

10. Probability distributions

These notes consider the Chapter 10 of the [handbook](#) on various probability distributions.

Discrete distributions

For random variables with countable number of possible values.

Binomial distribution

See page 200 for proper definition.

Useful when a random variable X has exactly two exclusive possible outcomes (e.g. success/fail) with known probabilities $p \in [0, 1]$ (success) and $q = 1 - p$ (fail). For $n \in \mathbb{N}$ trials with $k = 0, 1, 2, \dots, n$ successes

$$X \sim \text{Bin}(n, p)$$

$$P(X = k) = \binom{n}{k} p^k q^{n-k}$$

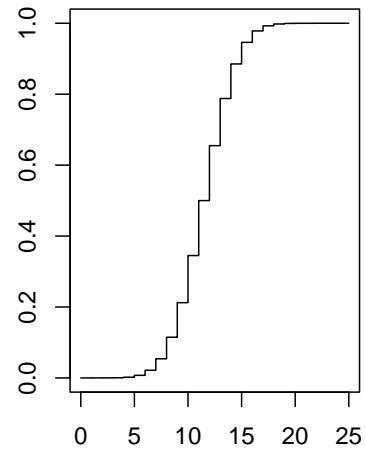
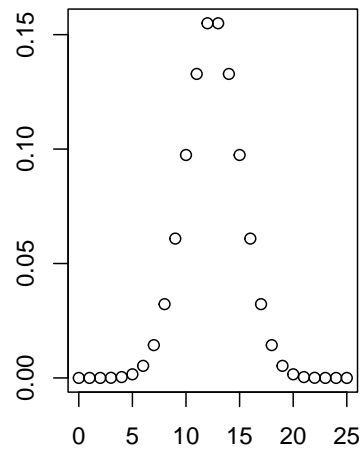
$$F(k) = P(X \leq k) = \sum_{i=0}^k \binom{n}{i} p^i q^{n-i}$$

In the following code examples, equal probabilities are assumed, $n = \text{size}$, $p = \text{prob} = 0.5$.

```
# Sequence for visualization
binomial_seq <- seq(0, 25, by = 1)

# Functions
binomial_pmf <- dbinom(x = binomial_seq, size = 25, prob = 0.5)
binomial_cdf <- pbinom(q = binomial_seq, size = 25, prob = 0.5)

# Plot
par(mfrow = c(1, 2))
plot(binomial_seq, binomial_pmf, ann = FALSE)
plot(binomial_seq, binomial_cdf, type = "S", ann = FALSE)
```



```
# Probability for exactly 3 successes out of 10 trials
dbinom(x = 3, size = 10, prob = 0.5)
```

```
## [1] 0.1171875
```

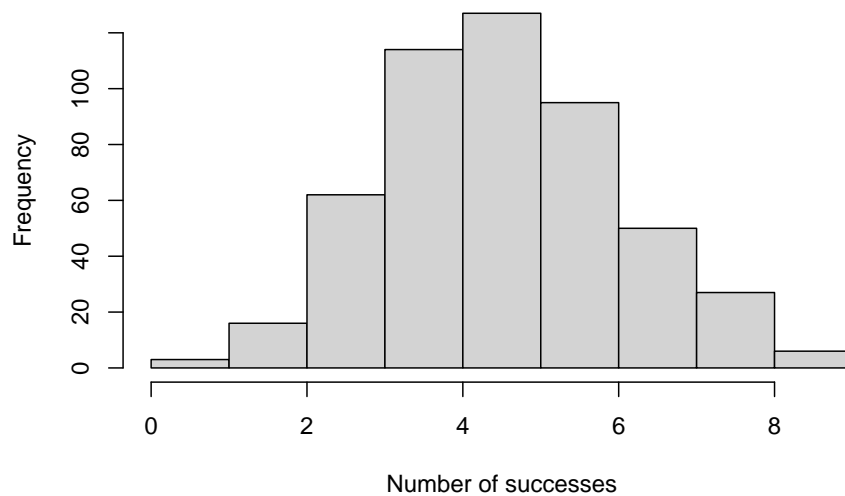
```
# Probability for up to 3 successes out of 10 trials
pbinom(q = 3, size = 10, prob = 0.5)
```

```
## [1] 0.171875
```

```
# Simulate 10 times how many successes there is using random numbers
rbinom(n = 10, size = 10, prob = 0.5)
```

```
## [1] 4 6 4 2 4 4 4 3 8 5
```

```
# With large enough n, expected value (np = 5) should become visible
rbinom500 <- rbinom(n = 500, size = 10, prob = 0.5)
hist(rbinom500, xlab = "Number of successes", ylab = "Frequency", main = NULL)
```



```
summary(rbinom(500))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  0.000   4.000   5.000   5.002   6.000   9.000
```

Poisson distribution

See page 204 for proper definition.

