Data representation

Lecture 2

IFT6758, Fall 2020; Reading: <u>IDS</u> - Chapters 8, 9, 10



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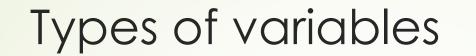
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"The greatest value of a picture is when it forces us to notice what we never expected to see."

PART-1: Visualizing data distributions



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 - ▶ Numerical: Height (continuous), Price (continuous), Population sizes (discrete)...

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 - ► Numerical: Height (continuous), Price (continuous), Population sizes (discrete)...
 - Discrete numeric data can be considered ordinal.
 - Conventionally, ordinal for variables belonging to a small number of different groups, with each group having many members: e.g.: the number of packs of cigarettes a person smokes a day, rounded to the closest pack
 - Discrete numerical for many groups with few cases in each group: the actual number of cigarettes in each pack



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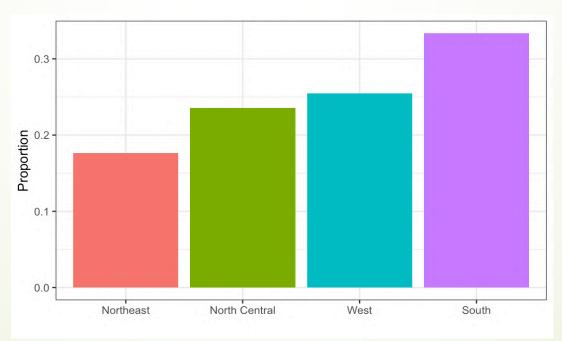
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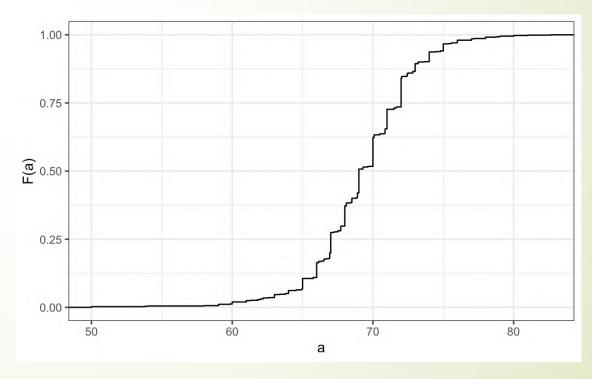
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Male height data



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- In statistics, the following notation is used: $F(a) = Pr(x \le a)$
- F(66) = 0.164, F(72) = 0.841
- Does not answer:
 - At what value is the distribution centered?
 - Is the distribution symmetric?
 - What ranges contain 95% of the values?



Histograms

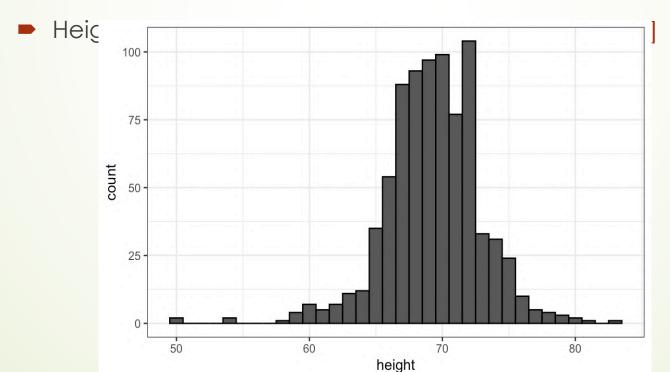
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Histograms

- Divide the span of our data into non-overlapping bins of the same size
- For each bin, we count the number of values that fall in that interval

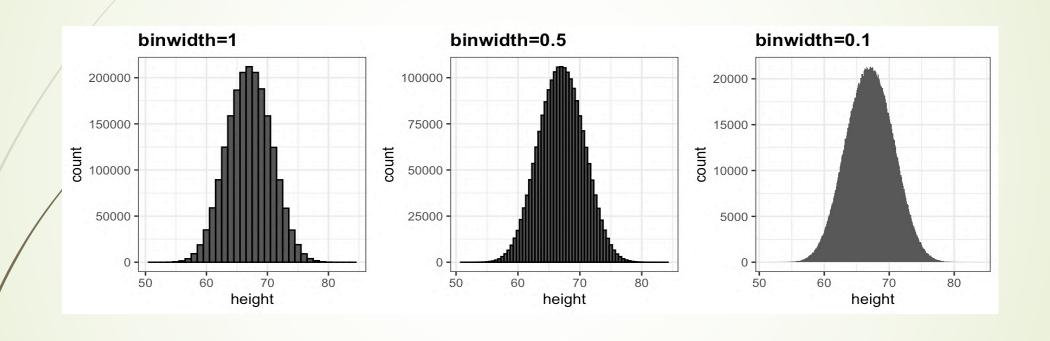
Histograms

- Divide the span of ou9.5, 50.5, 50.5, 51.5, ..., (82.5, 83.5)
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- For each bin, we count the number of values that fall in that interval

