

Wednesday, Jan 18

Summary Measures of a Distribution — Continued

A couple of properties of a distribution that we often want to measure are *location* and *variability*.

Measures of Variability

The **variance** for a sample of observations can be written as

$$s^2 = \frac{(x_1 - \bar{x})^2 + (x_2 - \bar{x})^2 + \cdots + (x_n - \bar{x})^2}{n - 1} = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n - 1}.$$

Example: Consider the following sample of observations: 1, 1, 7. The mean is $\bar{x} = 3$ and the variance is

$$s^2 = \frac{(1 - 3)^2 + (1 - 3)^2 + (7 - 3)^2}{3 - 1} = 12.$$

A related measure is the **standard deviation** which is the square root of the variance, so it can be written as

$$s = \sqrt{\frac{(x_1 - \bar{x})^2 + (x_2 - \bar{x})^2 + \cdots + (x_n - \bar{x})^2}{n - 1}} = \sqrt{\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n - 1}}.$$

Note that the symbol for the variance is s^2 because the variance equals the square of the standard deviation (s).

Another measure is the **range** which is simply defined as the difference between the largest and smallest values,

$$\text{range} = \max(x_1, x_2, \dots, x_n) - \min(x_1, x_2, \dots, x_n),$$

and the **interquartile range** which we will discuss later.

Cumulative Distributions

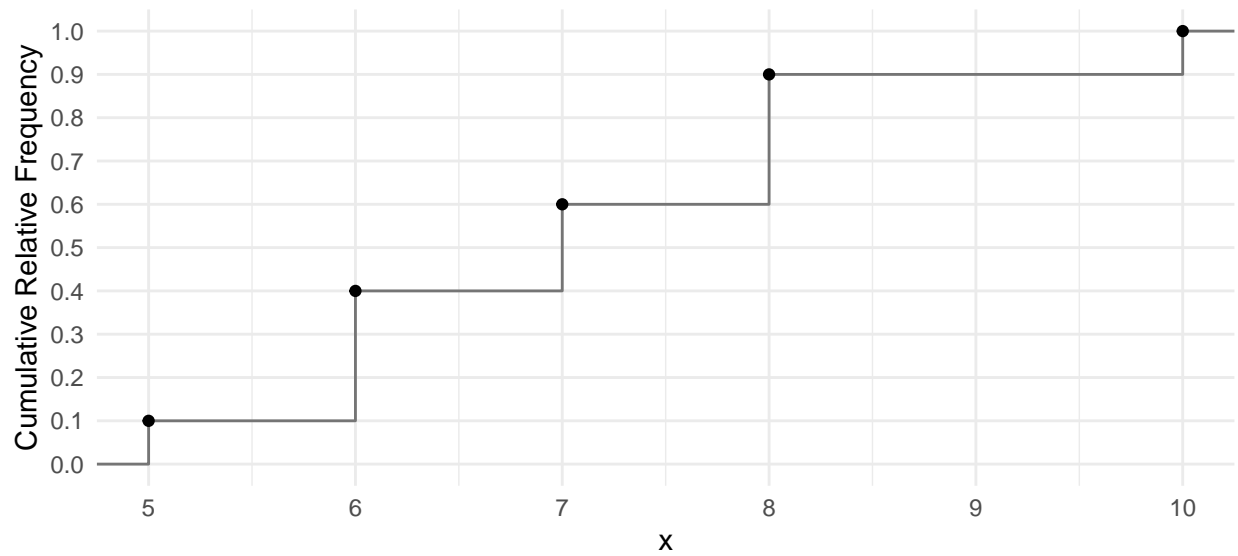
The **cumulative distribution** shows the relationship between the value of the variable and *cumulative relative frequency*.

Example: The following is a hypothetical set of observations of examination scores.

5, 6, 6, 6, 7, 7, 8, 8, 8, 10

| x | Frequency | Relative Frequency | Cumulative Relative Frequency |
|-----|-----------|--------------------|-------------------------------|
| 5 | 1 | 0.1 | 0.1 |
| 6 | 3 | 0.3 | 0.4 |
| 7 | 2 | 0.2 | 0.6 |
| 8 | 3 | 0.3 | 0.9 |
| 10 | 1 | 0.1 | 1.0 |

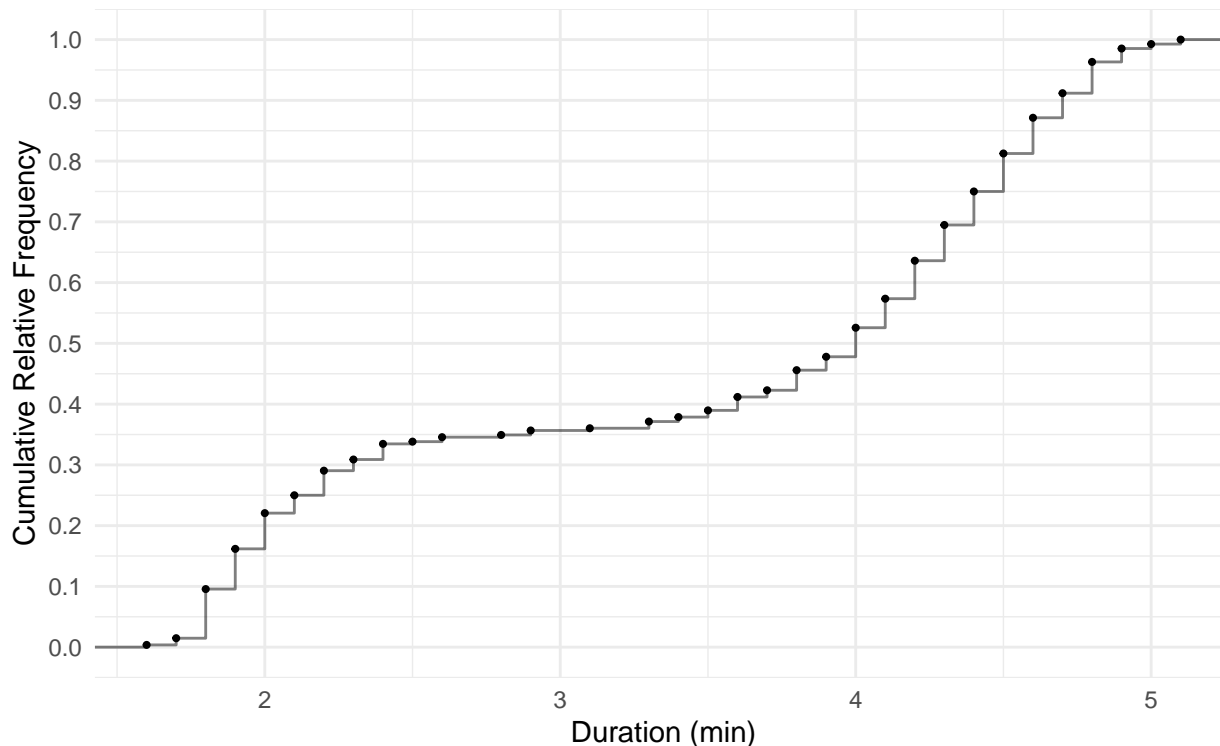
We can graph the cumulative distribution using a *step function*.



