

Background

Goal: Modernize my second semester, undergraduate statistics course. Want course to satisfy two popular but conflicting ideas:

- Teach the entire data analysis workflow, of which modeling is only one step.
- Teach a more diverse set of models, especially statistical learning techniques.

Problem: How do I find time to teach more of the data analysis workflow and to cover new modeling techniques?

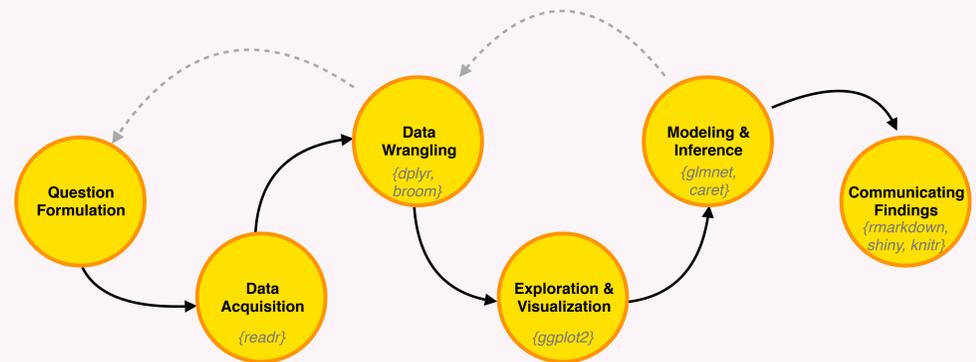
Proposed Solution:

- Streamline the process of teaching the data analysis workflow using the Tidyverse.
- Shorten the discussion of specific regression models.
- Use freed up class time to cover predictive modeling techniques.

Examples: In this poster, I present example activities which:

- Use Tidyverse packages.
- Emphasize the importance of the Data Wrangling and the Exploration and Visualization steps.
- Reflect an iterative approach to the data analysis workflow.
- Include statistical learning methods.
- Follow a reproducible workflow.

Data Analysis Workflow



Case Study 1: Are volcanic eruptions increasing?

Question Formulation:

- After learning simple linear regression, the students can frame this problem as:
 - *Is there a positive, linear relationship between time and number of eruptions?*

Data Acquisition:

- Data file from the Smithsonian Institution's Global Volcanism Program website.

`read_csv()`

Eruptions

```
## # A tibble: 11,078 × 24
##   Volcano_Number Volcano_Name Eruption_Number
##   <int>          <chr>          <int>
## 1      282080      Aira             22203
## 2      300010      Kambalny          22198
## 3      262000      Krakatau          22188
```

Data Wrangling:

- Filter by date and confirmed eruptions.
- Group by start year.
- Record year, number of eruptions, and average size of eruptions.

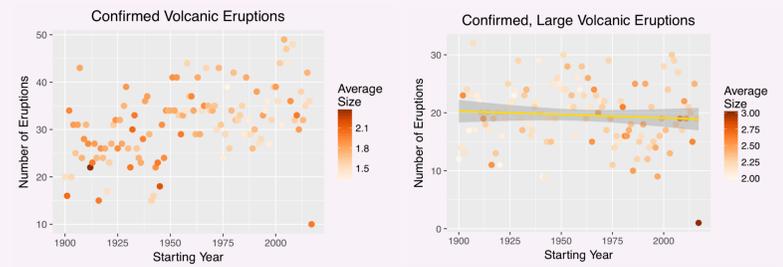
```
dataset %>%
  filter() %>%
  group_by() %>%
  summarize()
## # A tibble: 118 × 3
##   Start_Year count avg_VEI
##   <int> <int> <dbl>
## 1      1900     20  1.500000
## 2      1901     16  2.066667
## 3      1902     34  1.941176
```

Exploration and Visualization:

- Create scatterplots.


```
ggplot() +
  geom_point() +
  stat_smooth()
```
- Sampling bias issues:
 - World events impacting reporting.
 - Detection dependent on size of the eruption over time.
- Add one more wrangling argument to try to minimize bias.


```
filter()
```



Modeling and Inference:

- Construct model and summary table.

```
lm() %>%
  tidy() %>%
  kable()
```

term	estimate	std.error	statistic	p.value
(Intercept)	42.42	28.37	1.5	0.14
Start_Year	-0.01	0.01	-0.8	0.42

- Not a significant relationship.

