

# ECE 295: Lecture 03 Histograms

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The science of making sense of data!

# Why study statistics?

... Using fancy tools like neural nets, boosting, and support vector machines without understanding basic statistics is like doing brain surgery before knowing how to use a band-aid...

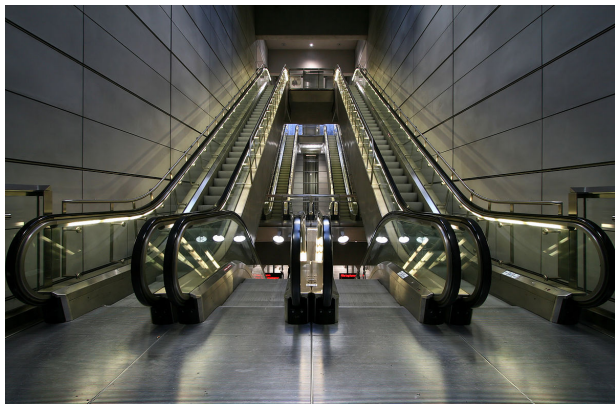
Larry Wasserman, "All of Statistics"

# Today's Plan

## **Histogram!**

Let's do a case study first ...

# The Escalator Problem

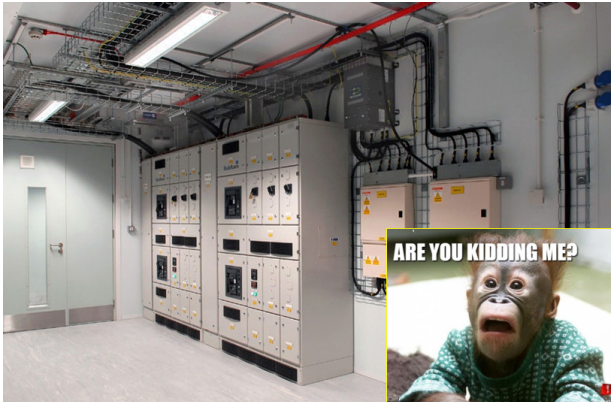


Energy efficient escalators:

- ▶ ON when there are pedestrians
- ▶ STAND-BY when there is no pedestrian for several seconds
- ▶ How much saving?

# That's Easy!

- ▶ Go to the meter room, and
- ▶ Measure it!!!



But what if you have not yet built the escalator?

# Let's collect data

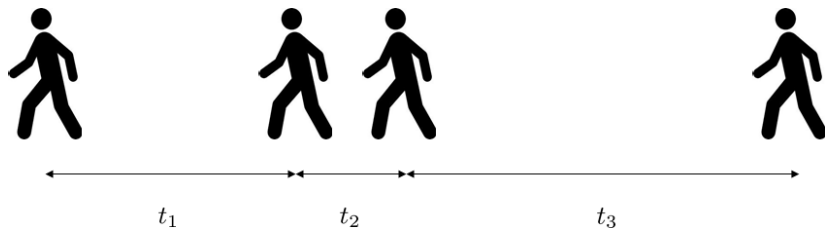




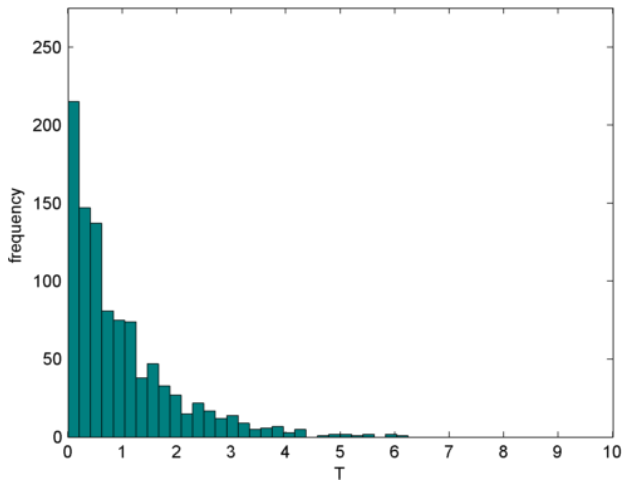
# Inter-arrival Time

Let  $T$  be the inter-arrival time.

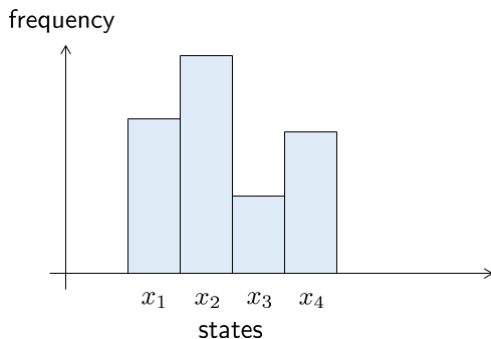
Possible values of  $T$ : Call them  $t_1, t_2, t_3, \dots$ ,



How does the histogram of  $T$  look like?