Example: Consider the cumulative distribution of the sample of observations of eruption duration of Old Faithful.

| Time | Frequency | Relative Frequency | Cumulative Relative Frequency |
|------|-----------|--------------------|-------------------------------|
| 1.6 | 1 | 0.004 | 0.004 |
| 1.7 | 3 | 0.011 | 0.015 |
| 1.8 | 22 | 0.081 | 0.096 |
| 1.9 | 18 | 0.066 | 0.162 |
| 2 | 16 | 0.059 | 0.221 |
| ⋮ | ⋮ | ⋮ | ⋮ |
| 5.1 | 2 | 0.007 | 1 |

Note: The relative and cumulative relative frequencies above have been rounded.



Finding Percentiles Using a Cumulative Distribution

The P th **percentile** is the value of the variable such that $P\%$ of the observations are *less than that value*.

Finding Percentiles: Finding percentiles from a set of observations is surprisingly complex! Consider the following distribution.

Here are a couple of examples.

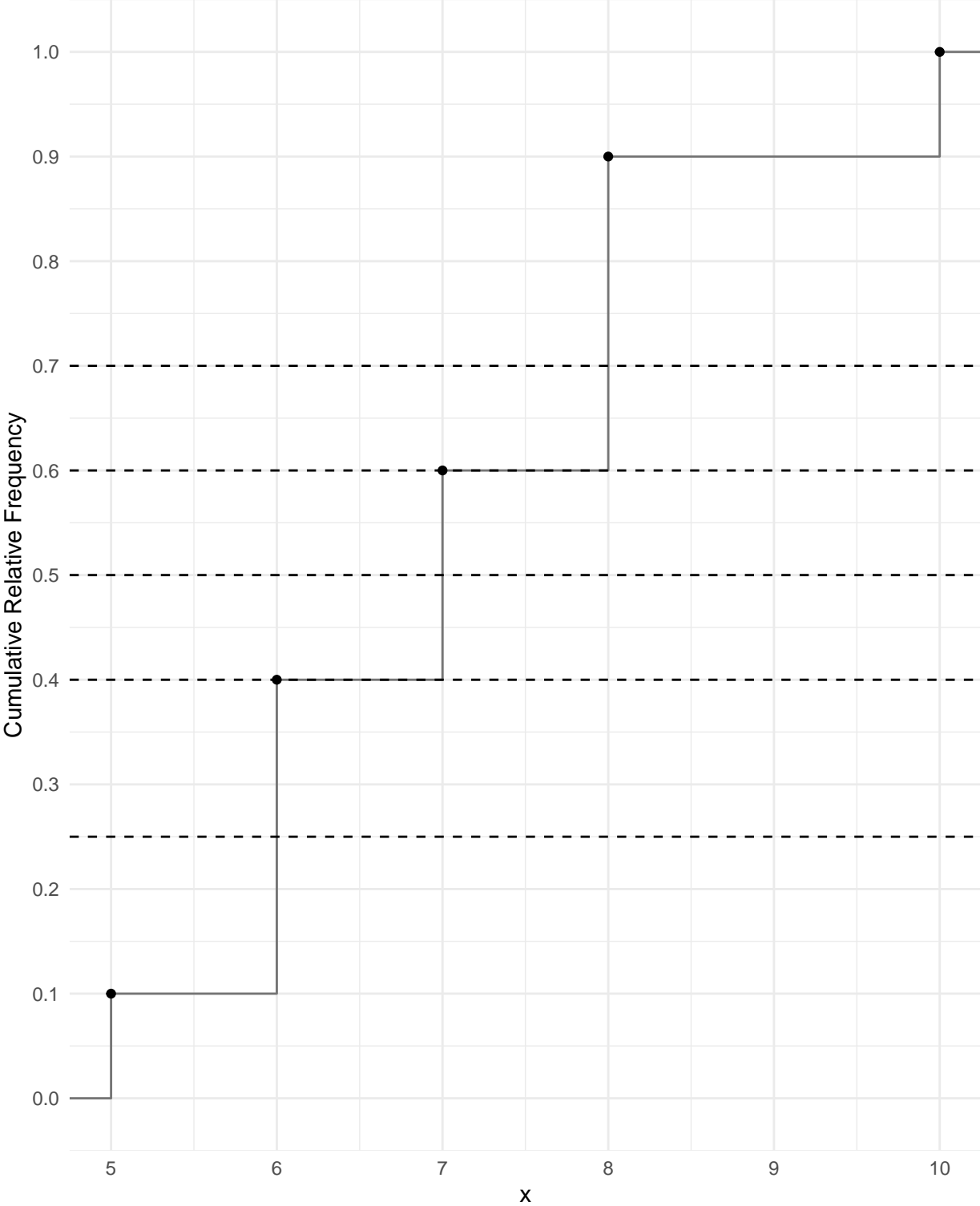
1. What is the 60th percentile? *Several* values of x would qualify! Any x such that $7 < x \leq 8$ has 60% of the observations that are less than it.
2. What is the 70th percentile? *No* values of x would qualify! The percent of observations less than 8 is 60%, and the percent of observations for any $x > 8$ is 90% or more.

| x | Frequency | Relative Frequency | Cumulative Relative Frequency |
|-----|-----------|-----------------------|-------------------------------------|
| 5 | 1 | 0.1 | 0.1 |
| 6 | 3 | 0.3 | 0.4 |
| 7 | 2 | 0.2 | 0.6 |
| 8 | 3 | 0.3 | 0.9 |
| 10 | 1 | 0.1 | 1.0 |

