Nonparametric methods

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Nonparametric methods

- You have probably encountered the term "nonparametric" (statistics/models/methods/...), and likely will again
- There are two rather different meanings that "nonparametric" means in the context of statistics:
- Methods that do not assume a distributional form underlying the data
 - E.g., sign test; Wilcoxon signed-rank test; the bootstrap
- Methods that involve parameters, but whose expressive flexibility grows with data scale and complexity
 - E.g., kernel density estimation; k nearest neighbors; certain hierarchical Bayesian models; GAMs
- Wikipedia's nice "parent" characterization:

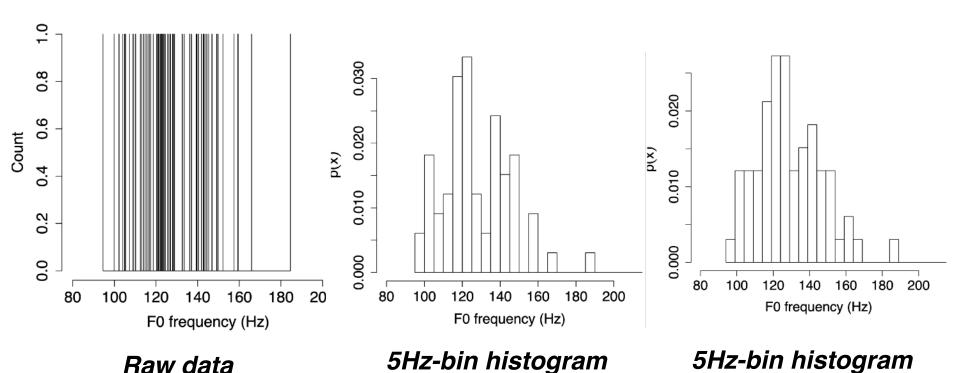
Nonparametric statistics

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From Wikipedia, the free encyclopedia

Example 1: kernel density estimation

- Task: estimate the distribution of adult male American English speakers' F0 ("pitch") formant for the vowel a
- With histograms:



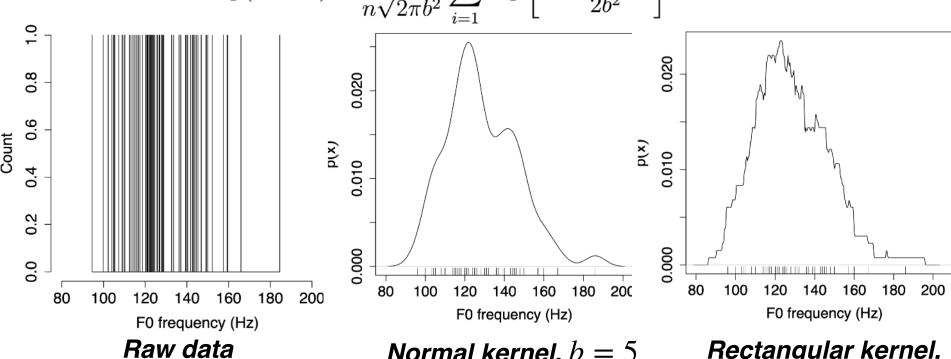
starting at 95Hz

starting at 94Hz

Example 1: kernel density estimation

- **Kernel:** an integrable non-negative real function K, typically **normalized** (integrates to 1) and **symmetric**
 - Examples: the normal and rectangular kernels
 - Normal kernel density estimate for **bandwidth** b:

$$\hat{p}(X=x) = \frac{1}{n\sqrt{2\pi b^2}} \sum_{i=1}^{n} \exp\left[-\frac{(x-x_i)^2}{2b^2}\right]$$



Normal kernel, b = 5

Rectangular kernel, b = 20

Example 2: the sign test

- Consider a vector of real-valued observations $y_1, \dots y_n$
- Assume they are generated IID
- Null hypothesis H_0 : P(y > 0) = 0.5
- This is a special case of the binomial test

Example 3: Wilcoxon rank sum test

- Consider two real-valued samples, $x_1, ..., x_m$ and $y_1, ..., y_n$, each IID
- Null hypothesis H_0 : the two samples come from the same distribution

Connecting to parametric statistics

- What is the relationship between the t test, the sign test, and the rank sum test (also the signed rank test which we didn't cover just now)?
- Mini-practicum: When would you want to use one versus the other? What are the pitfalls involved?

The bootstrap

- Common use case: confidence intervals
- Previously in this class, we have used parametric assumptions (typically some kind of normality) to compute confidence intervals
- When these assumptions are wrong, it can affect our confidence intervals
- The idea of the bootstrap: use the distributional properties of the data itself to estimate the statistic you're interested in
- Simplest use may be case resampling
- Mini-practicum: implement case resampling for estimating confidence intervals on non-normally (and normally, for sanity checking) distributed data