REPORT JAINAM NIMISH SHROFF MACHINE LEARNING FINAL PROJECT

Semi supervised learning is a branch of machine learning that combines small number of labeled data with large number of unlabeled data during the model training. It comes in between the supervised and unsupervised learning approaches.

1. a self-training algorithm

Self-training algorithm's working is like that of supervised learning, in this the classifier is trained on the labelled data and the trained classifier is used to predict the outcome of the unlabelled data.

STEPS:

- 1. The classifier is trained on the labelled data, the trained classifier is used to predict the outcome of the unlabeled data, the most confident predictions are included in the training data. (Either based on a particular confidence threshold or taking the top n confident results that is also known as k_best results)
- 2. The above step is continued until all the data becomes labelled data.
- 3. The model trained using the final trained data will be the resultant model.

The above experiment is done with different amount of unlabeled data (0%, 10%, 20%, 50%, 90%, 95%) and we analyze the interplay between the amount of unlabeled data and the performance of the resultant model. Also, the same experiment is done on the three datasets.

Experiment Setup

The best model and hyperparameters pair for a particular preprocessed dataset is being used as the base classifier of the self-training classifier. Three datasets are used namely heart disease dataset, marketing dataset, online shoppers' intention dataset. The accuracy, F1 score, ROC curve and runtime is calculated for each experiment.

For heart disease dataset and online shoppers' intention dataset the classifier of choice is

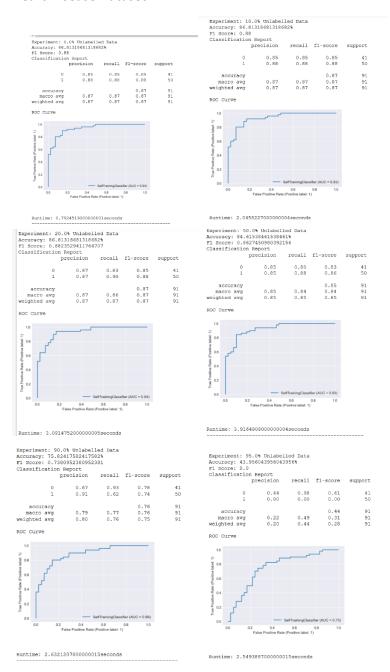
MLP classifier along with hyperparameters as: activation = 'tanh', alpha = 0.0001, hidden_layer_sizes = (50, 100, 50), learning_rate = 'constant', solver = 'sgd',

For Marketing dataset, the classifier of choice is: RandomForestClassifier

With hyperparameters as: max_depth = 15, max_features = 'sqrt', min_samples_leaf = 2, min_samples_split = 2, n_estimators = 100, random_state = 42)

Results Obtained.

Heart Disease Dataset

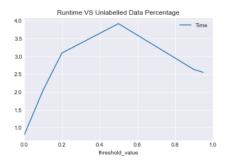


Here we can see the performance evaluation for different experiments done.

Above, the evaluation metrics for different levels of unlabeled data is being shown.

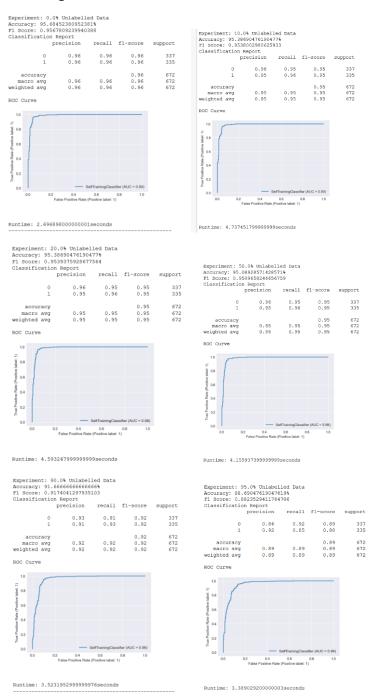




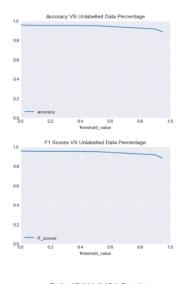


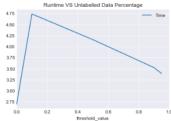
Here we can see the trend of accuracy, f1-score, and runtime with different amount of unlabeled data and can see that the accuracy remains like the supervised experiment until the point where the 90% of data is unlabeled and it drops sharply at 95% amount of unlabeled data.

Marketing Dataset.



Performance evaluation Is shown for different experiments





Here we can see the trend of accuracy, f1-score, and runtime with different amount of unlabeled data and can see that the accuracy remains like the supervised experiment until the point where the 90% of data is unlabeled and it drops a little at 95% amount of unlabeled data.

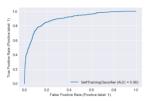
Online Shoppers intention dataset(imbalanced).

