Wave-2, the Upsell

What we know:

- 75,000 Sample Size
- Basic Demographics
- Purchased Tax Software
- Linear Regression Results

What we want to know:

- Who will buy out of the remaining 760,000
- New models

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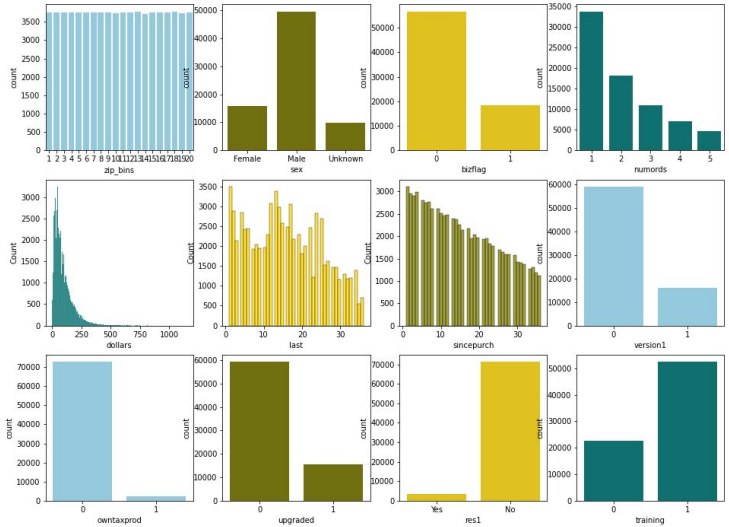
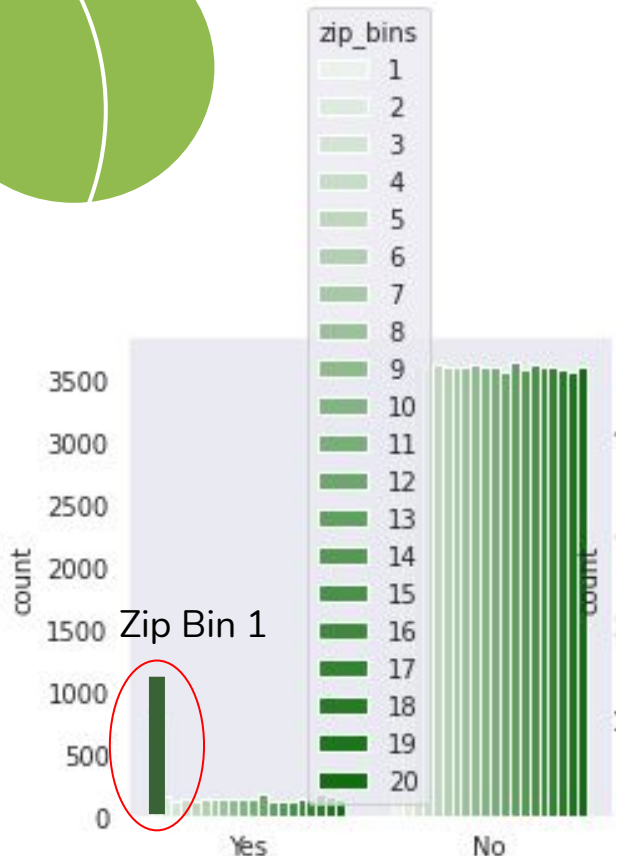


