

**DATA 3464: Fundamentals of Data Processing**

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# **Basic machine learning models**

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# This week's topics

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- Exploring and understanding your data
- Splitting your data
- Assignment 1: Exploring Calgary traffic data

**Resources used:**

# Basic things to look at

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- Data source - File? Database? API?
- Structured/unstructured
- Assumption 1: relatively small (fits in memory) tabular dataset
  - Data types - numeric/categorical, text, other
  - Assumption 2: numeric data
    - Ranges
    - Summary statistics
    - Missing values
- Next week: categorical data, after reading week unstructured

# Example: Anscombe's Quartet

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- Very small dataset, constructed by hand in 1973 by [Francis Anscombe](#)

Most textbooks on statistical methods, and most statistical computer programs, pay too little attention to graphs. Few of us escape being indoctrinated with these notions:

- (1) numerical calculations are exact, but graphs are rough;
- (2) for any particular kind of statistical data there is just one set of calculations constituting a correct statistical analysis;
- (3) performing intricate calculations is virtuous, whereas actually looking at the data is cheating.

- Not known exactly how he made it, but Drs. Roberta La Haye and Peter Zizler proposed a [compelling argument](#) for linear algebra

# Useful starting points

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- `pandas.DataFrame.info` : data type, number of non-null, names, dimensions
- `pandas.DataFrame.head` : return the first `n` rows (default 5)
- `pandas.DataFrame.describe` : Compute a bunch of summary statistics
- As soon as you have a general sense of the:
  - Data scales
  - Missing features
  - Distributions, particularly categorical
- It's time to split the data!

# Splitting your data - why

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- We need to set aside a final **test set** to evaluate our final model
- Humans are great at detecting patterns!
- Even looking at test data could influence decisions, causing **data leakage**

# Splitting your data - how

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# Types of exploratory visualizations

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- I will not provide an exhaustive list of visualizations!
- Pandas provides a [handy wrapper](#) around [matplotlib](#)
- So does [Seaborn](#) - check out the [example gallery](#)
- Some of my favourites:
  - Histograms
  - Scatter plots/hexbin plots
  - Box plots/violin plots



# Some simple tricks

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Try tweaking:

- Histogram bin sizes
  - Aiming for a smooth distribution that works for your data
- Transparency ( `alpha` )
  - Useful for both dense scatter plots and overlapping categories
- "Jitter"
  - Mostly for scatter plot of continuous vs categorical data
  - Add a tiny bit of random noise to spread out samples