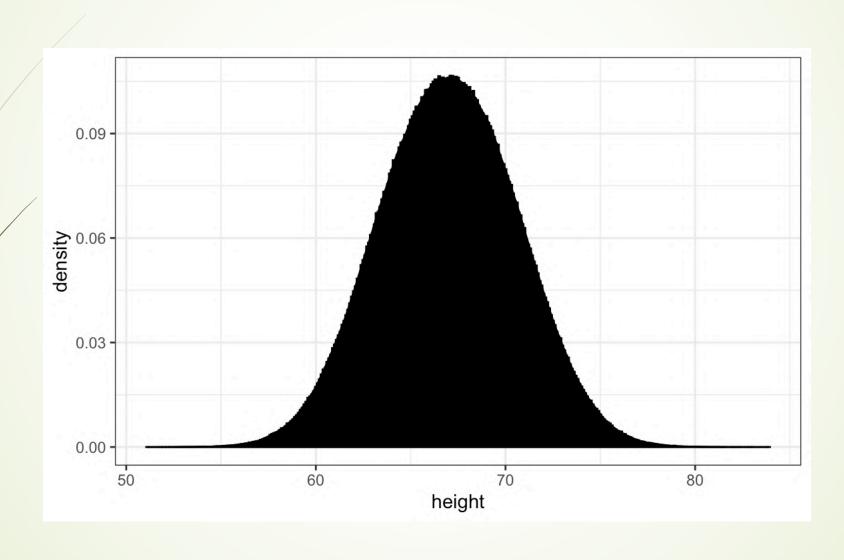




From histogram to smooth density

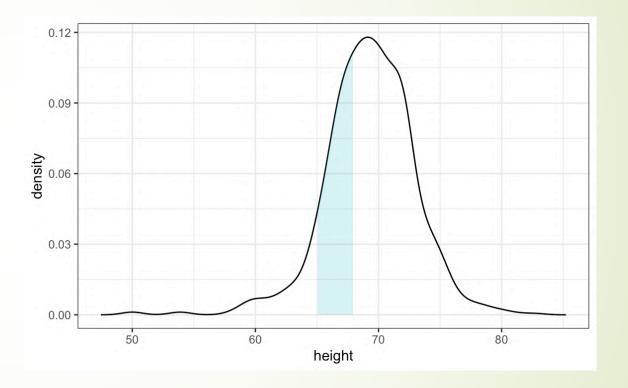


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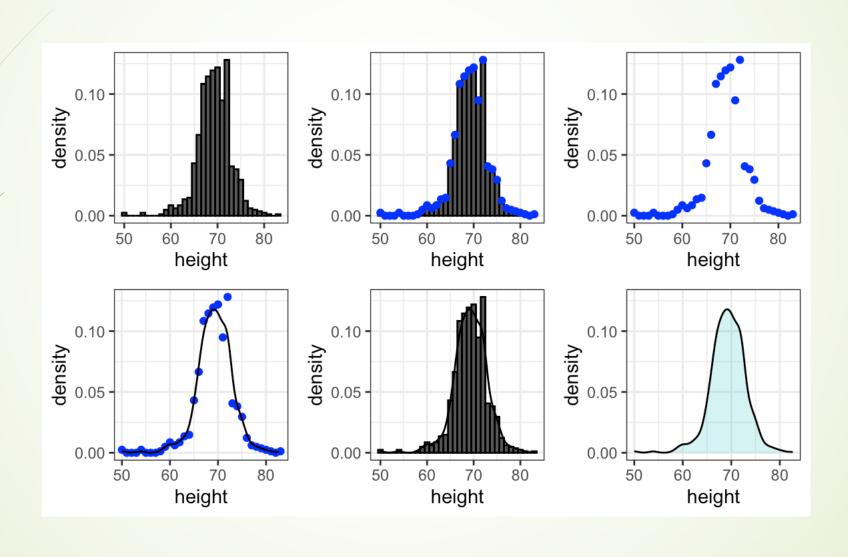


What is density?

- Proportion of values between 65 and 68
- The proportion of this area is about 0.3, meaning that about 30% of male heights are between 65 and 68 inches.

