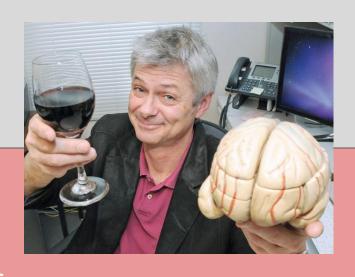
# Wine Sample Analysis

Economist - Andrew Dively GIS Data Scientist - Ashley Moss NLP Data Scientist - Patti Degner





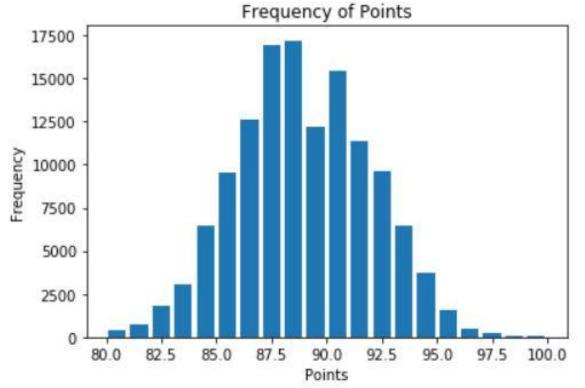
### Data Cleaning & Metadata

- Taster Name
- Description
- Wine Year & Wine Age
- Expected Points & Value
- Country Polygons & Country Codes
- Data Cleansing





