(§1.2) Descriptive statistics; (§1.3) Measures of location

Common naming conventions:

 \bullet Population size: N

• Sample size: n

• Sample from two different populations: n, m, or n_1, n_2

• Data: $x_1, x_2, x_3, \dots x_n$

Stem-and-leaf displays

```
> x sample(1:50, size=20, replace=TRUE)

> sort(x)

[1] 2 2 2 3 9 14 18 19 20 21 21 22 22 29 30 32 32

[18] 33 44 47

> stem(x)
```

The sample function generates numbers in the range provided as the first argument, with a size equal to the second argument. sort(x) sorts the values stored in x, and stem(x) does the following:

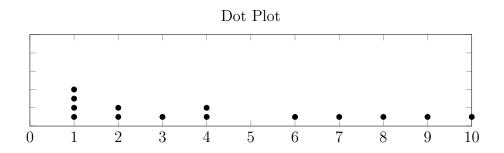
Each "stem" refers to the highest digits and each "leaf" is the latter digits. This is the stem-and-leaf display for the dataset stored in x:

| Stem | Leaves |
|------|---|
| 0 | 2 2 2 3 9 |
| 1 | $ \begin{array}{cccccccccccccccccccccccccccccccccccc$ |
| 2 | 0 1 1 2 2 9 |
| 3 | $egin{array}{cccccccccccccccccccccccccccccccccccc$ |
| 4 | 4 7 |

Dot plots

```
> x <- sample(1:10, size=15, replace=TRUE)
> x
[1] 7 8 1 3 4 10 1 2 2 1 1 4 4 9 6
> stripchart(x, method="stack", offset=0.5, at=0.15, pch=20)
```

stripchart() constructs the dot plot. The offset and at values affect the appearance. pch means p rinting ch aracter. Having it set to 20 makes the dot solid.



Histograms

- For quantitative data.
- May differ depending on whether a variable is discrete or continuous.

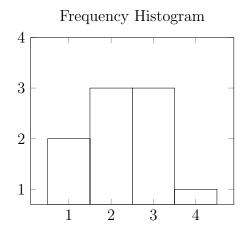
Discrete data is either infinite or countably infinite: orderable, typically integers. Includes ration numbers.

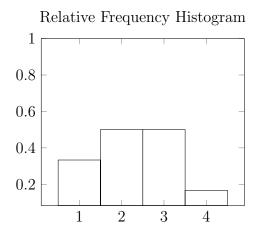
- $x_1 = \#$ of tuberculosis cases in the 50 largest U.S. cities in 2013. \leftarrow Might be 0, 1, 2, ... but $\neq \infty$.
- $x_i \# \text{ of "successes"}$ when 100 patients treated. $\leftarrow \text{Count data in which we know a hard upper limit (in this case 100).$
- $x_i \#$ of rooms in a rental unit. $\leftarrow 1, 2, \ldots, 100$. At least 1 guaranteed as a hard minimum limit.

Histograms, when data is discrete, use vertical bars with height proportional to the number of times each value occurs.

Frequency Histogram vs Relative Frequency Histogram

For the data: $\{1, 1, 2, 2, 2, 3\}$





Relative Frequency Histograms have a total height of 1.

R Code for Histograms

```
> # Use R to simulate 100 values from {1, 2, 3, ..., 10}
> my_data <- sample(1:10, size=100, replace=TRUE)
> # Here are the first 20 values:
> my_data[1:20]
[1] 4 3 5 6 5 4 10 2 6 2 5 8 7 10 4 10 7 6
[19] 5 7
> # Create a frequency histogram:
> hist(my_data. main="Plot_1_-_discrete_data")
> # Create a relative frequency histogram:
> hist(my_data, probability=TRUE, main="Plot_2")
```

Note: R uses # for comments. You're a CS major, you know what these functions and their arguments do.

R Code of a Histogram Using Continuous Data

```
> # Generate 100 rational values from 1:10.
> my_data <- runif(100, min=1, max=10)
> # Create a histogram using the same code for discrete data:
> hist(my_data, main="Plot_3_-_Continuous_Data")
> # And a relative frequency histogram:
> hist(my_data, probability=TRUE, main="Plot_4")
```

Here, \mathbf{r} refers to random and \mathbf{unif} refers to uniform in \mathbf{runif} , it does not refer to \mathbf{run} if.

Endpoint: 41:32.