Communicating Findings:

- Students write up their work using RMarkdown.
- Students also use this data to construct interactive maps of the world's volcanoes using *shiny* and *leaflet*.

Case Study 2: Build a model for household income.

Question Formulation:

- When covering model selection techniques, the students complete the following task:
 - *Build a model for income. Conduct model selection to determine an appropriate set of predictors.*

Data Acquisition:

- Data from the US Bureau of Labor Statistics Consumer Expenditure Survey.
- Two files from the fourth quarter of 2015:
 - Household data
 - Data on each individual

Data Wrangling:

- Merge the principal earner's information into the household dataset.

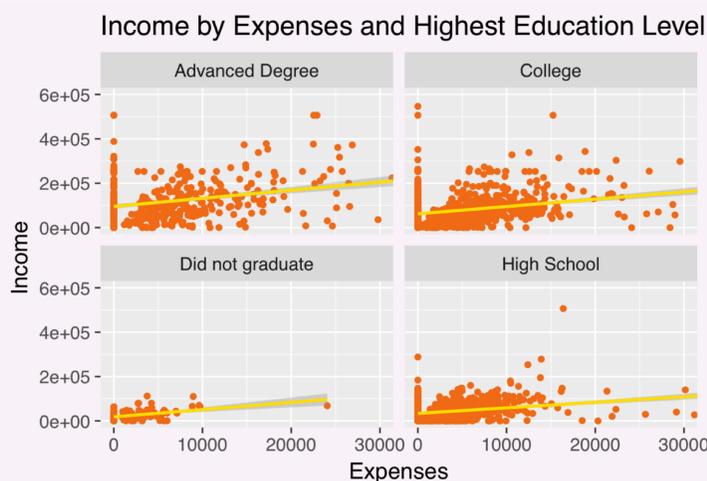
`left_join()`

- Resulting in 2,469 households.

Exploration and Visualization

- Students construct graphics to explore multivariate relationships.

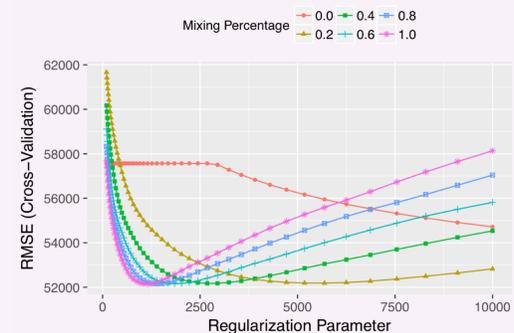
```
ggplot() +
  geom_point() +
  stat_smooth() +
  facet_wrap()
```



Modeling and Inference:

- Consider full two-way interaction model with 1,030 potential variables.
- Fit an elastic net model.
- Use cross-validation to select hyperparameters.

```
trainControl(method = "cv")
train(..., method = "glmnet")
```



- Resulting model contains 163 variables.

Communicating Findings:

- In an RMarkdown report, students compare the performance of the selected models between stepwise selection and elastic net and draw conclusions about how the predictors relate with income.

Conclusions

- Students get a lot of satisfaction out of making impressive plots with *ggplot2* and polished reports with *RMarkdown*.
 - This provides motivation to improve their skills and to overcome errors.
- Students struggle with data wrangling. My suggestions are:
 - Make LOTS of pictures.
 - Use the pipes to breakdown each step.
 - Stress the importance of the wrangling step to the entire workflow.
- Must drop some topics.
- With freely available or "found" data, it is so important to emphasize the potential pitfalls of generalizing results.

Acknowledgments

I would like to thank the Smithsonian Institution and the US Bureau of Labor Statistics for providing public use datasets. My classes have also greatly benefitted from the RStudio Server.

References

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