

Submission**[10 Points]**

In your Jupyter Notebook, **Restart Kernel and Run All...** Then on Compass submit both

- the Jupyter notebook
- a PDF export

IF YOU HAVE CHANGED TOPICS SINCE FP2: Please upload new .pkl file to your box folder.

Content	Full Points	No Points
Restarted kernel and ran all	4 yes	0 no
Submitted Jupyter notebook	3 yes	0 no
Submitted PDF export	3 yes	0 no

Format**[10 Points]**

Content	Full Points	No Points
Removed all non-essential code	2 yes	0 no
Figure Setup		
Increase size of figures, axis labels, title, and ticks	4 1 per item	0 none
Typos/grammatical errors	1 none to some	0 many
Section headers for:		
Inference, Prediction, and Comparison	3 1 per header	0 missing

Inference**[20 Points]**

Use a reasonable train-test split. Using OLS, binomial logit, or multinomial logit appropriately...

1. Use cross-validation with LASSO to find the optimal α
2. Refit a regularized model on the full training data with optimal α
3. Refit a non-regularized model on the full training data using the features selected from step 2
 - Deal with dummy features accordingly
 - If using multinomial logit, remove a feature if 50% or more of the coefficients have been zeroed
4. Interpret the top three most significant marginal effects

Content	Full Points	No Points
Reasonable train-test split	3 yes	0 no
LASSO CV optimal α	4 yes	0 no
Refit with optimal α	3 yes	0 no
Refit with selected features	4 yes	0 no
Correct interpretations	6 2 per feature	0 no

Prediction**[50 Points]**

Models	Classification	Regression
Naïve Bayes	✓	
KNN	✓	✓
SVM	✓	✓
Random Forest	✓	✓
AdaBoost or XGBoost	✓	✓
Neural Network	✓	✓

Print and store the inference model's performance (model 1). If performing classification, you may either tune or use the default threshold in prediction.

If an individual ([partner](#)) project, choose two ([three](#)) models from the table above.

- Create subsection headers for each chosen model
- Train the models dealing with hyperparameters, random states, early stopping, and/or refitting appropriately
- Print and store your models' performance (if you have a **classification problem**, also print a confusion matrix for 2 points of total performance points)
- 5% bonus if you train and test an additional model-based stacking ensemble on the inference and two ([three](#)) chosen models

INDIVIDUAL				PARTNER			
Content	Full Points	No Points		Content	Full Points	No Points	
Model 1: performance	4 yes	0 no		Model 1: performance	2 yes	0 no	
Subsection headers	6 3 per	0 no		Subsection headers	6 2 per	0 no	
Model 2: train	10 yes	0 no		Model 2: train	7 yes	0 no	
Model 2: performance	10 yes	0 no		Model 2: performance	7 yes	0 no	
Model 3: train	10 yes	0 no		Model 3: train	7 yes	0 no	
Model 3: performance	10 yes	0 no		Model 3: performance	7 yes	0 no	
				Model 4: train	7 yes	0 no	
				Model 4: performance	7 yes	0 no	
Bonus ensemble	5 yes	0 no		Bonus ensemble	5 yes	0 no	

Comparison**[10 Points]**

In a markdown cell...

- Produce a table over model performance
- Compare all models relative flexibility and ease of interpretation
- Explicitly identify the best performing model according to the metric of your choosing

Content	Full Points	No Points
All components in a markdown cell	3 yes	0 no
Complete table of model performance	3 yes	0 no
Comparisons over all models	3 yes	0 no
Best performing model	1 yes	0 no