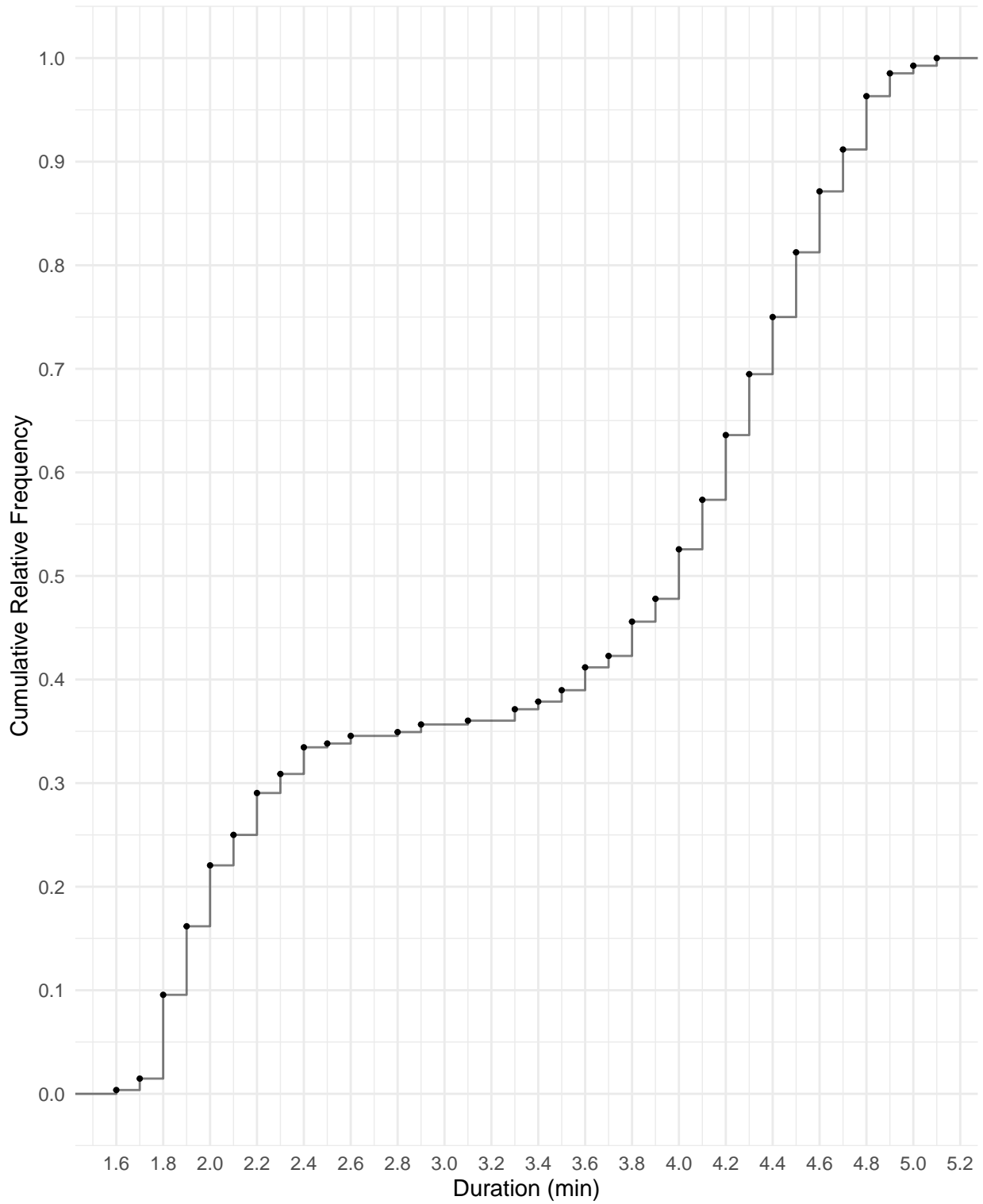
One solution is to use the *midpoint in the first case*, and *the smallest value of x that has more than $P\%$ of observations less than it* in the second case. This is easier to explain/do using a graph of the cumulative distribution.

Finding Percentiles Using the Cumulative Distribution: To find the approximate percentile using a *graph* of the cumulative relative distribution, find the value where the step function crosses the cumulative relative frequency of $P/100$. If more than one value crosses value, use the midpoint (i.e., average of the two values).

Example: We can confirm that the 25th, 40th, 50th, 60th, and 70th percentiles for the distribution show below are 6, 6.5, 7, 7.5, and 8, respectively.



Example: We can use the following plot to approximate the 25th, 50th, and 75th percentiles “by eye” (the actual percentiles are 2.15, 4, and 4.45).



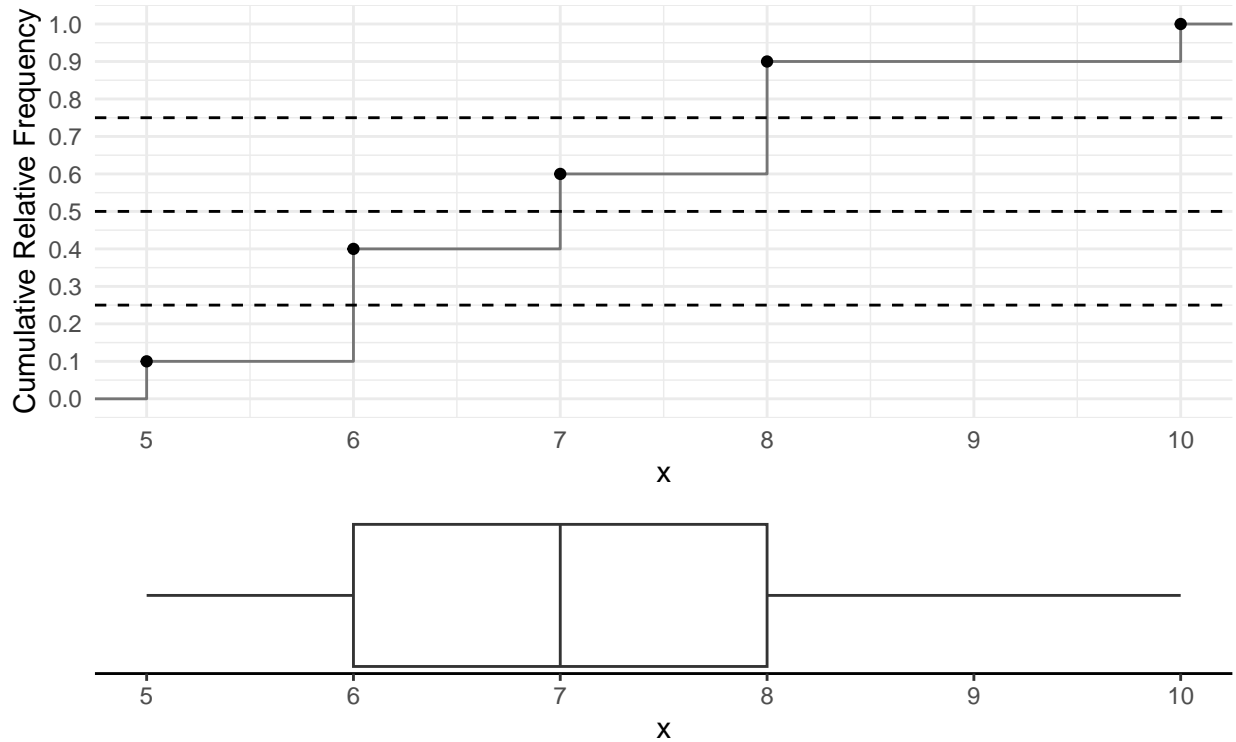
Box Plots

A **box plot** is a graphical representation of a distribution of a quantitative variable that uses what is called a **five-number summary**:

