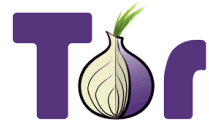


Workflow

Data Sources



Data Collection



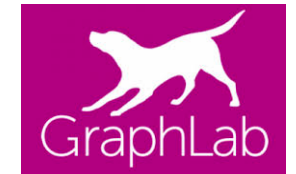
Data Munging



reg[ular]
expr[essio]n

Data Modeling

pandas
 $y_{it} = \beta' x_{it} + \mu_i + \epsilon_{it}$



Exploratory Data Analysis

Wine Data:

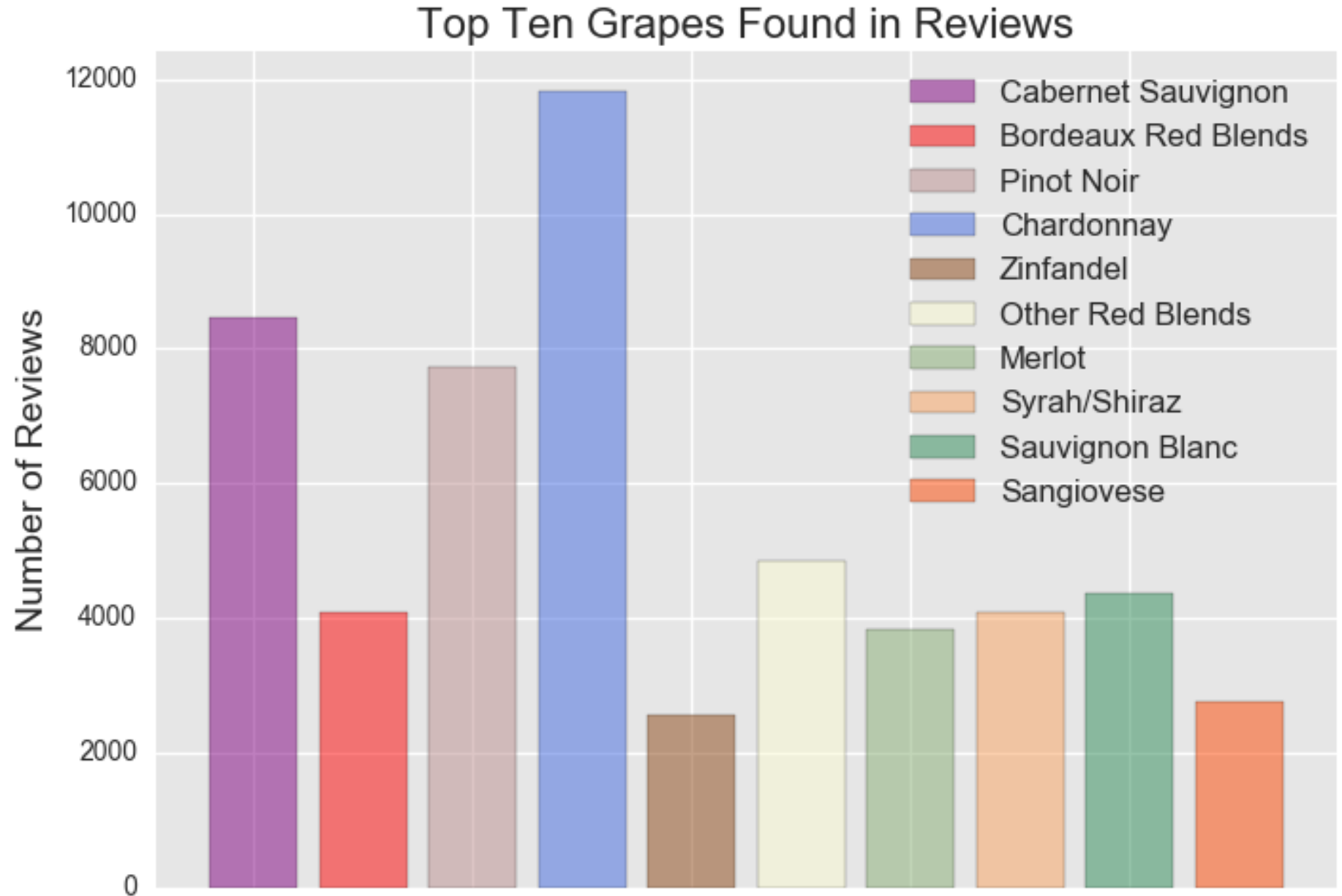
Num Wines:
78,076

