

# STA130H1 - Class # 1

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# A Brief Introduction to R

# What is R?



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## The R Project for Statistical Computing

### Getting Started

R is a free software environment for statistical computing and graphics. It compiles and runs on a wide variety of UNIX platforms, Windows and MacOS. To **download R**, please choose your preferred [CRAN mirror](#).

If you have questions about R like how to download and install the software, or what the license terms are, please read our [answers to frequently asked questions](#) before you send an email.

# R Coding basics

Go to console ...

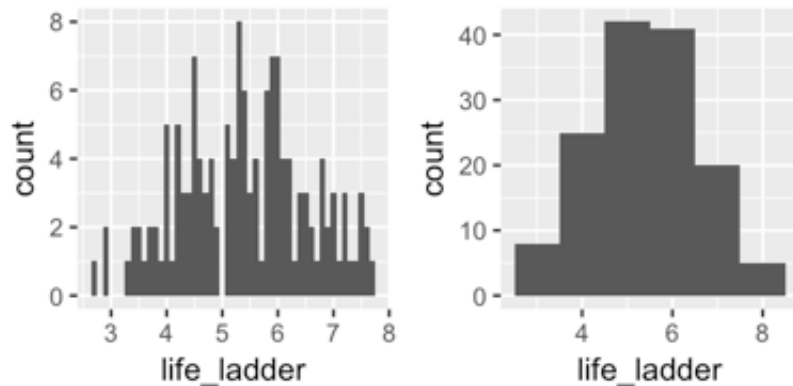
# Histograms and Density Functions

- The histogram of a variable is a graphical method to visualize the distribution of a variable.

# Histograms and Density Functions

- Different bin width will yield different histograms

```
p1 <- ggplot(data = happinessdata2016, aes(x = life_ladder)) +  
  geom_histogram(binwidth = 0.1)  
p2 <- ggplot(data = happinessdata2016, aes(x = life_ladder)) +  
  geom_histogram(binwidth = 1.0)  
grid.arrange(p1,p2,nrow = 1)
```



# Mathematical Definition of Histogram

- The bins of the histogram are the intervals:

$$[x_0 + mh, x_0 + (m + 1)h).$$

$x_0$  is the origin,  $m = \dots, -1, 0, 1, \dots$  indexes the bins, and  $h = (x_0 + (m + 1)h) - (x_0 + mh)$  is the bin width.

- The bins can be used to construct rectangles with width  $h$  and height  $\hat{f}(x)$ .
- The area of these rectangles is  $h\hat{f}(x)$ .
- The area of the rectangles is the same as the proportion of data in the same bin as  $x$ .



# Example - Mathematical Definition of Histogram

```
dat <- data_frame(x = c(1,2,2.5,3,7))
```

```
dat$x
```

```
[1] 1.0 2.0 2.5 3.0 7.0
```

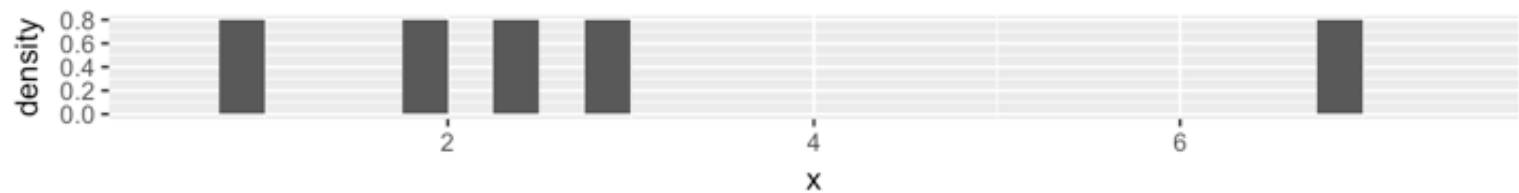
Let  $x_0 = 0.5, h = 0.25, m = 1, \dots, 29$

```
seq(0.5,7.5,by = 0.25)
```

```
[1] 0.50 0.75 1.00 1.25 1.50 1.75 2.00 2.25 2.50 2.75 3.00 3.25 3.50 3.75  
[15] 4.00 4.25 4.50 4.75 5.00 5.25 5.50 5.75 6.00 6.25 6.50 6.75 7.00 7.25  
[29] 7.50
```

The bins are:  $[0.50, 0.75), [0.75, 1.00), [1.00, 1.25), \dots, [7.25, 7.50)$ .