1. minimum
2. first quartile (Q_1) — i.e., 25th percentile
3. second quartile (Q_2) — i.e., 50th percentile and median
4. third quartile (Q_3) — i.e., 75th percentile
5. maximum

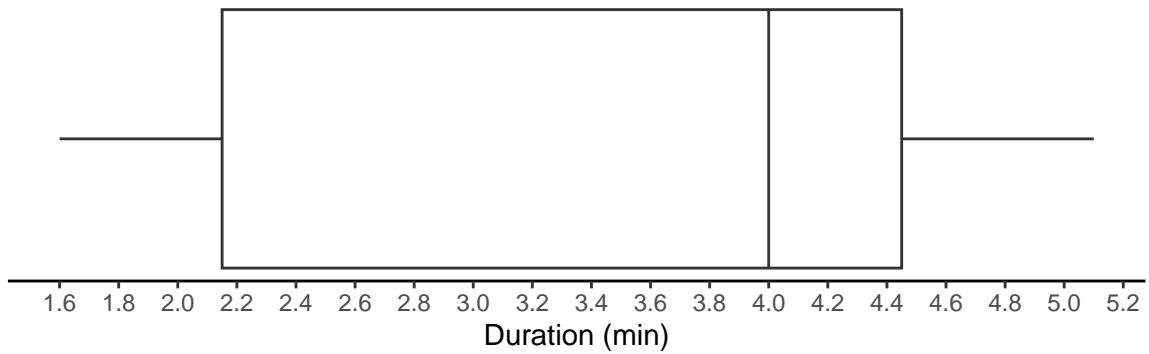
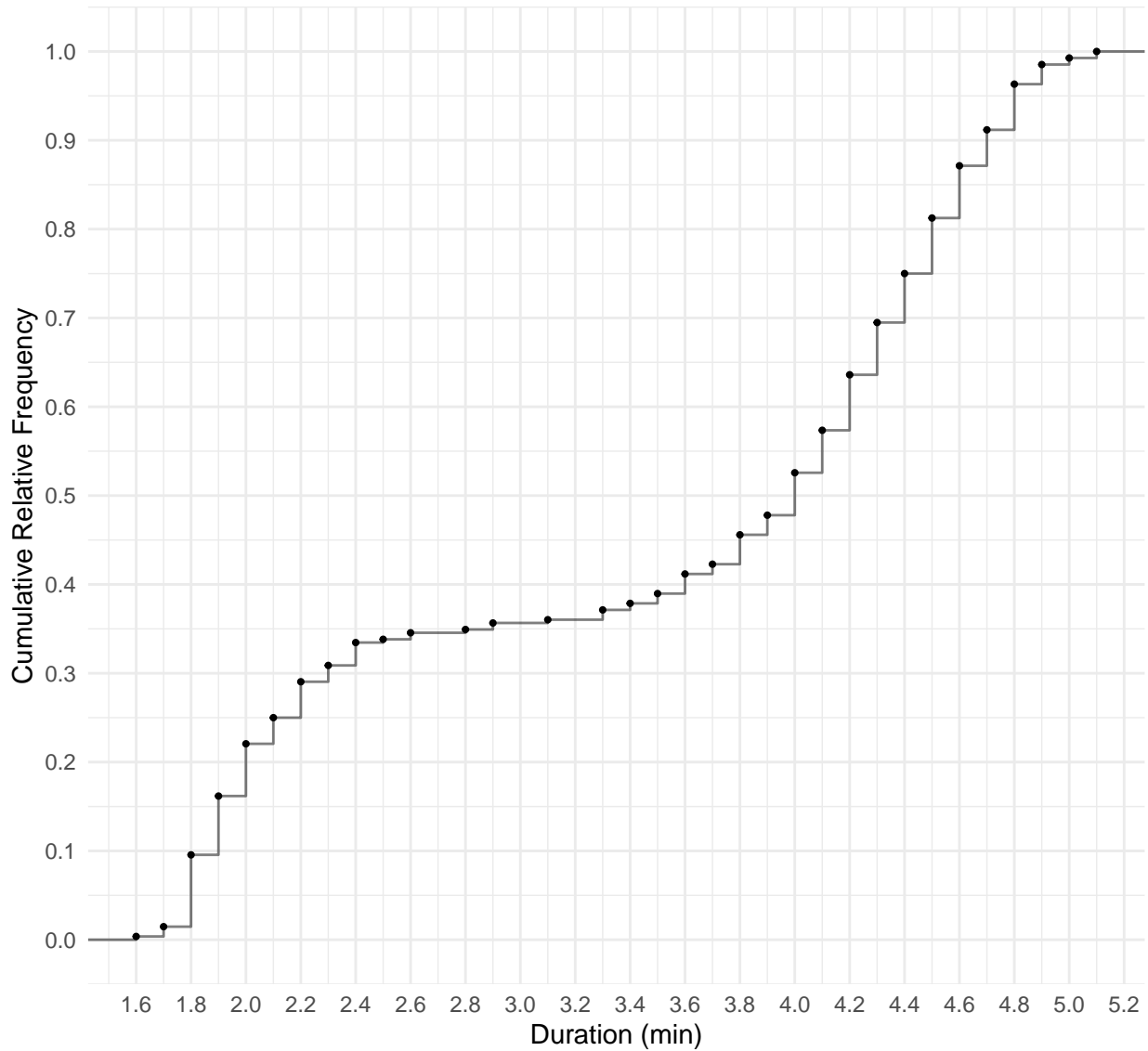
Comment: Because there is more than one way to approximate a percentile and thus a quartile, there is more than one way to draw a boxplot. For consistency we will use the approximation given earlier for finding percentiles.

Example: Box plot based on an earlier example.



Note that the five number summary is 5, 6, 7, 8, and 10.