01 Exploring the Data

Dummy Variables / 3-Factor labels
Transforming Dollars

ZIP Bins & Dummy Variables

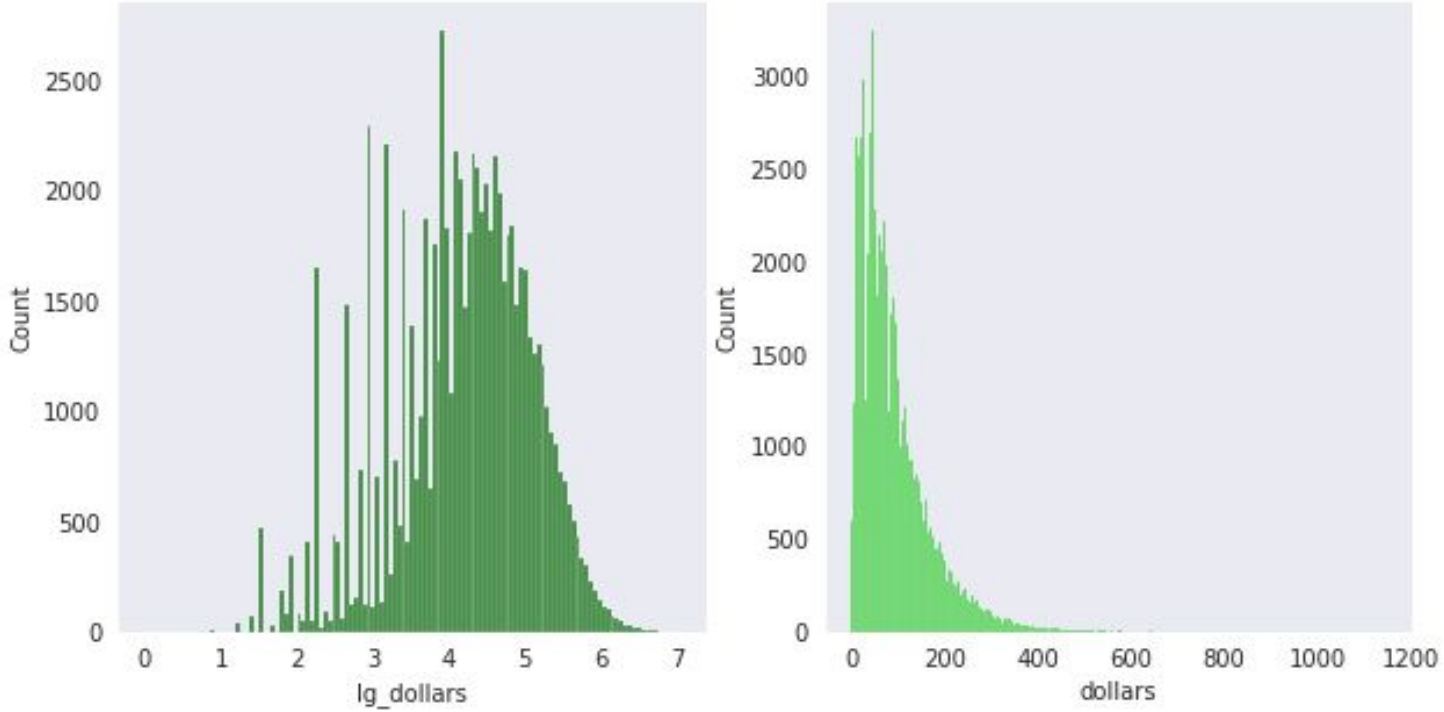


- Zip Bins
- Zip ‘00801’ & ‘00804’
- As New Dummy Variables

‘Upgraded’ & ‘Version 1’ 3 Factor Label

	Quickbooks Version 1 Only	Version 1 to Version 2	Quickbooks Version 2 Only
			
	Upgraded = 0 **Version 1 = 0**	**Upgraded = 1** **Version 1 = 0**	**Upgraded = 1** **Version 1 = 1**
New Variable:	only_ver02	from_v1_to_v02	only_ver01

Log Transform Dollars



02

Models and hyperparameters

Modeling Methods / Tuning Hyperparameters

Methods we used

Neural Networks to find and capture underlying relationships

Tree based models to capture predictions on highly non-linear and complex relationships

