An (extremely) Abbreviated Introduction to Data Science with sklearn

Doug Lloyd CS50 Hackathon Amsterdam 14 June 2018

Slides

• Slides and code are available at cs50.ly/amsterdam-ds. No need to write everything down now!

Data Science

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 For us, talking about applying programming techniques to process data for various purposes.

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 Come up with questions about that data you want to answer, then write code to answer those questions.

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- 100% complete (no "empty data")
- Fairly small, only ~500 rows and 14 columns worth of data.
- Downloadable in CSV form at cs50.ly/boston-data. (You'll see why this is available separately in a moment.)

• Collected by the U.S. federal government as ancillary census data.

 Useful because it has a number of variables that can be used as part of different regressions, almost always used as a test case for predicting median home value in general.

• Extremely popular data set to introduce data science (see also, *Titanic* data).

1978 Boston Housing Data

- crim per capita crime rate
- zn % of land zoned for lots over 25k sq. ft.
- indus % of land in town zoned for non-retail business
- chas boolean for if property borders Charles River
- nox concentration of nitric oxide
- rm avg rooms per dwelling
- age % of units built prior to 1940

1978 Boston Housing Data

- dis weighted distance to major employment centers
- rad accessibility to radial highways
- tax tax rate per \$10,000
- ptratio student-teacher ratio
- **b** a formula determining minority population
- **1stat** % of the town population that is "lower class"
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• sklearn is a very popular module that can be used for data regression and analysis, simplifying many of the calculations that you might otherwise do on your own!

Step by Step

• Now, let's walk through some basic introductory steps one might take for tinkering with this data.

boston-0.py

 Here, let's see if we can get the data properly imported so we can test some other stuff later.

 Looking at the output of this program, generated by line 13's print(boston), what is the data type of the Boston data?

boston-1.py

 Let's dig in a little bit deeper, and check out just the size of our data set, to confirm that it matches what we're expecting.

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- Hmm.... That doesn't seem quite right. Our dataset is supposed to have 14 columns.
- As it turns out, when used for testing, regression usually happens on the median value, the missing 14th column, which is actually in the "target" key-value pair. We could splice it in, but let's leave it out for now!

boston-2.py

 Using a popular Python data analysis module, pandas, we can determine some relatively basic calculations for each of the columns, such as mean, median, mode, and different quartile values.

 We can examine lots of different things about the data this way.

boston-3.py

• But what we really want to do is **analyze**! And we also probably want to actually see the data looking nice in a graph.

• This is where the machine learning aspect of things comes into play, and before we look at the code, we need to talk about *training sets* and *test sets*.

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- By doing a pairing of expected inputs to expected outputs on the machine's behalf (and having it perform analysis on the "perfect" data), the machine learns your model and readies itself to apply the model to more interesting (real) data.

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- Don't make the training set too large!

Target

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• Let's create our datasets!

boston-4.py

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• In a perfect world, the data would follow a single straight line or at least some polynomial.

Things to Try!

- What happens if we think some of the columns are irrelevant (like chas)? And we don't want them in our data?
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 This is why we have numpy!
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 - See **boston-5.py**, where we try to plot a *linear regression* and calculate a mean-squared error.
- What if we try a polynominal regression (e.g., n = 2)? Dig into sklearn!

Thanks!