# What can be told from a histogram?



- ▶ Set of all possible state:  $x_1, x_2, \dots, x_m$ .
- ▶ Empirical **frequency** of each state:  $\hat{p}_1, \hat{p}_2, \dots, \hat{p}_m$ .

**Important!**

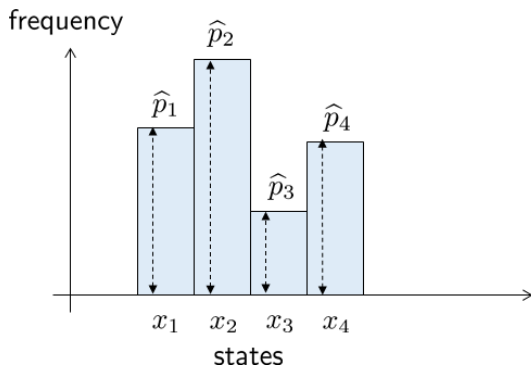
$$\hat{p}_1 + \hat{p}_2 + \dots + \hat{p}_m = 1.$$

# What can be told from a histogram?

## Sample Mean:

$$\bar{X} = \sum_{i=1}^m \hat{p}_i x_i$$

- ▶ “Average” of computed from the histogram
- ▶ Could be different if you run another experiment



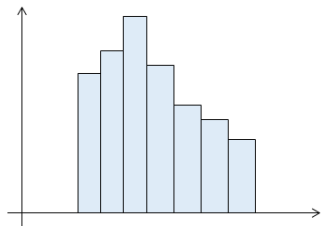
# What can be told from a histogram?

## Sample Variance:

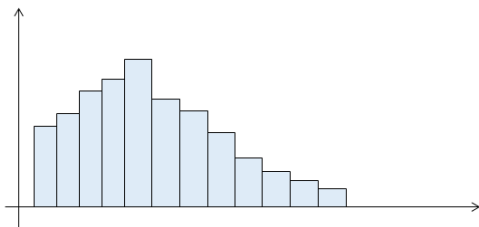
$$S^2 = \sum_{i=1}^m \hat{p}_i (x_i - \bar{X})^2.$$

- ▶ Measures the deviation
- ▶ Large  $S^2$  means that the histogram is wide-spread
- ▶  $S$  is the sample standard deviation

small variance

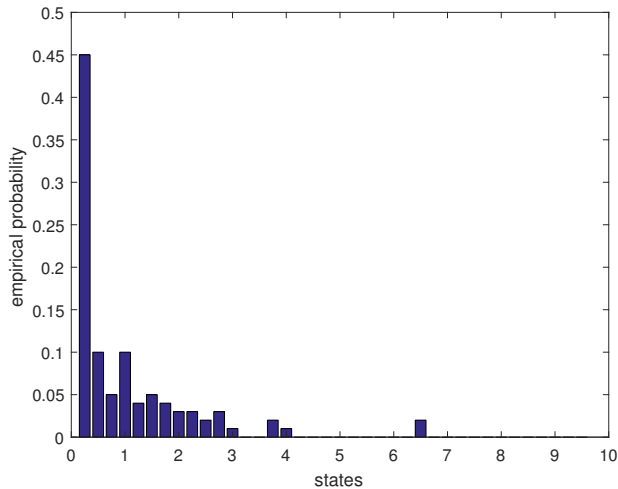


large variance



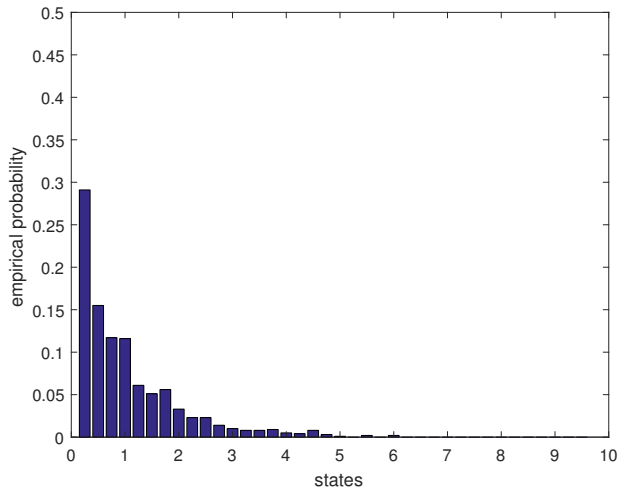
# Histogram Grows

What if we have 100 measurements?