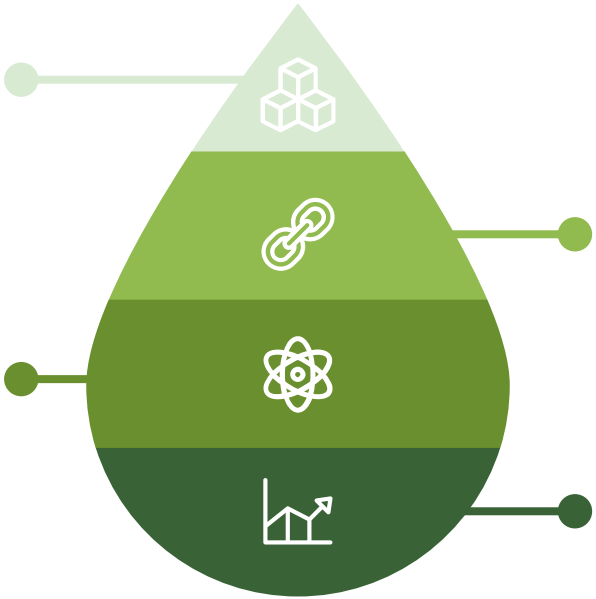
Method 1: Keras Neural Network

1. Specify Architecture

- ***Keras Sequential*** Model
- 100 to 300 nodes each layer
- 4 to 6 dense layers
- ***Relu*** activation
- ***Softmax*** activation

3. Fit

- Train on 52,500 data
- 30% validation split
- ***Early Stopping*** callbacks
- 20 epochs



2. Compile

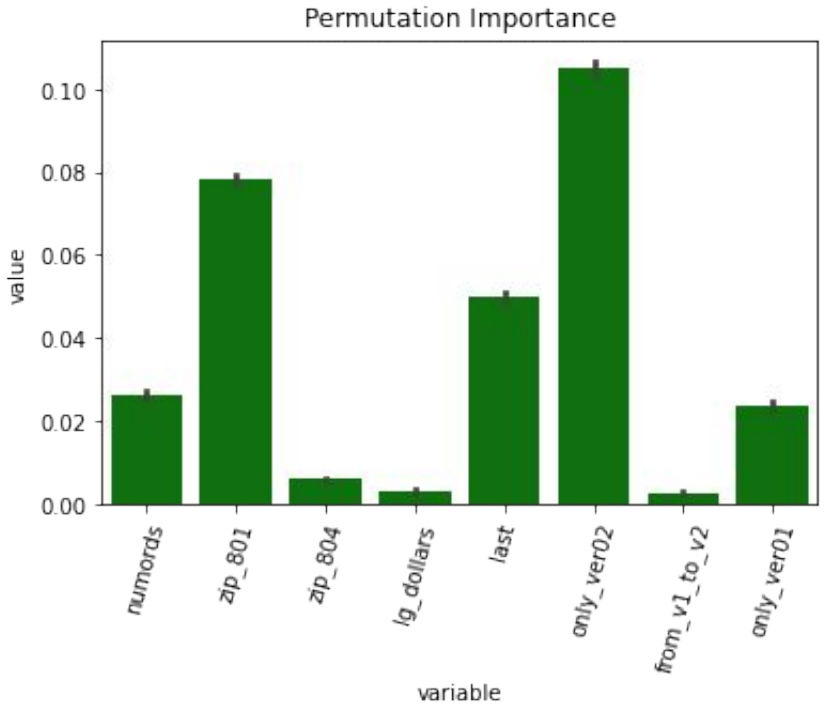
- ***Adam*** optimizer
- ***Categorical_crossentropy*** loss function
- ***Accuracy*** metrics

4. Predict

- Predict 22,500 test data
- Breakeven threshold

Best AUC Score: 76.5%

Method 2: MLP Neural Network



Activation = “tanh”
Solver = “lgfbs”
Max iter = 1000

Hidden Layer Sizes?

1 Layer: (1,), (2,), **(3,)**, (4,)

2 Layers: (1, 1), (2, 2), (3, 3), (4, 4)

Alpha?