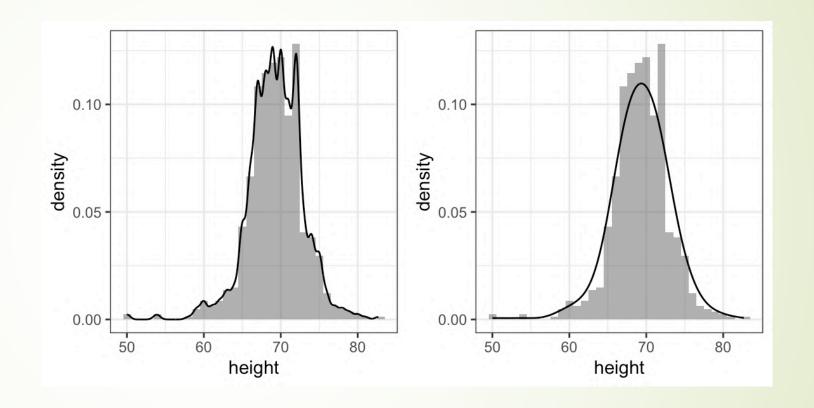


The transition to smoother density



Smoothness is relative

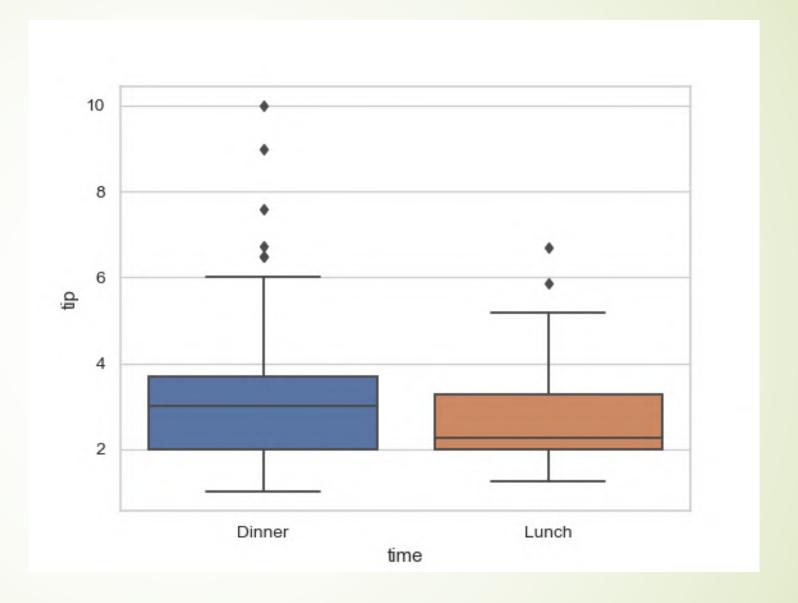
- Kernel Density Estimator
- Scipy, Scikit-Learn
- Smoothness varies with Bandwidth



Boxplots

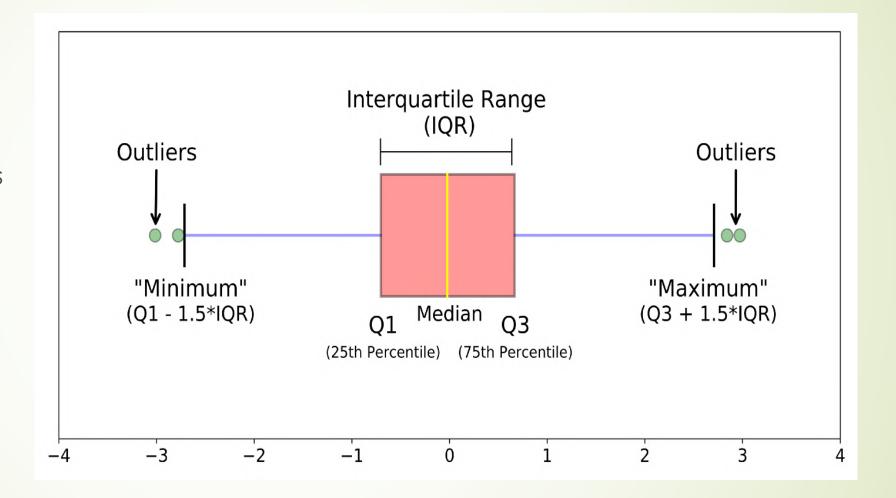
Boxplots

Matplotlib, Seaborn



Boxplots

- Matplotlib, Seaborn
- Helps detecting outliers



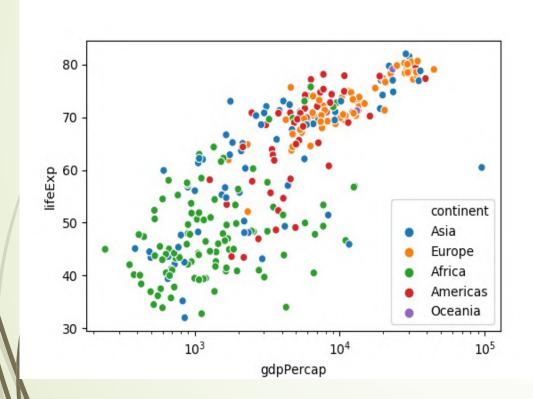


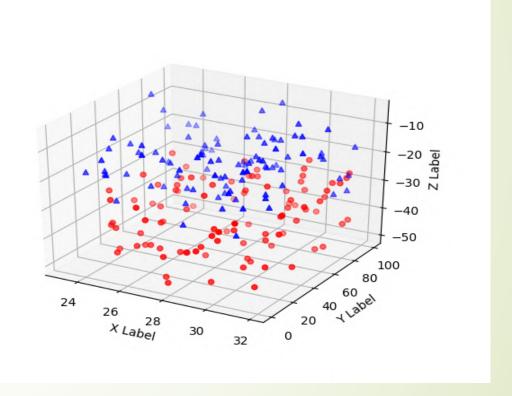
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