Price Range:
\$1.98 – \$12,429

Wine Type:
Red: 61%
White: 35%

Grape Varietals:
164
Chard: 15%, Cab: 10%

Regions:
109
Napa: 10%



Exploratory Data Analysis

Review Data:

Num Reviews:

211,452

83% had a rating

Num Unique Users:

334

Num Wines Reviewed:

95,264

61% only had one review

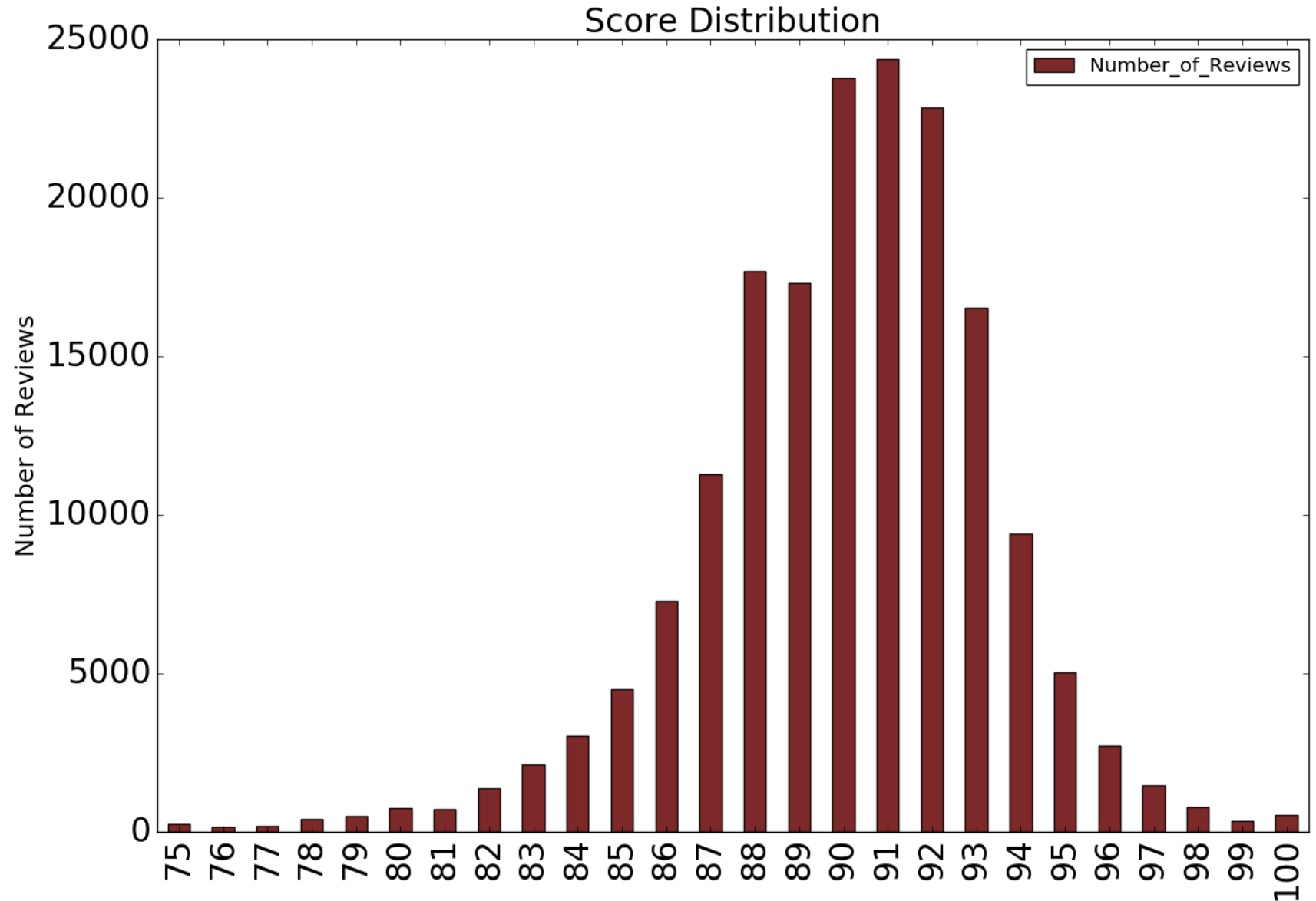
Rating Range:

0 - 100

Mode = 91

Mean = 89.85

Median = 90 $\sigma = 5$



Building the Recommender Models

I. Content-Based Filtering:

If you like an item then you will usually also like a “similar” item based on similarity of the items being recommended (rather than learned from the interaction data)

Similarity_type: cosine

Number of wines: 78076

Features: Wine Name, Vintage, Grape Varietal, Region/Location, Price, Avg Critic Rating

Predict on:

Grape	Location	Wine Name	Price	Avg Score
Cab Sauvignon	California	Freakshow Cab - 2014	20.99	90

Recommendations:

Grape	Location	Wine Name	Price	Avg Score
Cab Sauvignon	California	Smoking Loon Cab - 2014	8.99	85
Cab Sauvignon	California	Layer Cake Cab - 2014	14.99	90
Cab Sauvignon	California	Carnivor Cab - 2014	11.99	90



Building the Recommender Models

II. Collaborative Filtering: Item – User Similarity

Build a model to find users who share the same rating patterns with the user whom the prediction is for.

	Wine 1	Wine 2	Wine 3	Wine 4
Bob	91		?	89
Jake	90		92	93
Frank		82		
Kayla	92			90

Algorithm based on:

- K-nearest neighbors
- Pearson correlation

Recommendations:

user_id	wine_name	score	rank
bob_mickus	2003 E. Guigal Côte-Rôtie ...	98.5	1
bob_mickus	2007 Maybach Family Vineya...	98.25	2
bob_mickus	1959 Château Lafite Rothsc...	98.0	3
bob_mickus	1989 Domaine Huet Vouvray ...	98.0	4
bob_mickus	2001 Pride Mountain Vineya...	97.3658929312	5
bob_mickus	1976 Joh. Jos. Prüm Wehlen...	97.25	6
bob_mickus	2009 Saxum James Berry Vin...	97.25	7
bob_mickus	1996 Domaine Leflaive Chev...	97.0	8
bob_mickus	1989 E. Guigal Côte-Rôtie ...	97.0	9
bob_mickus	1997 Romano Dal Forno Reci...	96.8333333333	10



“a full-bodied, concentrated wine that has sweet tannin, excellent mid-palate depth and a terrific finish....”

Model Comparisons

RMSE (on predicting 0 – 100 score):

Model 1:

Item – User Similarity	12.4
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Model 2:

Default Matrix Factorization	3.8
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Model 3:

Hyper-parameters Optimized on MF	2.3
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Model 4:

Include Price as a Feature	1.7
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Precision (on predicting a good wine)

Model 5:

Logistic Regression Classifier	.997
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Next Steps...

Next Steps:

- Conduct sentiment analysis on the text comments contained in the reviews and incorporate into the model
- Continue to download more reviews
- Build front-end UX
- Integrate the Perfect Pour with the Matrix
- "Test" some more wines!



Can I recommend a wine for you?

Bob Mickus

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Project Repository: www.github.com/bobmickus/perfectpour