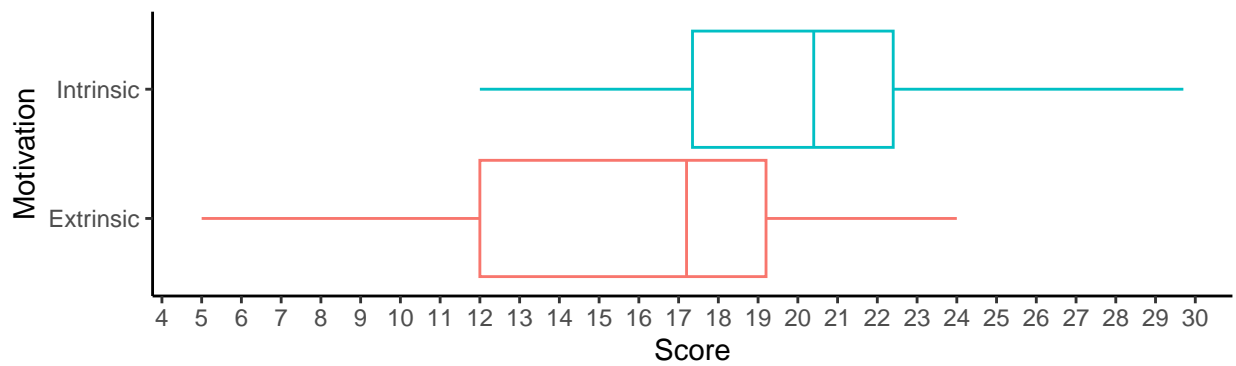
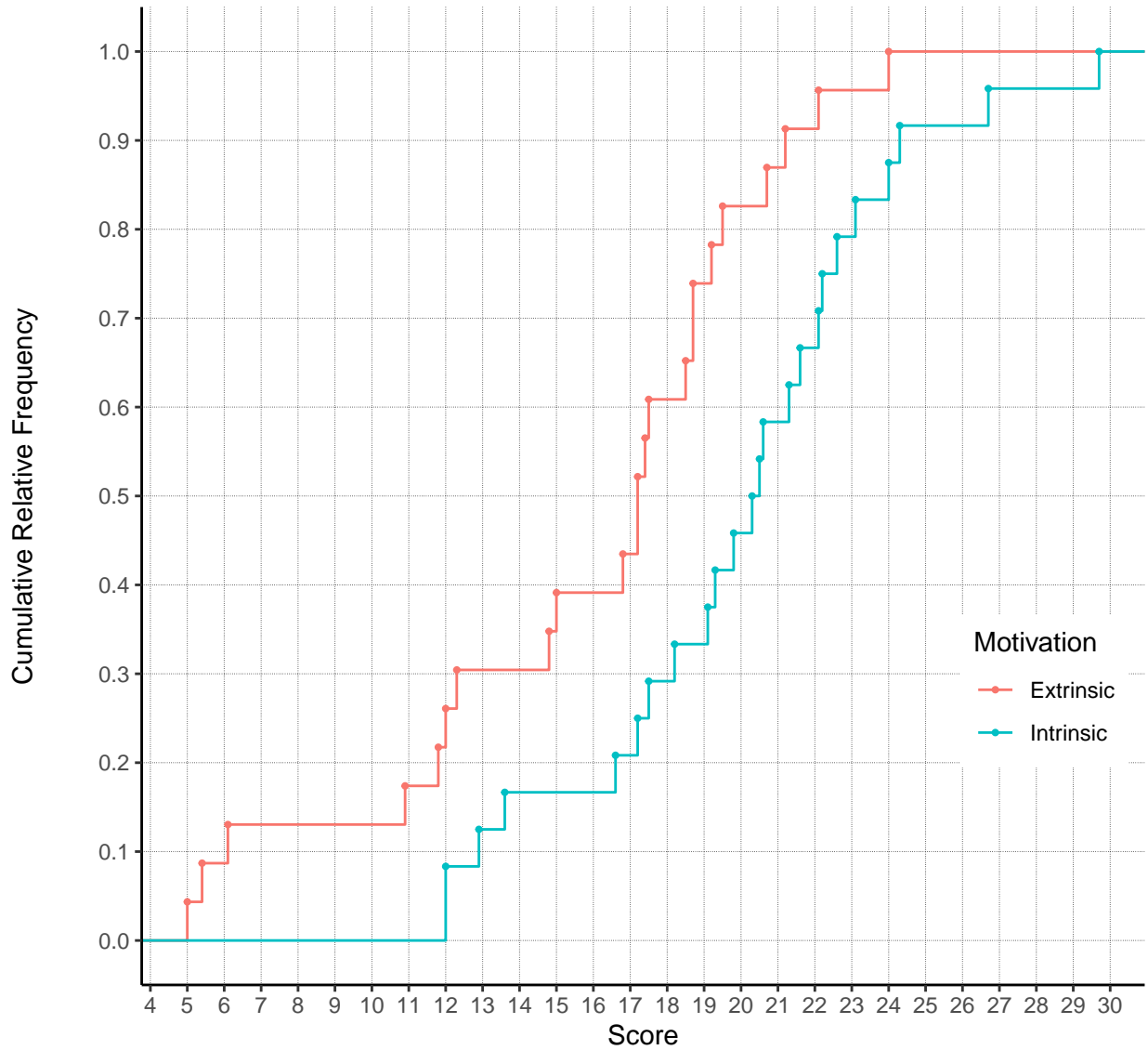
Example: Box plot of the Old Faithful data.



