor * np.exp(exponent)
```

Parameters:

x (np.array): The k-dimensional data point (column vector).

mu (np.array): The mean vector (column vector).

covarianceMatrix (np.array): The covariance matrix (k x k matrix).


```
import numpy as np
```

```
def multivariate_gaussian(x, mu, covarianceMatrix):
```

```
    k = len(mu)
```

```
    covarianceMatrix_det = np.linalg.det(covarianceMatrix)
```

```
    covarianceMatrix_inv = np.linalg.inv(covarianceMatrix)
```

```
    norm_factor = 1.0 / np.sqrt((2 * np.pi) ** k * covarianceMatrix_det)
```

```
    diff = x - mu
```

```
    exponent = -0.5 * np.dot(np.dot(diff.T, covarianceMatrix_inv), diff)
```

```
    return norm_factor * np.exp(exponent)
```

```
# Example usage:
```

```
x = np.array([1.0, 2.0])
```

```
mu = np.array([0.0, 0.0])
```

```
covMatrix = np.array([[1.0, 0.5], [0.5, 1.0]])
```

```
# Calculate the multivariate Gaussian value
```

```
result = multivariate_gaussian(x, mu, covMatrix)
```

```
print(result)
```


End Slide