## Random Variables and Probability Distributions

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### Random Variable

- A Random variable is used to map the outcome of a random process to numbers.
- A Random Variable X takes on a defined set of values with different probabilities.
  - For example, if you roll a die, the outcome is random(not fixed) and there are 6 possible, each of which occur with probability one-sixth.

| <b>Event</b> | Probability |
|--------------|-------------|
|--------------|-------------|

| X | p(x) = P(X = x) |
|---|-----------------|
| 1 | 1/6             |
| 2 | 1/6             |
| 3 | 1/6             |
| 4 | 1/6             |
| 5 | 1/6             |
| 6 | 1/6             |

Figure: Rolling of a die

#### Random Variables

It can be discrete or continuous

- **Discrete**: random variables have a countable number of outcomes.
  - For example yes/no, dice, counts etc.
- Continuous: random variables have an infinite continuum of possible values.
  - For example blood pressure, weight, the speed of a car, the real numbers from 1 to 6 etc.

## **Probability Functions**

- A probability functions maps the possible values of x against their respective probabilities of occurrence, p(x)
- p(x) is a number from 0 to 1.0.
- The area under a probability function is always 1.

## **Probability Density Functions**

- A probability density function is most commonly associated with absolutely continuous univariate distributions.
- A random variable X has density  $f_X$ , where  $f_X$  is a non-negative Lebesgue-integrable function, if:

$$\Pr[a \le X \le b] = \int_a^b f_X(x) \, dx.$$

• If  $F_X$  is the cumulative distribution function of X, then:

$$F_X(x) = \int_{-\infty}^x f_X(u) du, \text{ and (if } f_X \text{ is continuous at } x)$$

$$f_X(x) = \frac{d}{dx} F_X(x).$$

## Discrete example: roll of a die

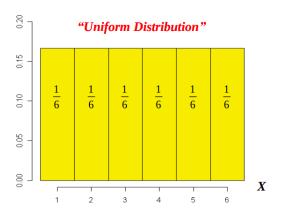


Figure: Rolling of a die

## Discrete example: roll of a die

### **Event Probability**

| X | p(x) = P(X = x) |
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| 1 | 1/6             |
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Figure: Rolling of a die

## Cumulative distribution function (CDF)

- The cumulative distribution function of a real-valued random variable X is the function given by:  $F_X(x) = P(X \le x)F_X(x) = P(X \le x)$  where the right-hand side represents the probability that the random variable X takes on a value less than or equal to x.
- The probability that X lies in the semi-closed interval (a, b], where a < b, is therefore</li>

$$P(a < X \le b) = F_X(b) - F_X(a)P(a < X \le b) = F_X(b) - F_X(a)$$

## Cumulative distribution function (CDF)

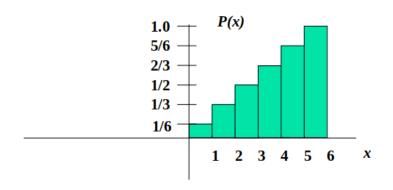


Figure: CDF example

## Cumulative distribution function (CDF)

| Х | $P(x \leq A)$      |
|---|--------------------|
| 1 | $P(x \le 1) = 1/6$ |
| 2 | $P(x \le 2) = 2/6$ |
| 3 | $P(x \le 3) = 3/6$ |
| 4 | $P(x \le 4) = 4/6$ |
| 5 | $P(x \le 5) = 5/6$ |
| 6 | $P(x \le 6) = 6/6$ |
|   |                    |

Table: CDF table

## Examples

• What's the probability that you roll a 3 or less?

What's the probability that you roll a 5 or higher?

## Examples

- What's the probability that you roll a 3 or less? Ans:  $P(x \le 3) = 1/2$
- ② What's the probability that you roll a 5 or higher? Ans :  $P(x \ge 5) = 1 P(x \le 4) = 1 2/3 = 1/3$

### Practice Problem

• Which of the following are probability functions?

$$f(x) = .25 \text{ for } x = 9,10,11,12$$

$$(x) = (x^2 + x + 1)/25 \text{ for } x = 0,1,2,3$$

### Practice Problem

• Which of the following are probability functions?

$$f(x) = .25 \text{ for } x = 9,10,11,12$$
 Ans : Yes

$$f(x) = (3-x)/2 \text{ for } x = 1,2,3,4$$
 Ans : No

• 
$$f(x) = (x^2 + x + 1)/25$$
 for  $x = 0,1,2,3$   
Ans : No

## Types of Discrete Distribution

There are a variety of discrete probability distributions that you can use to model different types of data.

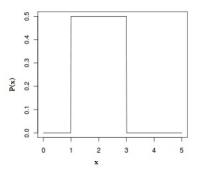
- Bernoulli distribution to model binary data, such as coin tosses.
- **Binomial distribution** is SUCCESS or FAILURE outcome in an experiment or survey that is repeated multiple times.
- Poisson distribution to model count data, such as the count of library book checkouts per hour.
- **Uniform distribution** to model multiple events with the same probability, such as rolling a die

#### Uniform distribution

- Uniform distributions are probability distributions with equally likely outcomes.
- This distribution is defined by two parameters, a and b:
  - a is the minimum.
  - b is the maximum.

The distribution is written as U(a,b).