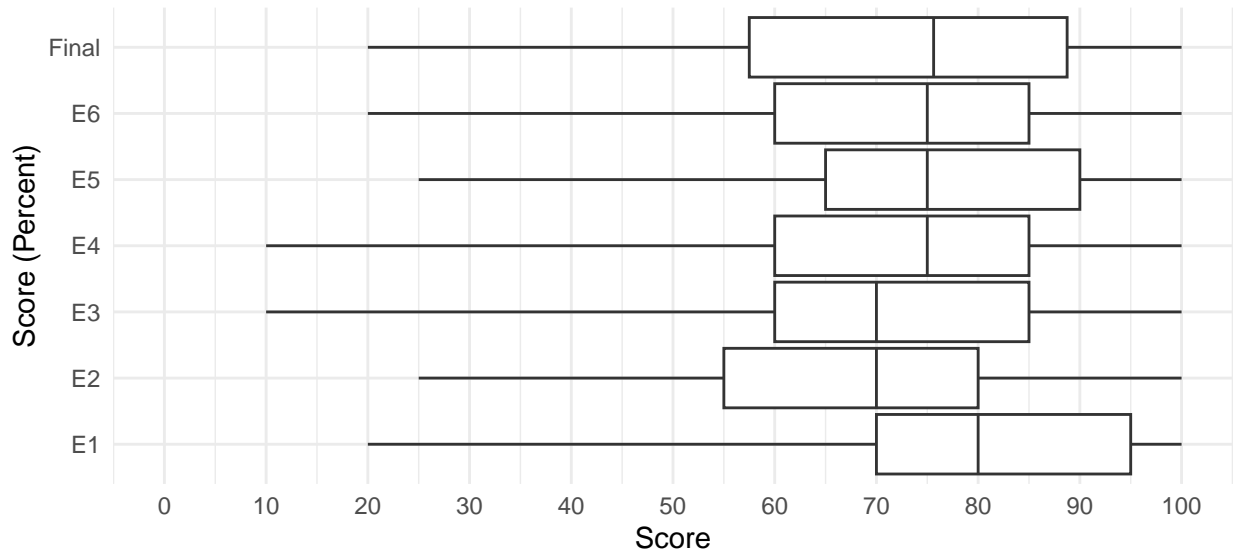
Note that the five number summary is 1.6, 2.15, 4, 4.45, and 5.1.

A box plot visually depicts three summary measures: the **median** (i.e., Q_2), **range** (i.e., maximum – minimum), and **interquartile range** (i.e., $Q_3 - Q_1$).

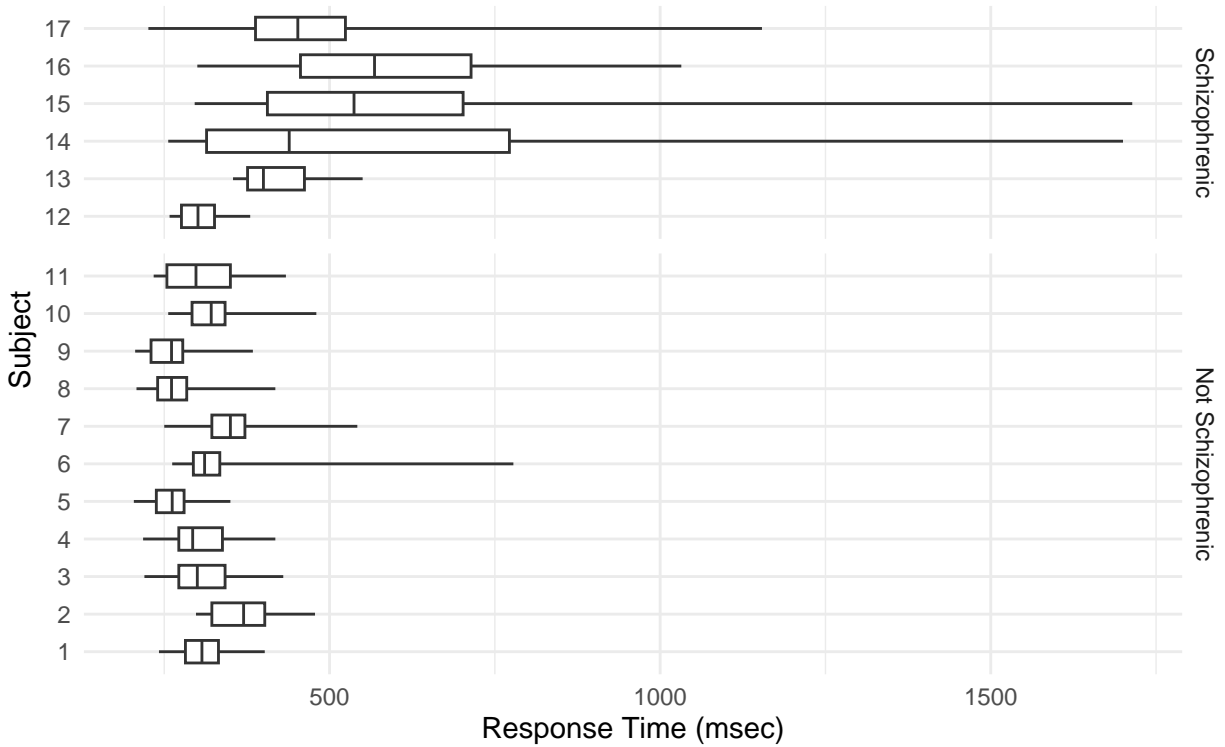
Example: Cumulative distributions and box plots of the data from the study on creativity and motivation.



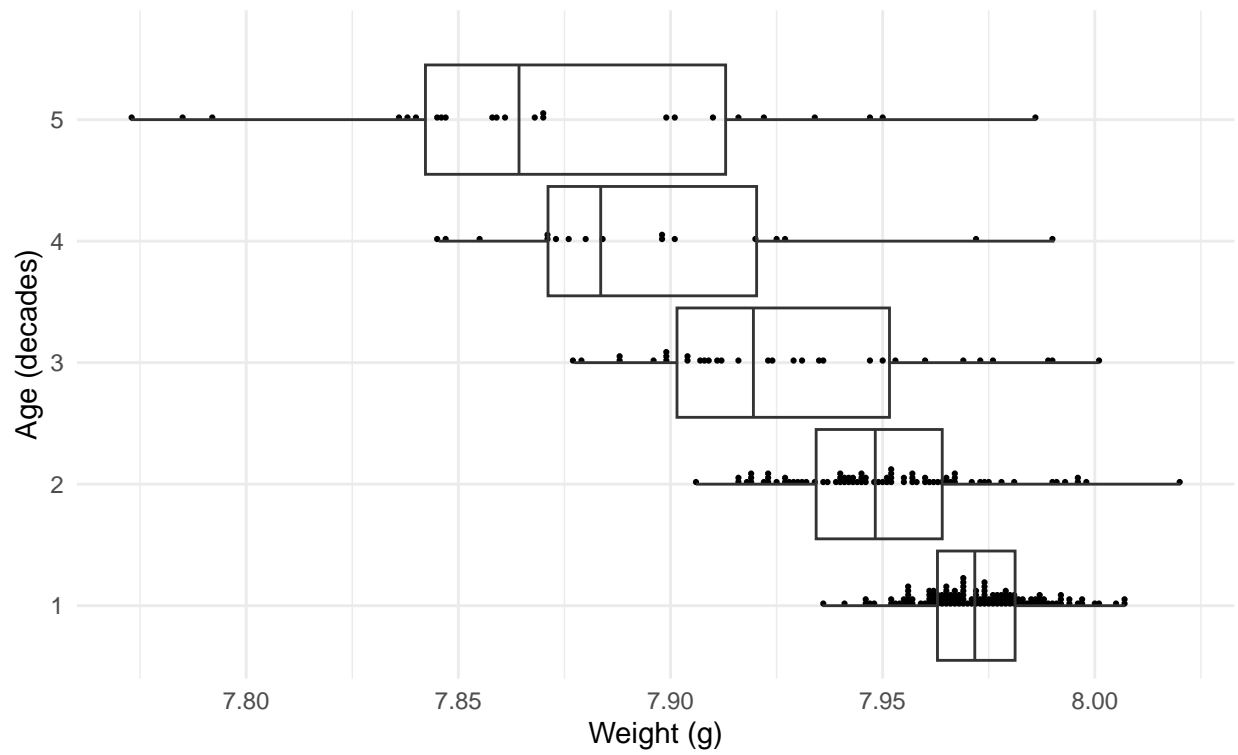
Example: Distribution of examination scores from Stat 251, Fall 2016.