0.001, 0.01, 0.05

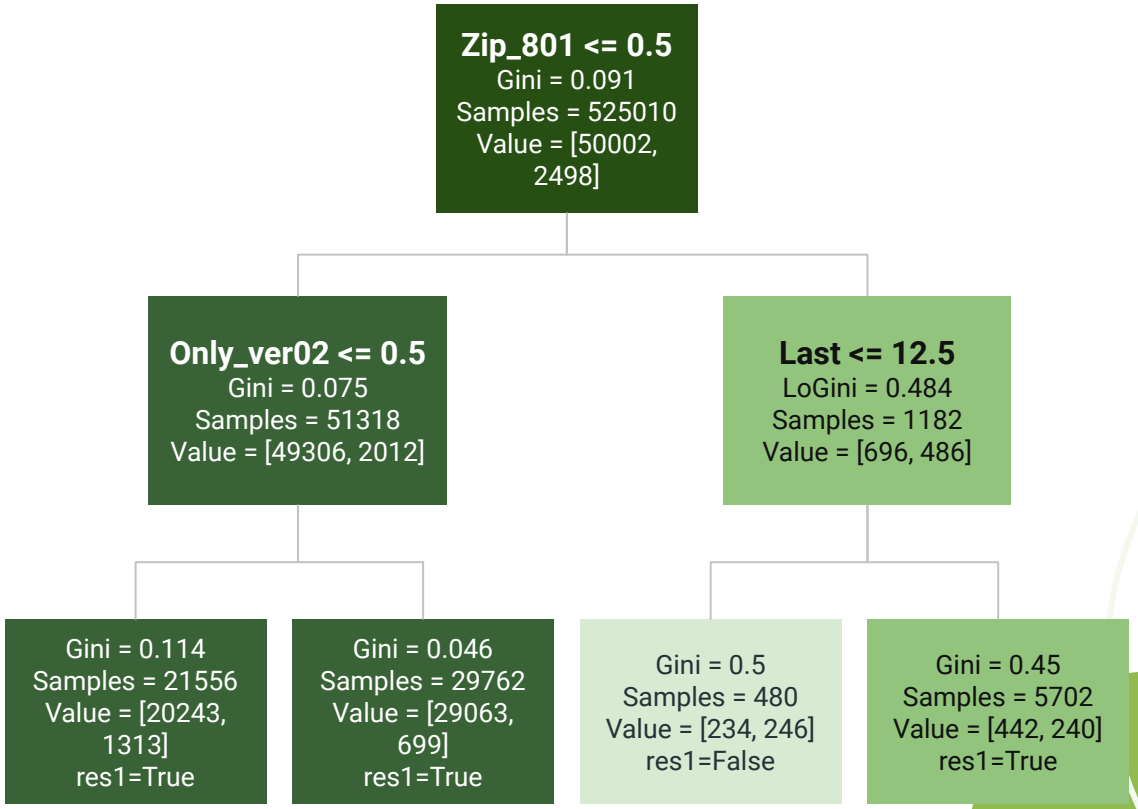
Best AUC Score: 77.02%

Method 3: Decision Tree Classifier

GridSearchCV: 5 Folds

AUC = 75%

Max depth = 6



Method 3: Decision Tree Classifier

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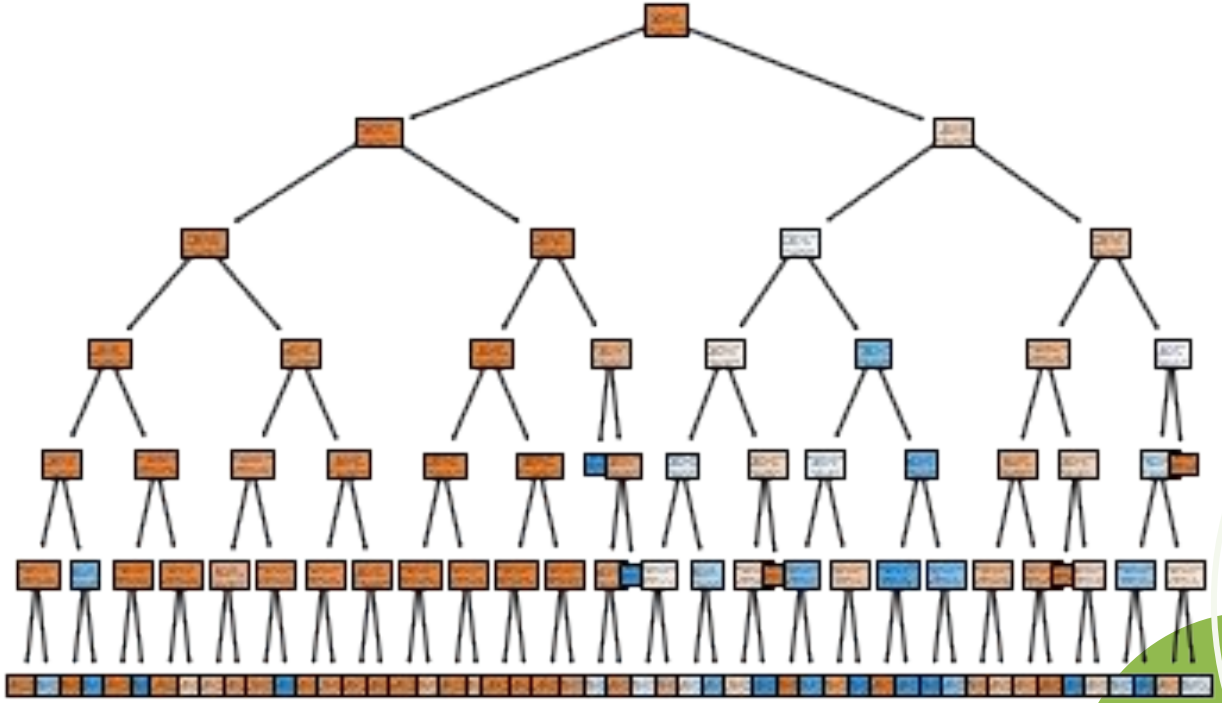


Method 4: Random Forest Classifier

1. Bootstrapping
2. Ensembling

max_features

[**auto**, sqrt]

n_estimators

[**50**, 100]

max_depth

[2, ,**5** 10]

min_samples_split

[**2**, 5]

bootstrap

[**True**, False]

min_samples_leaf

[**1**, 2, 4]

GridSearchCV with 4 folds:

Best AUC Score: 76.64%

Method 5: XGBoost Classifier

Residuals *= target — prediction*

Output *= average of residuals*

Pred $_i$ *= Pred* $_{(i-1)}$ *+ lr * output*

Tuning parameters

- **objective**: reg:logistic
- **colsample_bytree**: 0.3, 0.7, 1
- **n_estimators**: 50, 100, 150
- **max_depth**: 2, 5, 7, 10
- **eta**: 0.01,0.1,0.5,0.9

Cross validation

- GridSearchCV
- 4 folds

Best parameters

- **objective**: reg:logistic
- **colsample_bytree**: 1
- **n_estimators**: 100
- **max_depth**: 2
- **eta**: 0.1