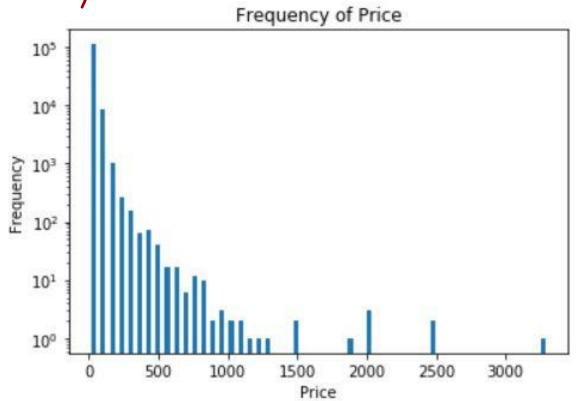
## Points Analysis







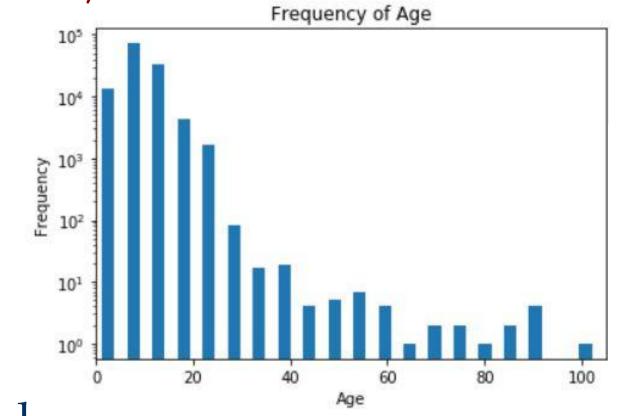
## Price Analysis







# Age Analysis



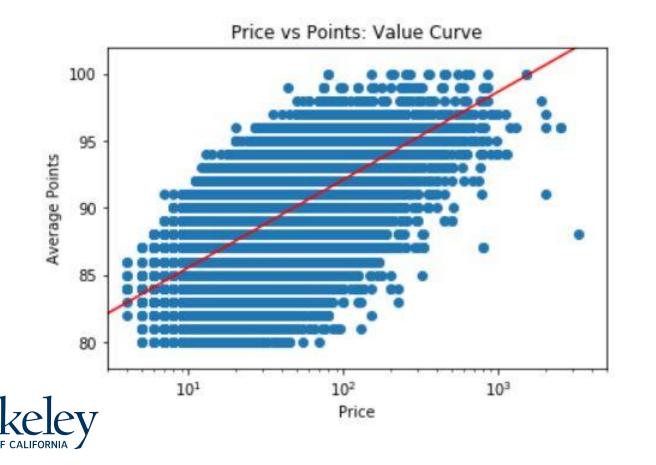


## Points Vs Price Range



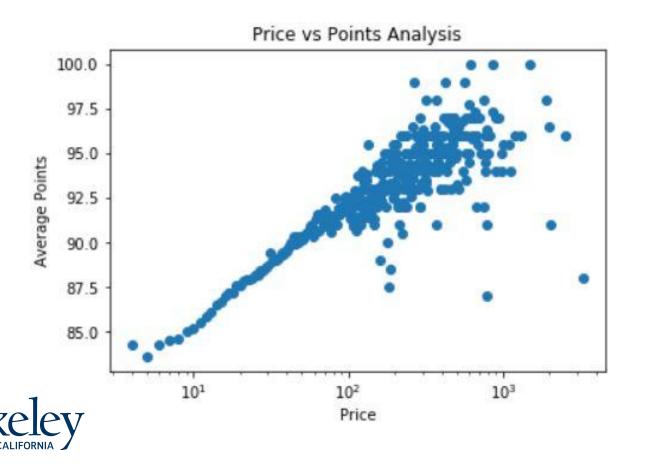


#### Price Vs Points: Value Curve



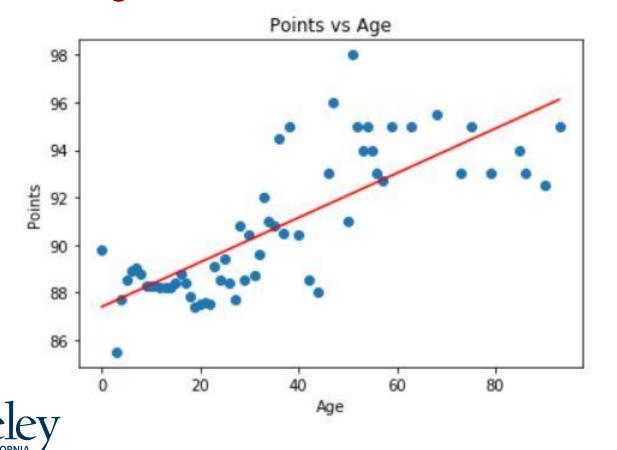


## Price Vs Average Points



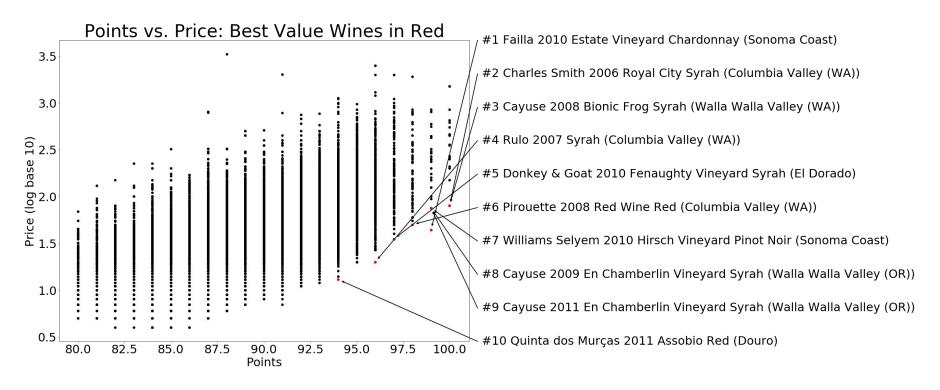


# Points Vs Age



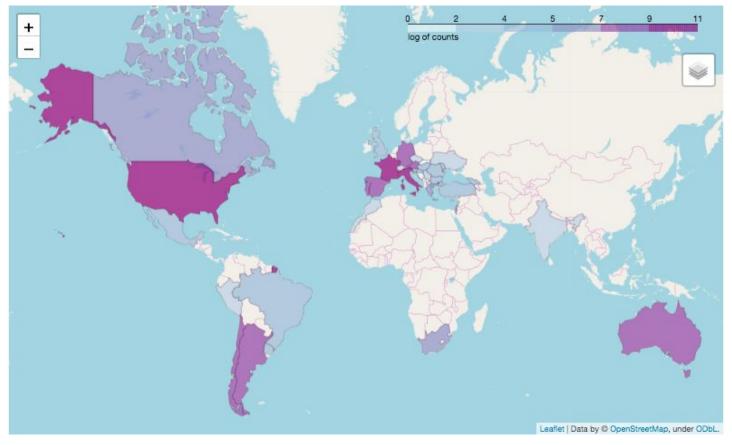


#### Top 10 Best Value Wines

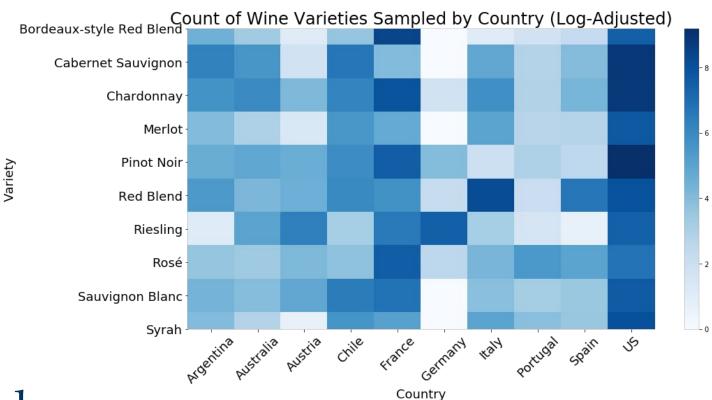




# Log-Adjusted Country Count

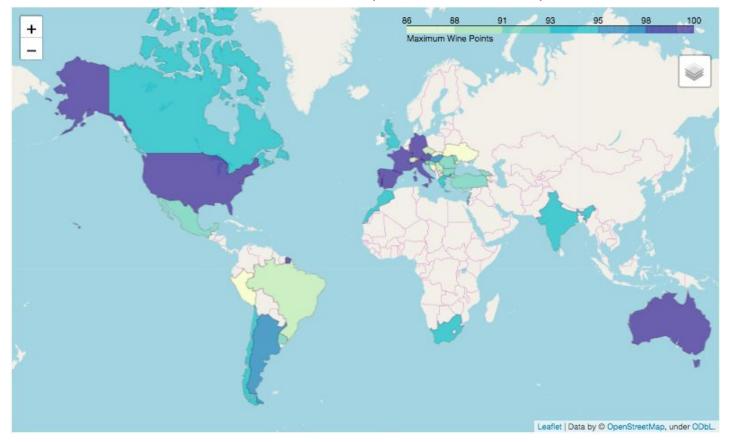


