LOGISTIC REGRESSION MODEL - SMOTE RFM

2011 - 2015

logistic regression model - SMOTE RFM training set recall f1-score precision support 0 0.68 0.63 0.65 1883 1 0.65 0.70 0.67 1883 0.66 3766 accuracy macro avg 0.66 0.66 0.66 3766 weighted avg 0.66 0.66 3766 0.66 test set precision recall f1-score 0 0.95 0.63 0.75 4354 1 0.16 0.68 0.26 465 accuracy 0.63 4819 0.55 0.65 0.51 4819 macro avg weighted avg 0.87 0.63 0.71 4819 2013 - 2015

□→ logistic regression model - SMOTE RFM training set precision recall f1-score support 0.72 0.69 0.70 1883 1 0.70 0.71 1883 0.73 accuracy 0.71 3766 0.71 3766 macro avg 0.71 0.71 weighted avg 0.71 0.71 0.71 3766 test set recall f1-score precision support 0.96 0.68 0.79 4354 0.19 0.73 0.31 465 0.68 4819 accuracy 0.58 0.70 0 55 4819 macro avg weighted avg 0.89 0.68 0.75 4819 2014 - 2015

□→ logistic regression model - SMOTE RFM training set precision recall f1-score support 0 0.70 0.67 0.68 1871 0.68 1 0.72 0.70 1871 3742 accuracy 0.69 macro avg 0.69 0.69 0.69 3742 weighted avg 0.69 0.69 0.69 3742 test set precision recall f1-score support 0 0.95 0.67 0.78 4328 1 0.18 0.67 0.28 465 0.67 4793 accuracy 0.56 0.67 0.53 4793 macro avg weighted avg 0.87 0.67 0.74 4793

LOGISTIC REGRESSION MODEL - SMOTE CLV

2011 - 2015

₽	logistic training		ession model	- SMOTE	CLV	
	Ü		precision	recall	f1-score	support
		0	0.69	0.63	0.66	1883
		1	0.66	0.72	0.69	1883
	accur	racy			0.67	3766
	macro	avg	0.67	0.67	0.67	3766
	weighted	avg	0.67	0.67	0.67	3766
	test set					
			precision	recall	f1-score	support
		0	0.95	0.63	0.76	4354
		1	0.16	0.66	0.26	465
	accur	racy			0.63	4819
	macro	avg	0.55	0.65	0.51	4819
	weighted	avg	0.87	0.63	0.71	4819

2013 - 2015

₽	logistic regr training set	ession model	- SMOTE	CLV	
		precision	recall	f1-score	support
	0	0.72	0.70	0.71	1883
	1	0.71	0.72	0.71	1883
	accuracy			0.71	3766
	macro avg	0.71	0.71	0.71	3766
	weighted avg	0.71	0.71	0.71	3766
	test set				
		precision	recall	f1-score	support
	0	0.96	0.69	0.80	4354
	1	0.20	0.73	0.31	465
	accuracy			0.69	4819
	macro avg	0.58	0.71	0.56	4819
	weighted avg	0.89	0.69	0.76	4819

2014 - 2015

+	logistic training	_	ession model	- SMOTE	CLV	
			precision	recall	f1-score	support
		0	0.72	0.67	0.69	1871
		1	0.69	0.74	0.71	1871
	accur	acy			0.70	3742
	macro	avg	0.70	0.70	0.70	3742
	weighted	avg	0.70	0.70	0.70	3742
	test set					
			precision	recall	f1-score	support
		0	0.95	0.66	0.78	4328
		1	0.18	0.68	0.28	465
	accur	acy			0.66	4793
	macro	avg	0.56	0.67	0.53	4793
	weighted	avg	0.88	0.66	0.73	4793

XGBOOST MODEL - SMOTE RFM

2011 - 2015

XGBoost model - SMOTE RFM validation 0-auc:0.582917 Will train until validation 0-auc hasn't improved in 5 rounds. validation 0-auc:0.678763 validation 0-auc:0.659691 validation 0-auc:0.648997 [4] validation 0-auc:0.669309 [5] validation 0-auc:0.677161 validation 0-auc:0.67469 Stopping. Best iteration: validation_0-auc:0.678763 training set precision recall f1-score support 0.76 0 0.83 0.80 1883 0.78 0.85 0.81 1883 3766 0.80 accuracy 0.81 3766 macro avg 0.80 0.80 weighted avg 0.81 0.80 0.80 3766 test set precision recall f1-score support 0.94 0.71 0.81 4354 0.17 0.56 0.26 465 accuracy 0.69 4819 macro avg 0.55 0.63 4819 weighted avg 0.86 0.69 0.75 4819

2013 - 2015

₽	[9] vali [10] vali [11] vali [12] vali [13] vali [14] vali [15] vali [16] vali Stopping. Be	dation_0-auc: dation_0-auc: dation_0-auc: dation_0-auc: dation_0-auc: dation_0-auc: dation_0-auc: st iteration: dation_0-auc:	0.716909 0.718758 0.720424 0.715269 0.71471 0.710705 0.715432 0.717446		
	training set			_	
		precision	recall	f1-score	support
	0	0.94	0.93	0.93	1883
	1	0.93	0.94	0.93	1883
	accuracy macro avg weighted avg	0.93	0.93 0.93	0.93 0.93 0.93	3766 3766 3766
	test set				
		precision	recall	f1-score	support
	0 1	0.93 0.20	0.85 0.35	0.89 0.26	4354 465
	accuracy macro avg weighted avg	0.57	0.60 0.80	0.80 0.57 0.83	4819 4819 4819

