

Statistical Methods and Data Analysis I

Lecture 6: Probability Distributions

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Probability distribution

Lists all possible events and the probabilities with which they occur.

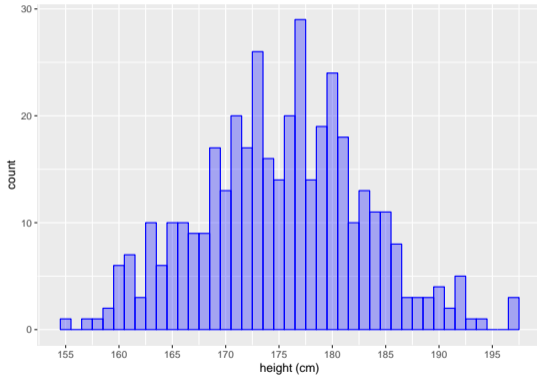
Rules:

- 1 The events must be disjoint.
- 2 All the probabilities must be between 0 and 1 inclusive.
- 3 The sum of all probabilities (the probability of “*something happening*”) must be 1.

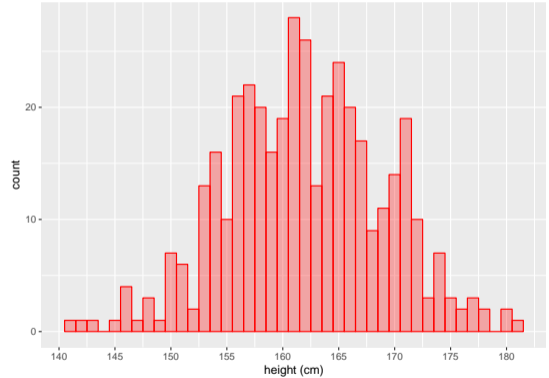
Easy enough if the events are discrete and their number is small enough.

Visualizing Discrete Probability Distributions: Histograms

Distribution of men's height

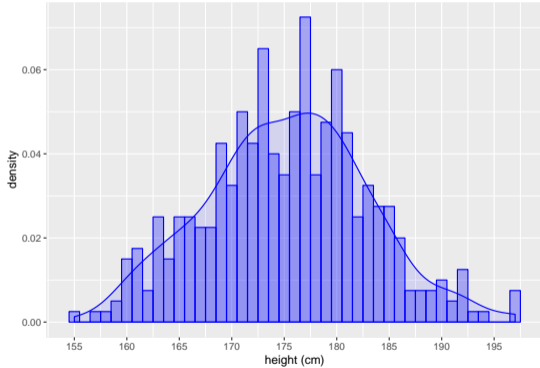


Distribution of women's height

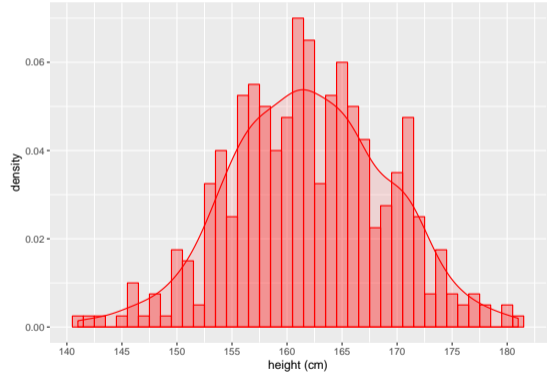


Normalizing Counts: Probability Density

Distribution of men's height

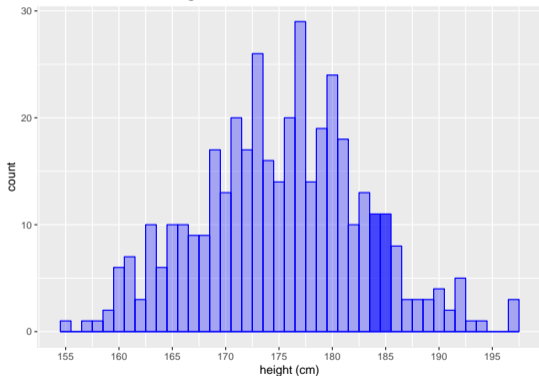


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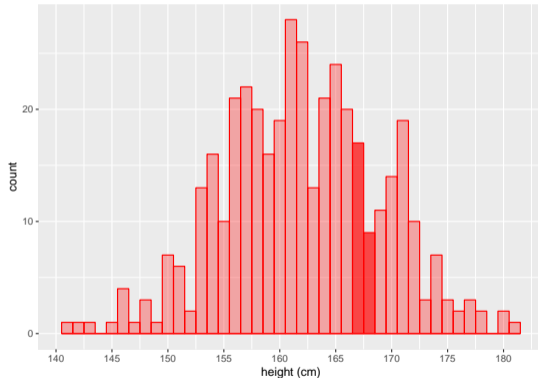