# Example - Mathematical Definition of Histogram



y	count	x	xmin	xmax	density
0.0	0	0.625	0.50	0.75	0.0
0.8	1	0.875	0.75	1.00	0.8
0.0	0	1.125	1.00	1.25	0.0
0.0	0	1.375	1.25	1.50	0.0
0.0	0	1.625	1.50	1.75	10/18 0.0

# Mathematical Definition of Histogram

- Suppose we have data:  $X_1, X_2, \dots, X_n$ .
- Let  $\#\{X_i \text{ in same bin as } x\}$  be the number of data points  $X_i$  in the same bin as  $x$ .
- Let  $n$  be the total number of data points. So,  $\frac{\#\{X_i \text{ in same bin as } x\}}{n}$  is the proportion of data in the same bin as  $x$ .
- Area of rectangle containing  $x \approx \frac{\#\{X_i \text{ in same bin as } x\}}{n}$ .
- $$hf(x) = \frac{\#\{X_i \text{ in same bin as } x\}}{n}.$$

# Mathematical Definition of Histogram

$$\hat{f}(x) = \frac{1}{hn} \#\{X_i \text{ in same bin as } x\}$$

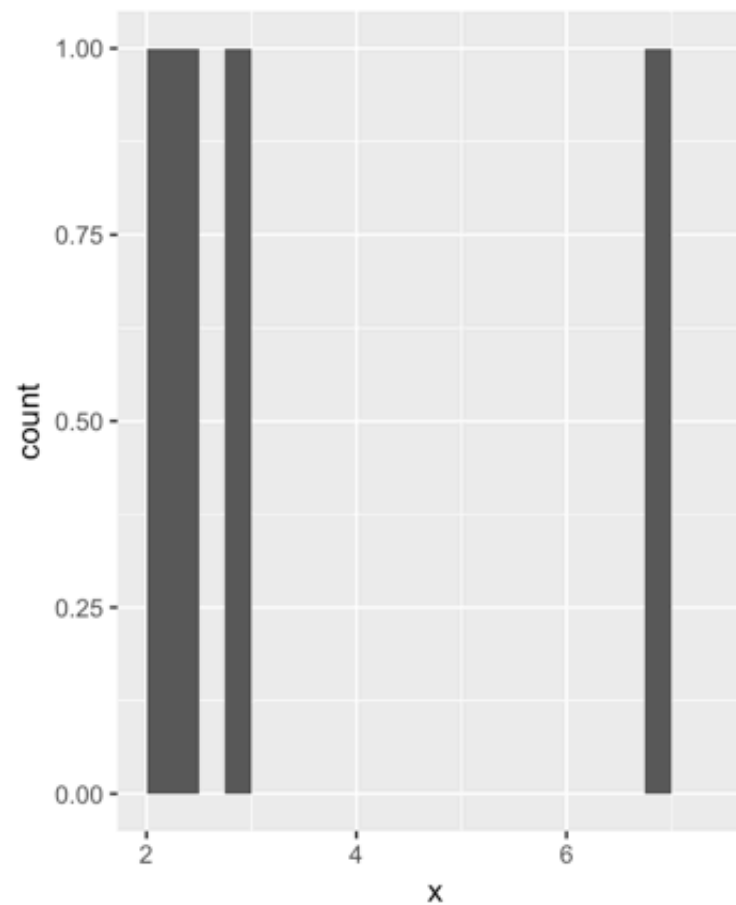
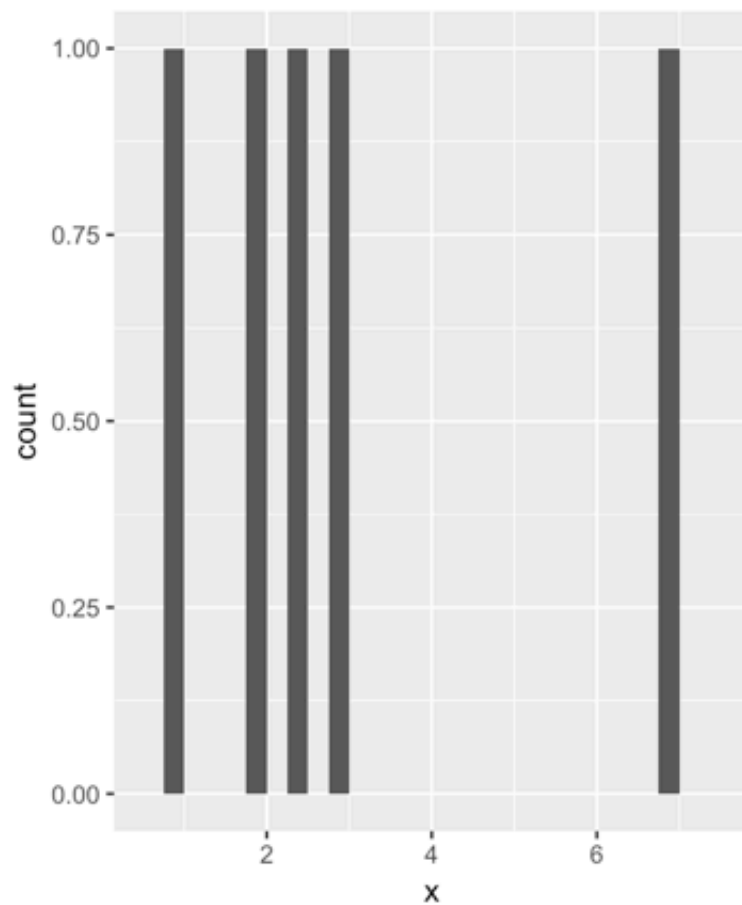
is called the **histogram estimator**.