2014 - 2015

[14] valid [15] valid [16] valid [17] valid Stopping. Bes [12] valid	lation_0-auc: lation_0-auc: st iteration:	0.703378 0.704289 0.702937		
training set				
	precision	recall	f1-score	support
0	0.94	0.94	0.94	1871
1	0.94	0.94	0.94	1871
accuracy			0.94	3742
macro avg	0.94	0.94		3742
weighted avg		0.94		3742
test set				
	precision	recall	f1-score	support
0	0.92	0.86	0.89	4328
1	0.20	0.32	0.25	465
accuracy			0.81	4793
macro avg	0.56	0.59	0.57	4793
weighted avg			0.83	4793
3				,

XGBOOST MODEL - SMOTE CLV'

2011 - 2015

XGBoost model - SMOTE CLV validation 0-auc:0.641503 Will train until validation 0-auc hasn't improved in 5 rounds. validation 0-auc:0.650185 [2] validation 0-auc:0.659211 validation 0-auc:0.662832 validation 0-auc:0.657371 validation_0-auc:0.661992 validation 0-auc:0.653851 validation 0-auc:0.652762 [8] validation 0-auc:0.649685 Stopping. Best iteration: validation 0-auc:0.662832 training set precision recall f1-score support 0.97 0.92 0.94 1883 0.92 0.97 1883 0.95 accuracy 0.95 3766 macro avg 0.95 0.95 0.95 3766 weighted avg 0.95 0.95 0.95 3766 test set precision recall f1-score support 0.84 0.88 4354 0.92 0.16 0.29 0.21 465 0.79 4819 accuracy 0.54 0.56 macro avg 4819 weighted avg 0.84 0.79 0.81 4819

2013 - 2015

[4] \\ [5] \\ [6] \\ [7] \\ [8] \\	[4] validation_0-auc:0.721801 [5] validation_0-auc:0.721163 [6] validation_0-auc:0.71842 [7] validation_0-auc:0.715367							
		lation_0-auc:	0.715573					
		t iteration:						
[4] \	/alid	lation_0-auc:	0.721801					
training	set							
		precision	recall	f1-score	support			
	0	0.97	0.96	0.96	1883			
	1	0.96	0.90	0.96	1883			
	1	0.90	0.97	0.90	1003			
accur	racy			0.96	3766			
macro	avg	0.96	0.96	0.96	3766			
weighted	avg	0.96	0.96	0.96	3766			
test set								
		precision	recall	f1-score	support			
	0	0.92	0.90	0.91	4354			
	1	0.22	0.27	0.24	465			
accur	-			0.84	4819			
macro	_	0.57	0.58	0.58	4819			
weighted	avg	0.85	0.84	0.85	4819			
					•			

2014 - 2015

[9] validation_0-auc:0.690573 Stopping. Best iteration:

[4] validation_0-auc:0.690764

training set	precision	recall	f1-score	support
0 1	0.96 0.94	0.94 0.96	0.95 0.95	1871 1871
accuracy macro avg weighted avg	0.95 0.95	0.95 0.95	0.95 0.95 0.95	3742 3742 3742
test set	precision	recall	f1-score	support
0 1	0.92 0.18	0.87 0.26	0.89 0.21	4328 465
accuracy macro avg weighted avg	0.55 0.84	0.57 0.81	0.81 0.55 0.83	4793 4793 4793

AFTER PIPELINE & HYPERPARAMETER TUNING - CLV

2011 - 2015

Best AUC Score: 0.7328937313118873

Accuracy: 0.6526250259389915

[[2853 1501] [173 292]] test set

weighted avg

support	f1-score	recall	precision	
4354	0.77	0.66	0.94	0
465	0.26	0.63	0.16	1
4819	0.65			accuracy
4819	0.52	0.64	0.55	macro avg

0.65

4819

0.87

2013 - 2015

Best AUC Score: 0.7722449050026589 Accuracy: 0.6955799958497614

[[3027 1327]

[140 325]] test set				
	precision	recall	f1-score	support
0	0.96	0.70	0.80	4354
1	0.20	0.70	0.31	465
accuracy			0.70	4819
macro avg	0.58	0.70	0.56	4819
weighted avg	0.88	0.70	0.76	4819

2014 - 2015

```
P→ Best AUC Score: 0.7720679007189519
    Accuracy: 0.6951804715209681
    [[3033 1295]
    [ 166 299]]
    test set
                              recall f1-score support
                 precision
                      0.95
                                0.70
                                          0.81
                                                    4328
                      0.19
                                0.64
                                          0.29
                                                    465
                                          0.70
                                                   4793
        accuracy
                      0.57
                                          0.55
                                                   4793
      macro avg
                                0.67
   weighted avg
                      0.87
                                0.70
                                                   4793
```

Summary:

- Logistic regression model accuracy is affected by different time duration. Therefore, this creates
 difficulties in finding exactly appropriated duration in which the model will work best. To conclude,
 this is not the most appropriate model to be used in the case where time duration effects model
 effectiveness in its nature.
- Timeframe should be relevant. Shouldn't be too further away or aged
- When the appropriate timeframe is selected, you will see that no matter the duration is result should be at the same effectiveness level (Recommend Pipeline parameter tuning or XGBoost SMOTE)
- However, in case of selecting XGBoost SMOTE we need to also consider on features used. The
 selected feature should not be affected by duration and should have enough dimensions which
 brings in stability and accuracy in itself. (Recommend RFM rather than CLV)