Best AUC score

- **76.98%**

03

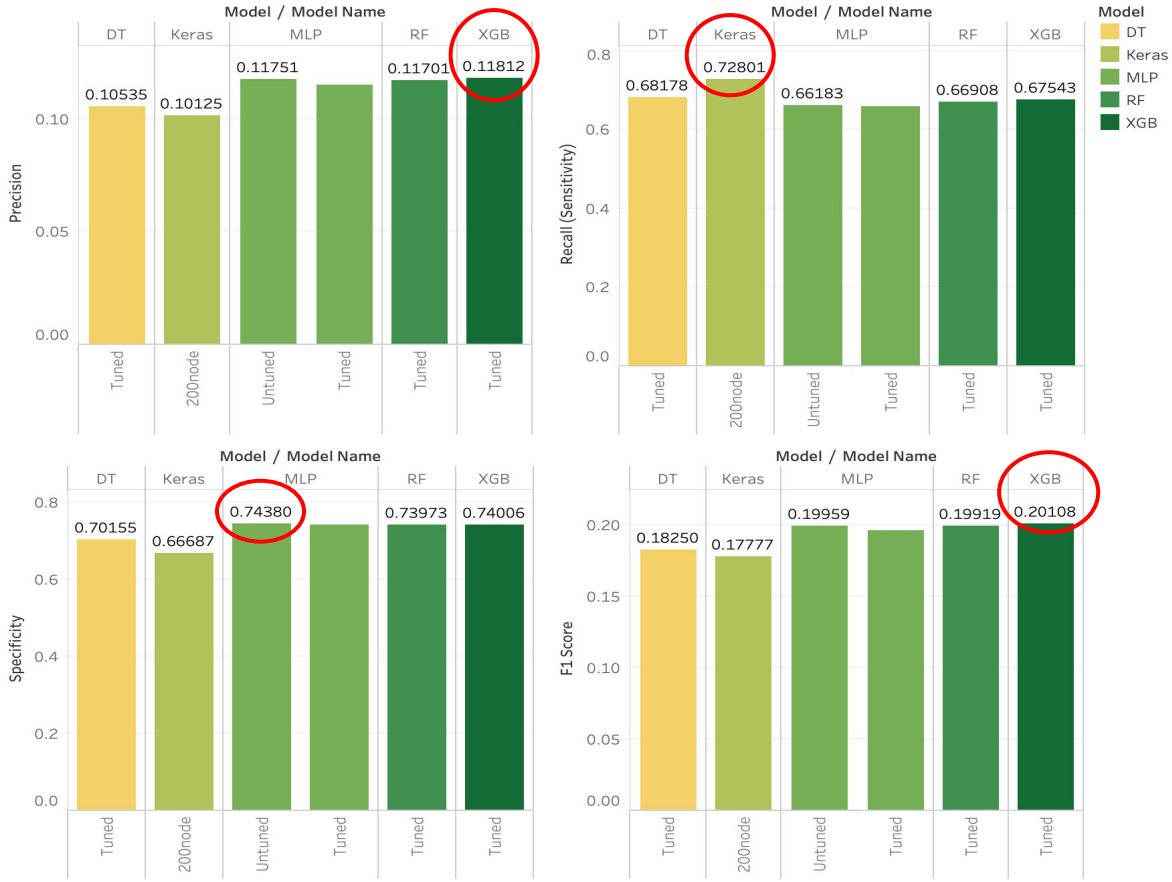
Model Performance

Accuracy / Gains / ROC / Test Set Profit

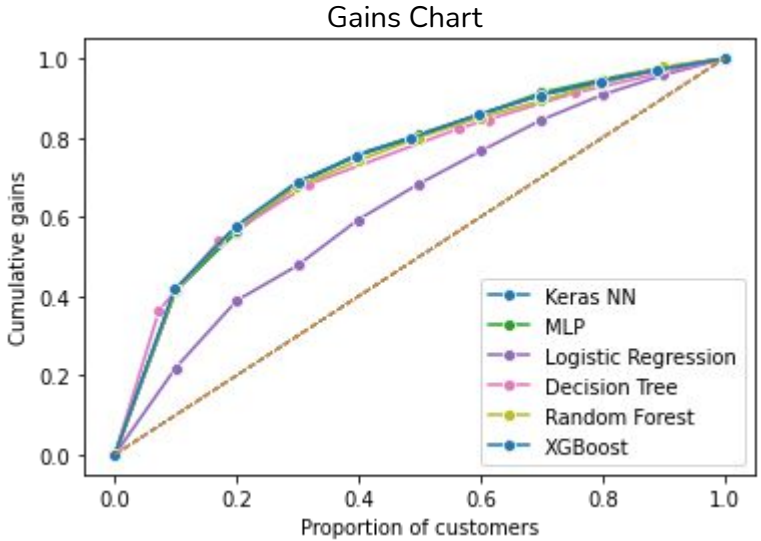
Confusion Matrices for Different Models

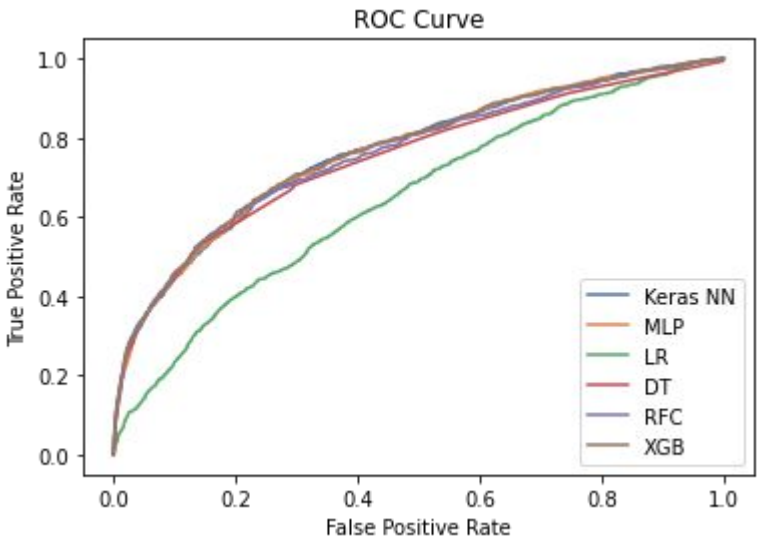
	Keras NN	MLP NN		XGBoost	Random Forest	Decision Tree
	200 Node 4 Layer	Untuned (1,) alpha= 0.01	Tuned (3,) alpha= 0.001	Tuned eta=0.1 max_depth=2 estimators=100	Tuned Max depth=5 estimators=50 bootstrap=True	Tuned Max depth = 6
TN %	63.42	70.73	70.29	70.38	70.35	66.72
TP %	3.57	3.24	3.23	3.31	3.28	3.34
FN %	1.33	1.66	1.68	1.59	1.62	1.56
FP %	31.68	24.36	24.81	24.72	24.75	28.38
ACU%	66.99	73.98	73.52	**73.69**	73.63	70.06

Model Performance Comparison



Gains Chart & ROC Curve