Example: Distribution of response times for 11 non-schizophrenic individuals and six schizophrenic individuals.



Example: Distributions of samples of observations of the weights of gold sovereigns collected from circulation in Manchester, England.

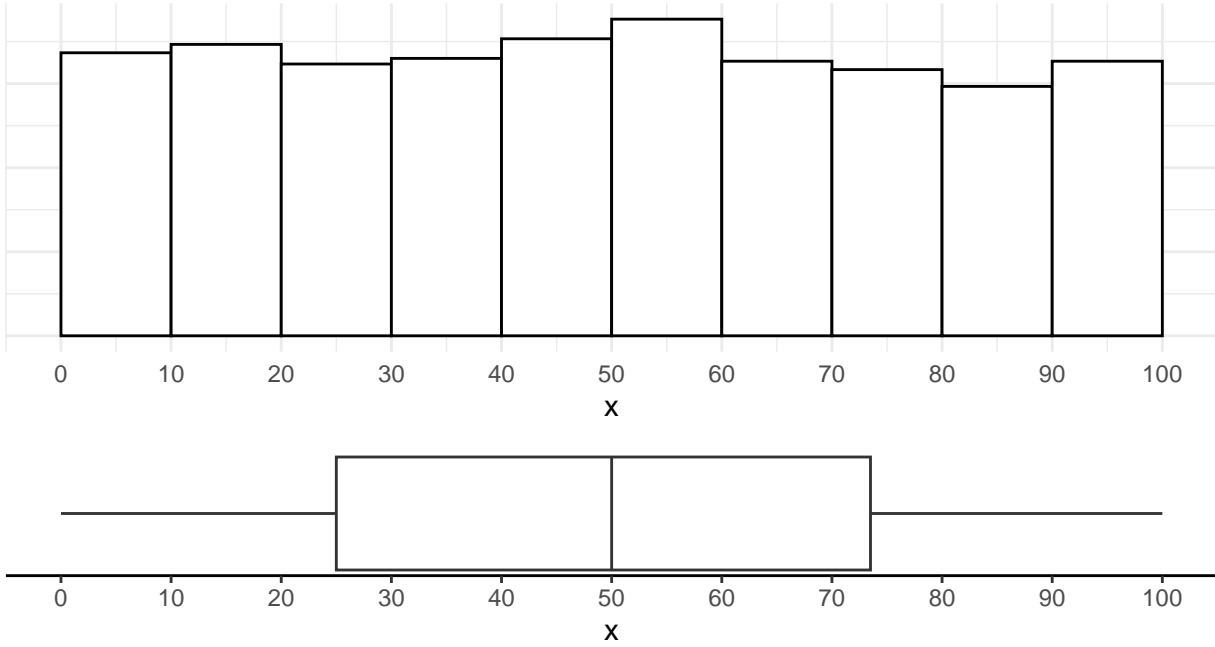


| Age | n | mean | sd | min | Q1 | Q2 | Q3 | max | IQR |
|-----|-----|-------|--------|-------|-------|-------|-------|-------|--------|
| 5 | 24 | 7.873 | 0.0535 | 7.773 | 7.842 | 7.864 | 7.913 | 7.986 | 0.0708 |
| 4 | 17 | 7.896 | 0.0406 | 7.845 | 7.871 | 7.883 | 7.920 | 7.990 | 0.0492 |
| 3 | 32 | 7.928 | 0.0343 | 7.877 | 7.902 | 7.920 | 7.952 | 8.001 | 0.0501 |
| 2 | 78 | 7.950 | 0.0227 | 7.906 | 7.934 | 7.948 | 7.964 | 8.020 | 0.0297 |
| 1 | 123 | 7.973 | 0.0141 | 7.936 | 7.963 | 7.972 | 7.981 | 8.007 | 0.0183 |