• The following graph shows the distribution with a = 1 and b = 3:



### Uniform distribution

#### Expected Value and Variance

#### Expected Value

The expected value (i.e. the mean) of a uniform random variable X is:

$$E(X) = (1/2) (a + b)$$

This is also written equivalently as:

$$E(X) = (b + a) / 2.$$

"a" in the formula is the minimum value in the distribution, and "b" is the maximum value.

#### Variance

The variance of a uniform random variable is:

$$Var(x) = (1/12)(b-a)2$$

#### Poisson distribution

 The Poisson distribution is used to model the number of events occurring within a given time interval. The formula for the Poisson probability mass function is

$$p(x;\lambda) = \frac{e^{-\lambda}\lambda^x}{x!} for x = 0, 1, 2, \tag{1}$$

 $\lambda$  is the shape parameter which indicates the average number of events in the given time interval.

## Shape of Poisson Distribution

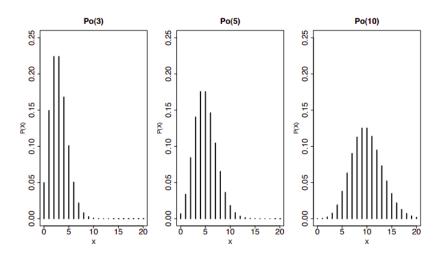


Figure: Poisson Distribution

## Shape of Poisson Distribution

We observe that the Poisson distributions

- are unimodal;
- **2** exhibit positive skew (that decreases as  $\lambda$  increases);
- **3** are centred roughly on  $\lambda$ ;
- have variance (spread) that increases as  $\lambda$  increases.

### Poisson distribution

Expected Value and Variance

#### Expected Value

The expected value (i.e. the mean) of a Poisson random variable X is:  $E(X) = \lambda$ 

#### Variance

The variance of a Poisson random variable is:

$$Var(x) = \sqrt{\lambda}$$

### Bernoulli Distribution

- The Bernoulli distribution is a discrete distribution having two possible outcomes labelled by  $\mathsf{n}=0$  and  $\mathsf{n}=1$  in which  $\mathsf{n}=1$  ("success") occurs with probability  $\mathsf{p}$  and  $\mathsf{n}=0$  ("failure") occurs with probability  $\mathsf{q}=1$ -p, where  $0<\mathsf{p}<1$ .
- It therefore has probability density function

$$p(n) = \begin{cases} 1 - p, & \text{for } n = 0 \\ p, & \text{for } n = 1 \end{cases}$$
 (2)

which can also be written

$$p(n) = p^{n}(1-p)^{1-n}.$$
 (3)

• The performance of a fixed number of trials with fixed probability of success on each trial is known as Bernoulli trial.



### Bernoulli distribution

#### Expected Value and Variance

#### Expected Value

The expected value (i.e. the mean) of a Bernoulli random variable  $\boldsymbol{X}$  is:

$$E(X) = p$$

#### Variance

The variance of a Bernoulli random variable is:

$$Var(x) = p(1-p)$$

### Binomial Distribution

- The binomial distribution gives the discrete probability distribution  $P_p(\mathsf{n/N})$  of obtaining exactly n successes out of N Bernoulli trials (where the result of each Bernoulli trial is true with probability p and false with probability  $\mathsf{q}{=}1\text{-p}$ ).
- The binomial distribution is therefore given by

$$P_p(n/N) = {N \choose n} p^n q^{N-n} = \frac{N!}{n!(N-n)!} p^n q^{N-n}$$

where  $\binom{N}{n}$  is a binomial coefficient.

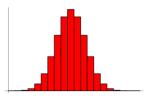


Figure: The above plot shows the distribution of n successes out of N=20 trials with p=q=1/2.

### Binomial distribution

#### Expected Value and Variance

#### Expected Value

The expected value (i.e. the mean) of a Binomial random variable X is:

$$E(X) = np$$

#### Variance

The variance of a Binomial random variable is:

$$Var(x) = np(1-p)$$

#### Continuous Distribution

#### Gaussian Distribution

- Continuous random variables are described with probability density function (pdfs) curves
- Normal pdfs are recognized by their typical bell-shape.

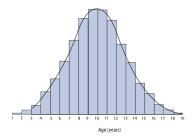


Figure: Age distribution of a pediatric population with overlying Normal pdf.

### Gaussian Distribution

• The general form of its probability density function is

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} \tag{4}$$

- The parameter  $\mu$  is the mean or expectation of the distribution (and also its median and mode), while the parameter  $\sigma$  is its standard deviation.
- The variance of the distribution is  $\sigma^2$
- A random variable with a Gaussian distribution is said to be normally distributed, and is called a normal deviate.

### Gaussian Distribution

#### Properties of a gaussian distribution

- The mean, mode and median are all equal.
- The curve is symmetric at the center (i.e. around the mean,  $\mu$ ).
- Exactly half of the values are to the left of center and exactly half the values are to the right.
- The total area under the curve is 1.

### Gaussian distribution

#### Expected Value and Variance

#### Expected Value

The expected value (i.e. the mean) of a Standard Gaussian random variable X is:

$$E(X) = 0$$

#### Variance

The variance of a Standard Gaussian random variable is:

$$Var(x) = 1$$

# The End