Useful when estimating amounts in random processes where the expected value (λ) is known

$$X \sim Poi(\lambda)$$

$$P(X = k) = \frac{\lambda^k e^{-\lambda}}{k!}, \quad k = 0, 1, 2, \dots, n$$

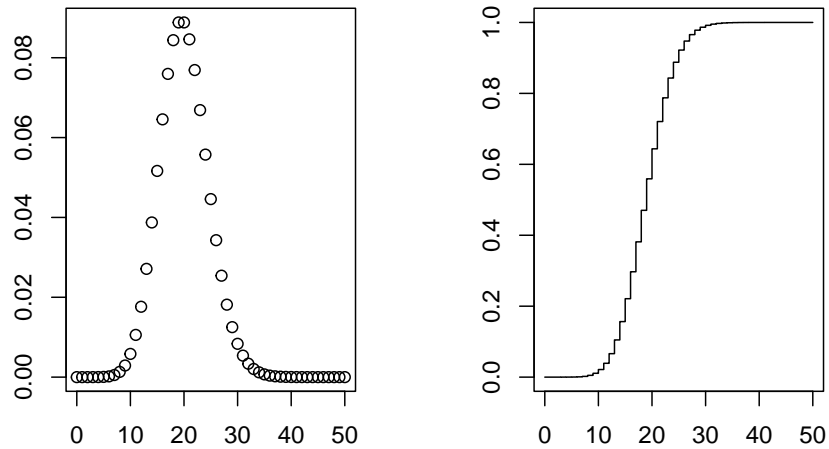
Note that the Poisson distribution can be used to approximate the binomial distribution when n is large and p is small. In this case, $\lambda = np$ and

$$P(X = k) = \frac{(np)^k e^{-np}}{k!}$$

```
# Sequence for visualization
poisson_seq <- seq(0, 50, by = 1)

# Functions
poisson_pmf <- dpois(x = poisson_seq, lambda = 20)
poisson_cdf <- ppois(q = poisson_seq, lambda = 20)
```

```
# Plot
par(mfrow = c(1, 2))
plot(poisson_seq, poisson_pmf, ann = FALSE)
plot(poisson_seq, poisson_cdf, type = "S", ann = FALSE)
```



```
# A bridge is crossed by 12 people per minute, on average.
# What is the probability that exactly 16 people crosses it in a minute?
dpois(x = 16, lambda = 12)
```

```
## [1] 0.05429334
```

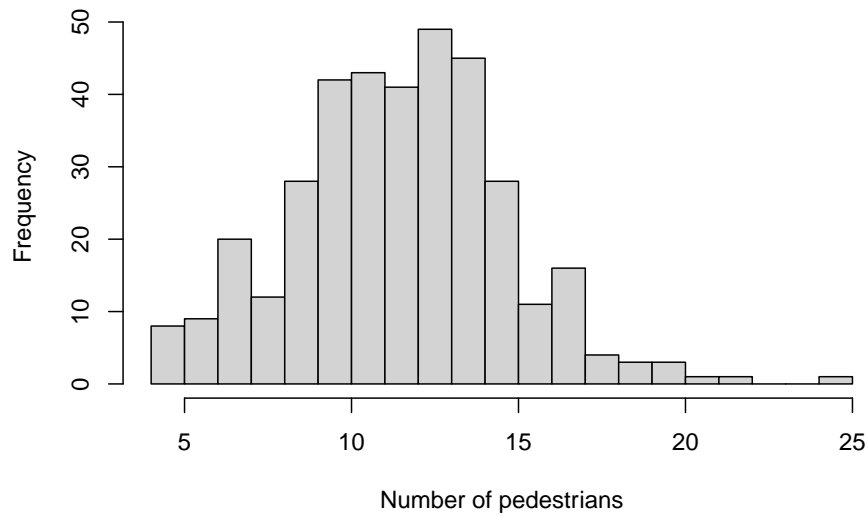
```
# A bridge is crossed by 12 people per minute, on average.
# What is the probability that up to 16 people crosses it in a minute?
ppois(q = 16, lambda = 12)
```

```
## [1] 0.898709
```

```
# Simulate amount of people per minute 10 times
rpois(n = 10, lambda = 12)
```

```
## [1] 6 18 13 16 7 12 8 14 18 14
```

```
# By repeating the observation once every day for a year,
# the expected value (lambda = 12) should become visible
poisson_pedestrians <- rpois(n = 365, lambda = 12)
hist(poisson_pedestrians, breaks = 20,
     xlab = "Number of pedestrians", ylab = "Frequency", main = NULL)
```



```
summary(poisson_pedestrians)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      4.00   10.00   12.00   11.94   14.00   25.00
```

Continuous distributions

For random variables with infinite number of possible values.

Generic definitions

PDF

$$P(a \leq X \leq b) = \int_a^b f(x) \, dx$$

Note that the point probabilities for continuous random variables are intrinsically zero

$$\int_a^a f(x) \, dx = 0$$

CDF

$$P(X \leq t) = \int_{-\infty}^t f(x) \, dx$$