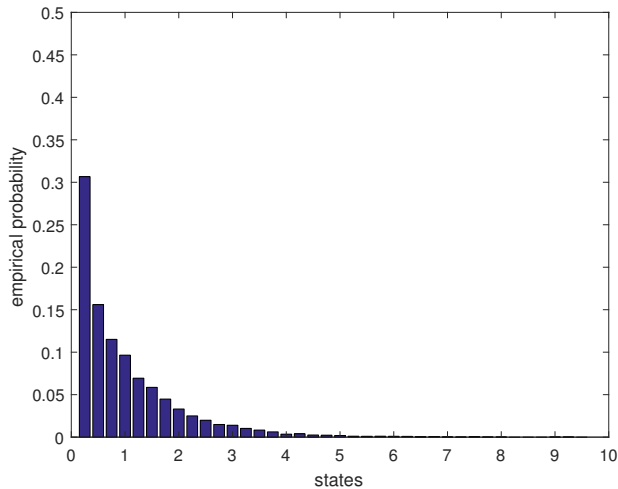
# Histogram Grows

What if we have 1000 measurements?



# Histogram Grows

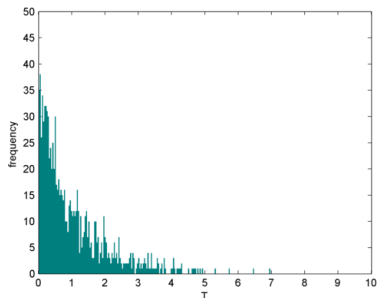
What if we have 10000 measurements?



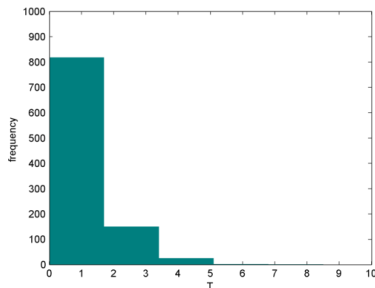


# Bin-width of Histogram

Bad choice of bin-width:



200 bins



5 bins

- ▶ Too many bins: Not enough data!
- ▶ Too few bins: Not descriptive!

# Optimal Bin-width

Here is a method to estimate the bin-width. The method is called **Cross-Validation**.

## Notations

- ▶  $n$ : number of data points
- ▶  $m$ : number of bins
- ▶  $h$ : bin-width:  $n/m$ . (Can round off to nearest integer.)
- ▶  $\hat{p}_j$ : frequency of the  $j$ -th bin.

Cross-validation Score:

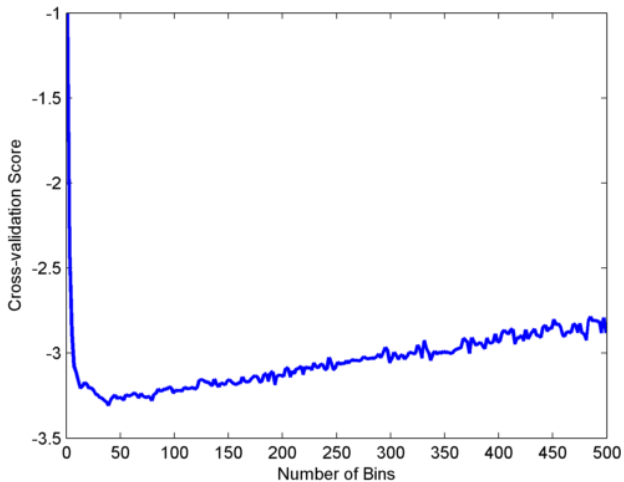
$$J(h) = \frac{2}{(n-1)h} - \frac{n+1}{(n-1)h} (\hat{p}_1^2 + \hat{p}_2^2 + \dots + \hat{p}_m^2).$$

# Optimal Bin-width

## Procedure:

- ▶ Pick the number of bins  $m$ .
- ▶ Since  $n$  is fixed, we can compute  $h = n/m$ .
- ▶ Build a histogram of  $m$  bins.
- ▶ The heights of the histogram bars are  $\hat{p}_j$ .
- ▶ Calculate the Cross-Validation Score  $J(h)$ .
- ▶ If  $J(h)$  is high, try another  $m$  until  $J(h)$  is low enough.

# Optimal Bin-width



# Summary

## Histogram:

- ▶ The most **basic** tool we use to analyze data.
- ▶ **Three components**: states, empirical probability, bin-width.
- ▶ Bin-width can be controlled by **Cross Validation**.
- ▶ **Sample Mean**: average of computed from the histogram.
- ▶ **Sample Variance**: deviation found of the states in the histogram.
- ▶ High-dimensional histograms.