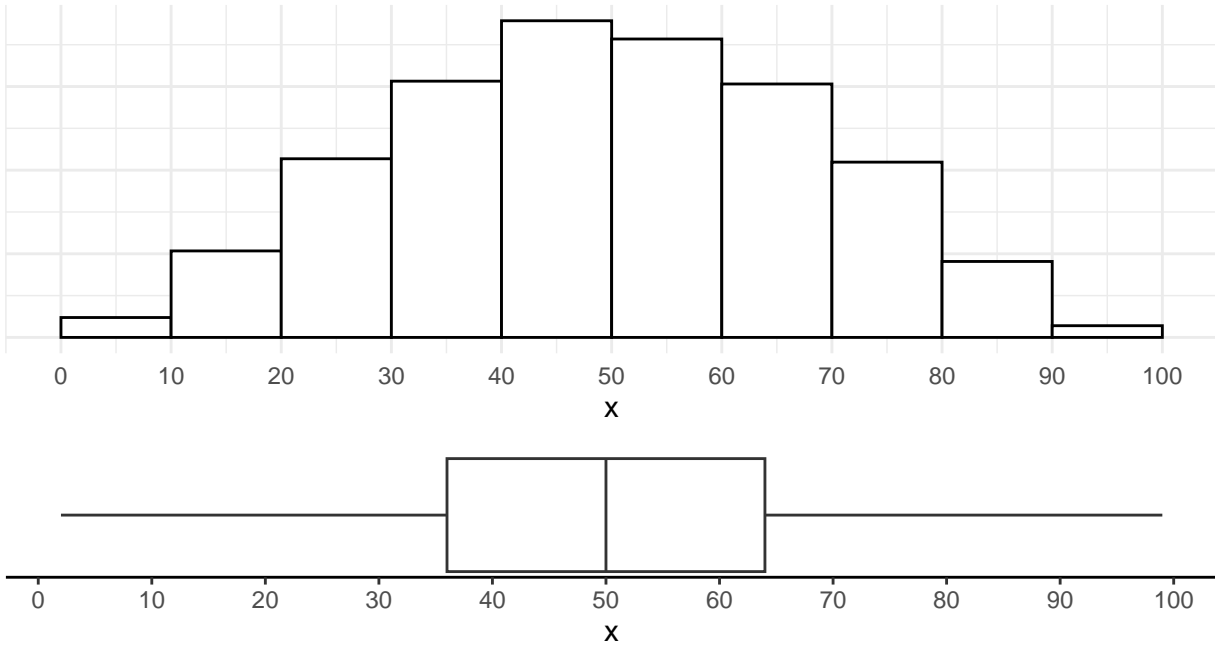
Distribution Shapes

Some terms we use to describe the *shape* of the distribution of a quantitative variable: *symmetric*, *uniform*, *left-skewed*, *right-skewed*, *uni-modal*, *bi-modal*.

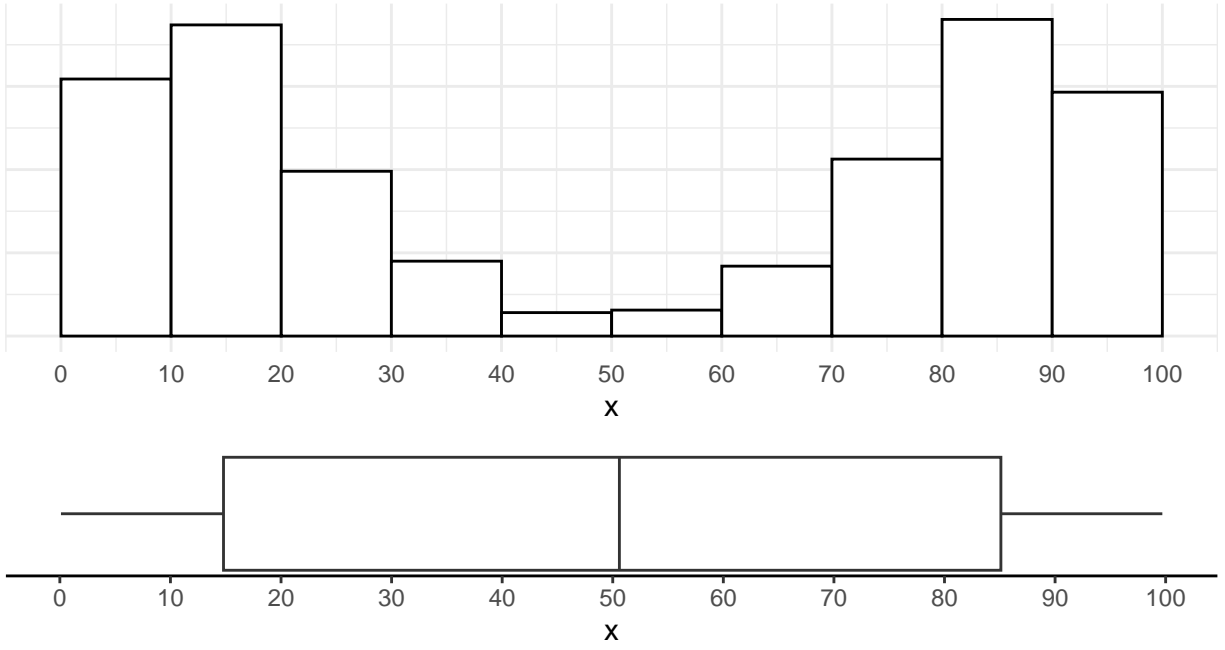
Example: Uniform and symmetric.



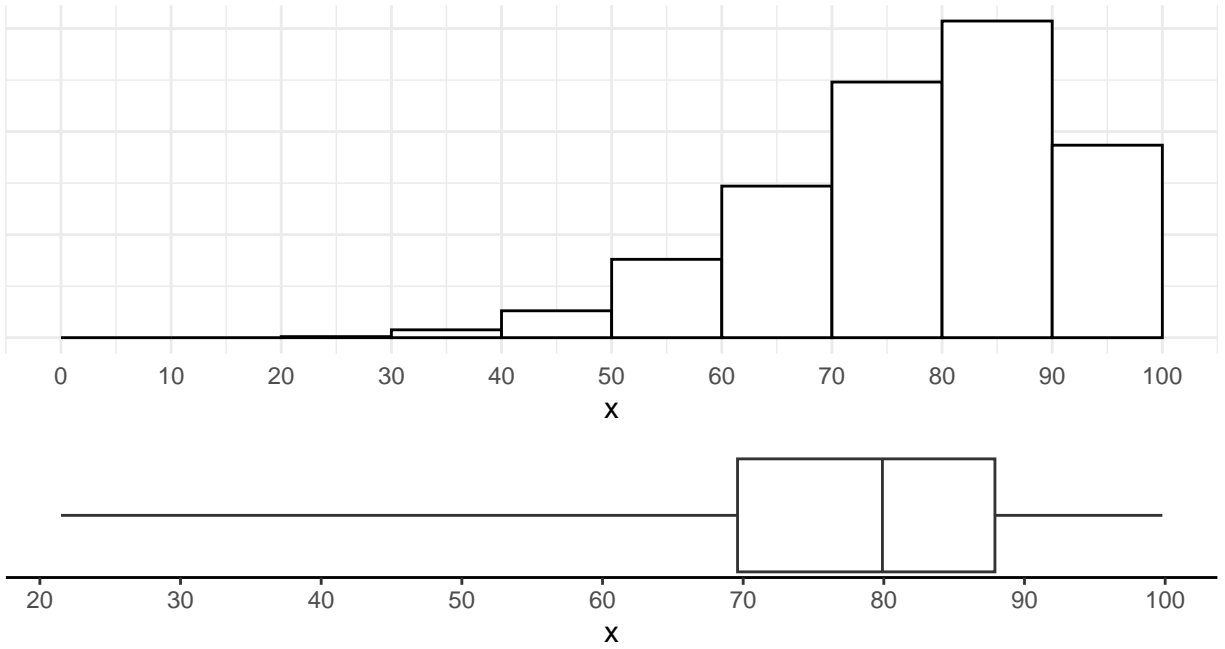
Example: Uni-modal and symmetric.



Example: Bi-modal and symmetric.



Example: Left-skewed and uni-modal.



Example: Right-skewed and uni-modal.