$\hat{f}(x)$  is an estimate of the density at a point  $x$ .

To construct the histogram we have to choose an origin  $x_0$  and bin width  $h$ .

# Choosing Origin and Bin Width in R

Same bin width but different origin



# Naive Estimator of Density

The histogram can be centered on a point  $x$ .

$$\hat{f}(x) = \frac{1}{2hn} \#\{X_i \in (x - h, x + h)\}$$

This estimate is often called the **naive estimator** of the density.

This can be expressed by defining the weight function  $w(x)$  by

$$w(x) = \begin{cases} 1/2 & \text{if } |x| < 1 \\ 0 & \text{if } |x| \geq 1. \end{cases}$$

Then the naive estimator can be written

$$\hat{f}(x) = \frac{1}{n} \sum_{i=1}^n \frac{1}{h} w\left(\frac{x - X_i}{h}\right).$$

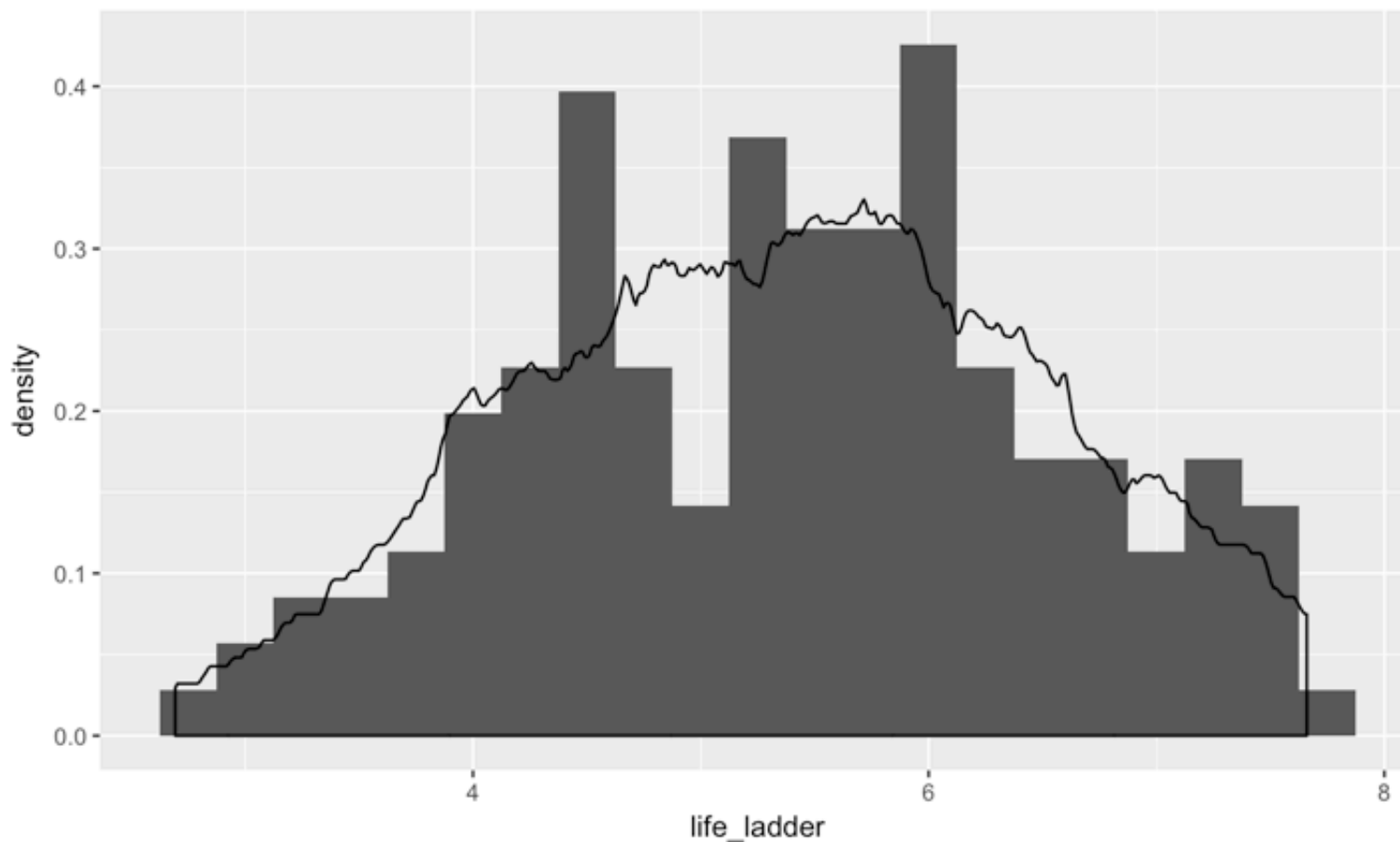
# Naive Estimator of Density

$$\hat{f}(x) = \frac{1}{n} \sum_{i=1}^n \frac{1}{h} w\left(\frac{x - X_i}{h}\right).$$

- Consider the histogram constructed from data using bins of width  $2h$ .
- Assume that  $x$  is at the centre of one of the histogram bins then the naive estimate will be the same as the  $y$  value of the histogram estimate.

# Naive Estimator in R

```
ggplot(data = happinessdata2016, aes(x = life_ladder, ..density..)) +  
  geom_histogram(binwidth = 0.25) + geom_density(kernel = "rectangular" )
```





# Kernel Estimator

Replace the weight function is by a kernel function  $K(x) \geq 0$  which satisfies  $\int_{-\infty}^{\infty} K(x) = 1$ . The **kernel estimator** of the density function is defined by

$$\hat{f}(x) = \frac{1}{n} \sum_{i=1}^n \frac{1}{h} K\left(\frac{x - X_i}{h}\right).$$

The Gaussian kernel is a popular choice

$$K(x) = \frac{1}{\sqrt{2\pi}} \exp(-x^2/2), -\infty < x < \infty.$$

# Kernel Estimator in R

```
p_gauss <- ggplot(data = happinessdata2016, aes(x = life_ladder, ..density..)) +  
  geom_histogram(binwidth = 0.25) + geom_density(kernel = "gaussian" )  
p_rect <- ggplot(data = happinessdata2016, aes(x = life_ladder, ..density..)) +  
  geom_histogram(binwidth = 0.25) + geom_density(kernel = "rectangular" )  
grid.arrange(p_gauss, p_rect)
```