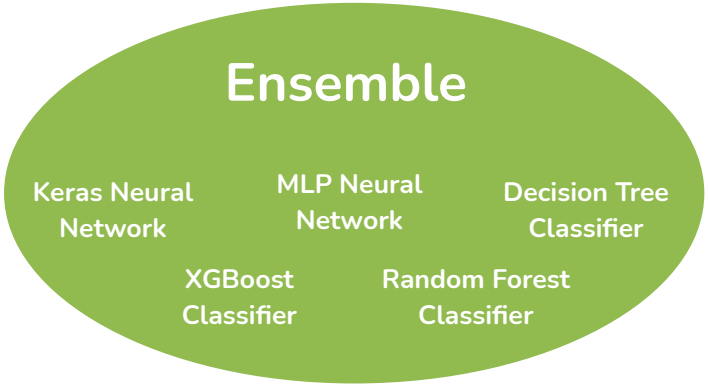
Profit Comparison

	Keras NN	MLP NN		XGBoost	Random Forest	Decision Tree
	200 Node 4 Layer	(1,) alpha=0.01	Tuned (3,) alpha= 0.001	Tuned eta=0.1 max_depth=2 estimators=100	Tuned Max depth=5 estimators=50 bootstrap=True	Tuned Max depth = 6
RR %	10.12%	11.7%	11.7%	11.8%	17.7%	10.5%
ROME %	115.42%	146.17%	148.7%	**151.3%**	149.0%	124.2%
$ Profit (test set)	12.3k	12.5k	12.2k	**12.8k**	12.6k	11.9k

Ensemble vs. XGBoost

XGBoost	**Ensemble Highest Prediction**
73.7%	73.1%
3.31%	3.37%
12.86k	**12.99k**

Lower Accuracy
Higher True Positive Rate
Higher Expected Profit



04

Adopting the Model

Wave 2 Results / Future Projects

73%
Accurate

150%
ROME

$441,000
Dollars in Profit

Thank You