Computing Probability: Discrete Case

Distribution of men's height

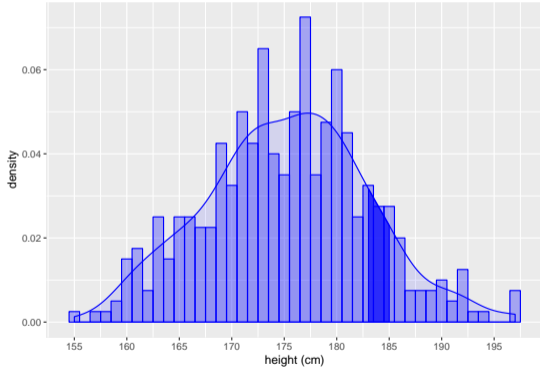


Distribution of women's height

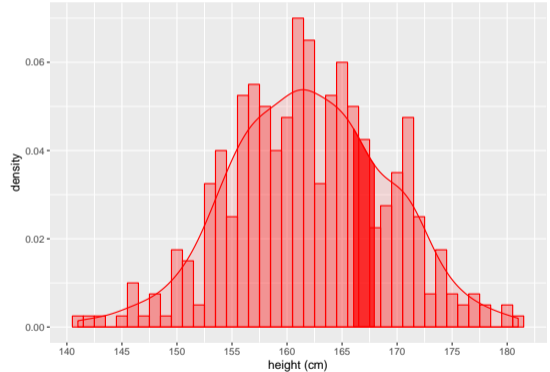


Computing Probability: Continuous Case

Distribution of men's height



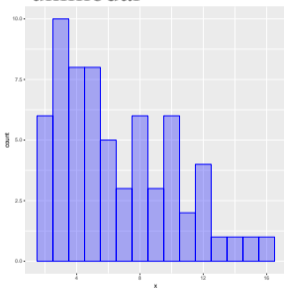
Distribution of women's height



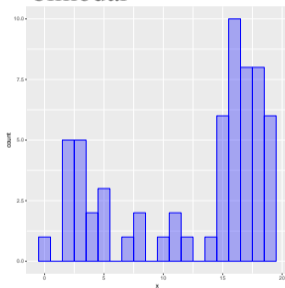
Common Distribution Shapes: Modality

Does the histogram have a single prominent peak? Two peaks? Several peaks? No obvious peaks?

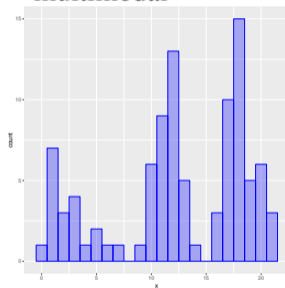
unimodal



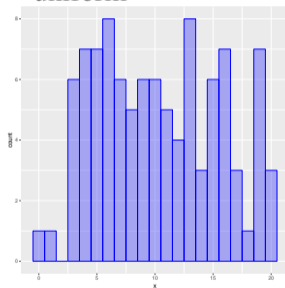
bimodal



multimodal



uniform

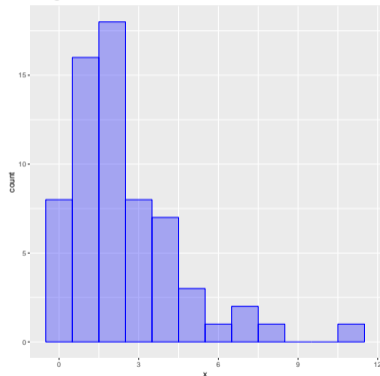


Imagining the corresponding density function may help.