## Variety vs. Geography

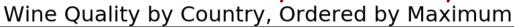


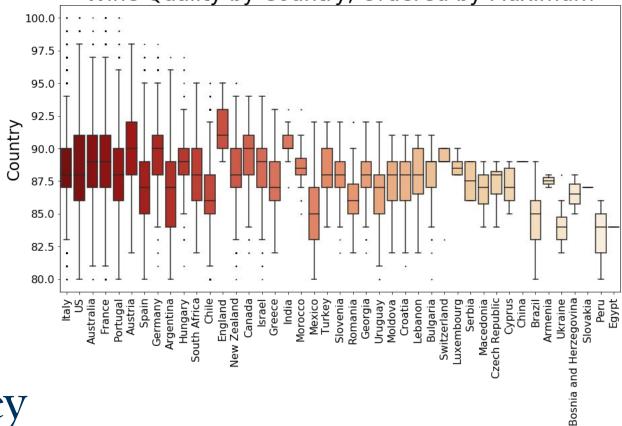


# Maximum Wine Score by Country



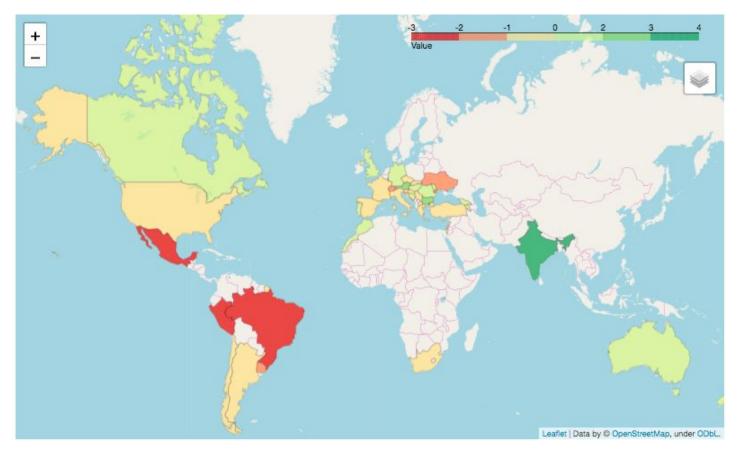
### Wine Score Distribution by Country



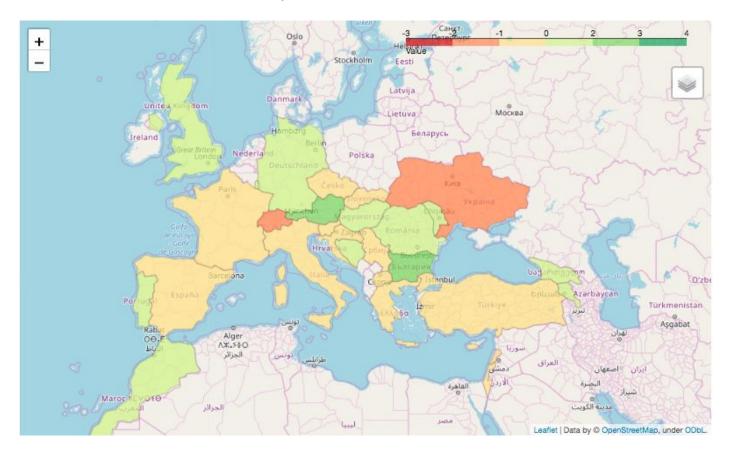




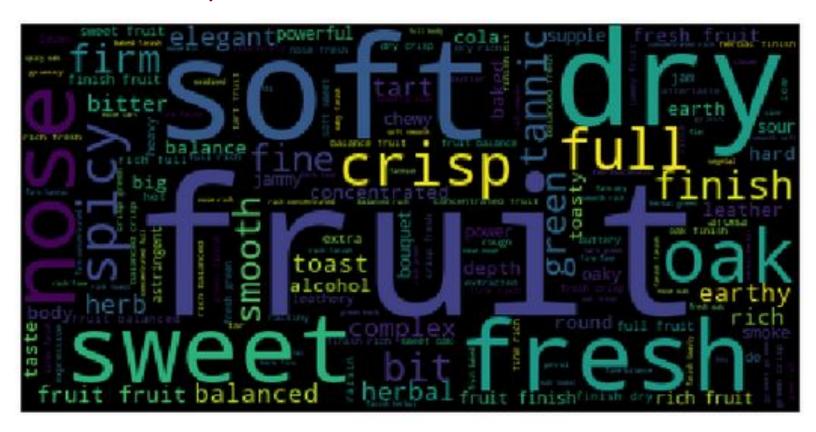
# Wine Value by Country



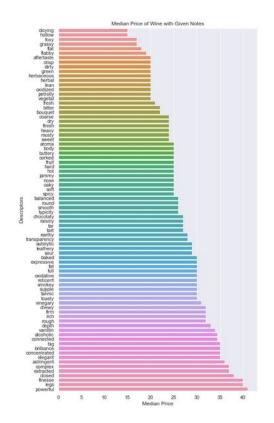
# Wine Value in Europe



## Wine Descriptors

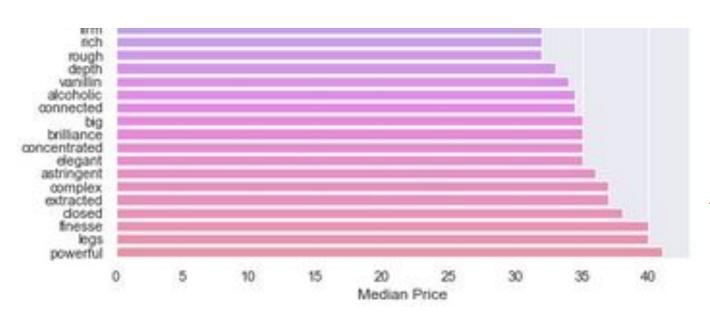


#### Median Price By Wine Notes



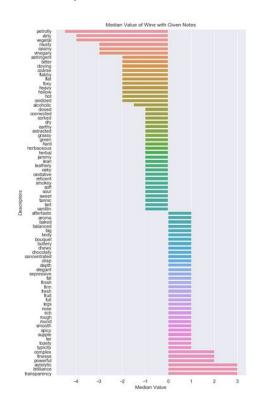
- 1. Powerful
- 2. Legs
- 3. Finessed
- 4. Dosed
- 5. Extracted
- 6. Astringent
- 7. Elegant
- 8. Concentrated
- 9. Brilliance
- 10. Big

### Median Price By Wine Notes



- 1. Powerful
- 2. Legs
- 3. Finessed
- 4. Dosed
- 5. Extracted
- 6. Astringent
- 7. Elegant
- 8. Concentrated
- 9. Brilliance
- 10. Big

# Median Value By Wine Notes



- 1. Transparency
- 2. Brilliance
- 3. Autolytic
- 4. Powerful
- 5. Finesse
- 6. Complex
- 7. Typicity
- 8. Toasty
- 7. Tar
- 10. Supple

## Median Value By Wine Notes



- 1. Transparency
- 2. Brilliance
- 3. Autolytic
- 4. Powerful
- 5. Finesse
- 6. Complex
- 7. Typicity
- 8. Toasty
- 9. Tar
- 10. Supple