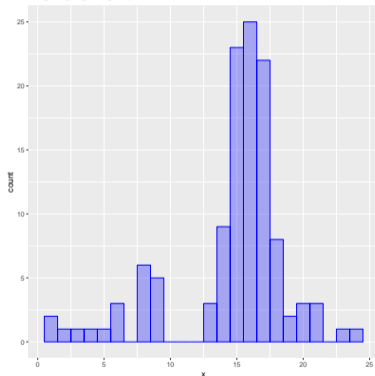
Common Distribution Shapes: Skewness

Is the histogram *right skewed*, *left skewed*, or *symmetric*?

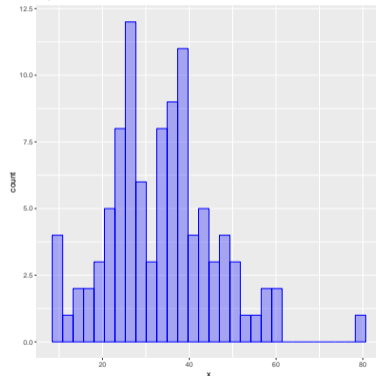
right skew



left skew



symmetric

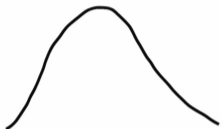


Histograms are said to be skewed to the side of the long tail.

Common Distribution Shapes: Continuous

- modality

unimodal



bimodal



multimodal



uniform



Common Distribution Shapes: Continuous

- modality

unimodal



bimodal



multimodal



uniform



- skewness

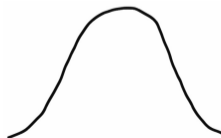
right skew



left skew



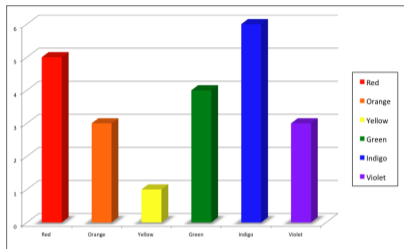
symmetric



Central Tendency: Mode

Mode

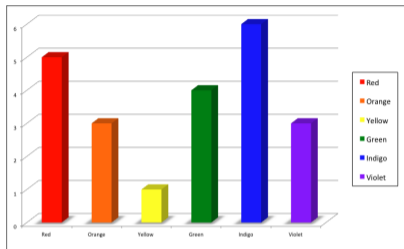
The most common, most probable, most popular, or most typical value of a random variable (outcome of a random process).



Central Tendency: Mode

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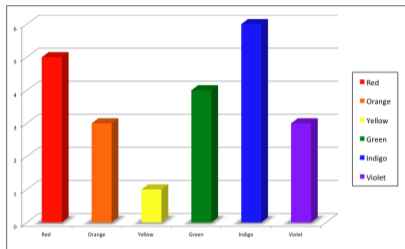
Less boring examples than “what is your favourite colour?”:

- *Which party will form a government after elections?*

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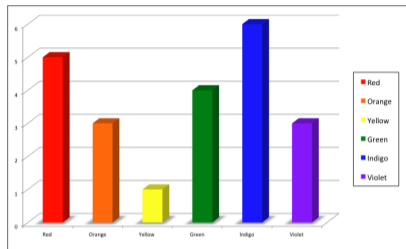
Less boring examples than “what is your favourite colour?”:

- *Which party will form a government after elections?*

The one that got the most votes (but this won't tell you how stable the coalition will be).

Mode

The most common, most probable, most popular, or most typical value of a random variable (outcome of a random process).



Less boring examples than “what is your favourite colour?”:

- *Which party will form a government after elections?*

The one that got the most votes (but this won't tell you how stable the coalition will be).

NB: It does not make sense to compute the average colour or the average number of votes among the competing parties (except maybe to assess how fragmented the parliament is).

Discrete case:

($p(x_i)$ is the probability of outcome x_i)

$$\bar{x} = \mu = \sum_i x_i p(x_i)$$

Continuous case:

($p(x)$ is the probability density function, PDF)

$$\bar{x} = \mu = \int_{-\infty}^{+\infty} xp(x)dx$$

If the probabilities of all outcomes x_i are the same (uniform distribution) then, for n possible outcomes, we have our usual definition of arithmetic mean:

$$p(x_i) = \frac{1}{n}, \quad \bar{x} = \mu = \frac{1}{n} \sum_i x_i$$

If we know the probabilities of all the outcomes (we know the probability density function in the continuous case) then our mean is the (*mathematical*) *expectation*, *expected value*: $E[x] = \mu$.