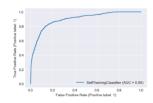


	precision	recall	f1-score	support
0	0.92	0.95	0.93	3114
1	0.66	0.56	0.61	585
accuracy			0.88	3699
macro avg	0.79	0.75	0.77	3699
weighted avg	0.88	0.88	0.88	3699

accur

re		





Experiment: 10.0% Unlabelled Data Accuracy: 88.75371722087051% Fl Score: 0.5961165048543688 Classification Report precision recall

recall fl-score support

Runtime: 20.015071999999996seconds

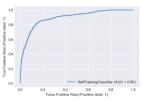
Accuracy: 88				
F1 Score: 0.	577380952380	9524		
Classificati	on Report			
	precision	recall	fl-score	support
0	0.91	0.96	0.93	3114
1	0.69	0.50	0.58	58
accuracy			0.88	3699
macro avg	0.80	0.73	0.76	3699

Runtime: 195.0724231seconds

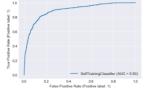
Experiment: 50.0% Unlabelled Dat Accuracy: 89.1051635577183% F1 Score: 0.6013847675568744

	precision	recall	fl-score	support
0	0.91	0.96	0.94	3114
1	0.71	0.52	0.60	585
accuracy	,		0.89	3699
macro avo	0.81	0.74	0.77	3699
weighted avo	0.88	0.89	0.88	3699

ROC Curve

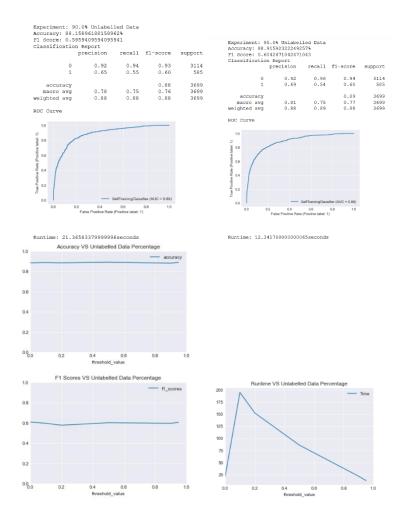


ROC Curve



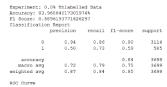
Runtime: 152.55382440000005second

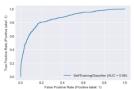
Runtime: 85.98173889999998seconds



Here we can see that the accuracy and the f1 scores are pretty much the same for all the experiments.

Online Shoppers intention dataset (Balanced - Up sampled).





Runtime: 32.77689499999997seconds

Experiment: 20.0% Unlabelled Data Accuracy: 83.48202216815356% F1 Score: 0.583503749147921 Classification Report

	precision	recall	II-score	support
0	0.94	0.85	0.90	3114
1	0.49	0.73	0.58	585
accuracy			0.83	3699
macro avg	0.71	0.79	0.74	3699
weighted avg	0.87	0.83	0.85	3699
ROC Curve				

Runtime: 308.30809080000006second

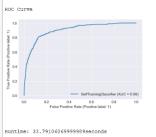
Classificati				
	precision	recall	fl-score	supi
0	0.95	0.85	0.90	
1	0.49	0.75	0.59	
accuracy			0.84	3
macro avg		0.80	0.74	3
weighted avg	0.87	0.84	0.85	3
True Poetive Rate (Postive labet 1)				
Mg 02				
2 02				
0.0	_	SelfTrainingClass	ifier (AUC = 0.87)	
0.0	0.2 0.4 False Positive Rate	0.6 (Positive label: 1	0.8 1.0	
			*	

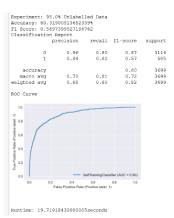
Experiment: 50.0% Unlabelled Data Accuracy: 83.40091916734252% F1 Score: 0.5944517833553502 Classification Report

	precision	recall	fl-score	support
0	0.95 0.48	0.85 0.77	0.90 0.59	3114 585
accuracy macro avg weighted avg	0.72 0.88	0.81	0.83 0.75 0.85	3699 3699 3699
ROC Curve				

1.0				_				
0.8			/					
0.8								
0.6		/						
0.4	1							
	-							
0.2	-1							
					QeHTe	ninina?la	ssifier (AUC	-00
0.0	-			0.4	0	u	0.8	1

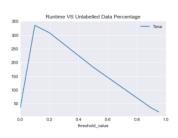
Runtime: 182.98427289999995seconds



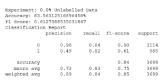


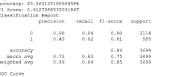


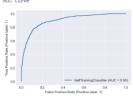




Online Shoppers intention dataset (Balanced – Down sampled)



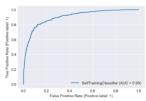




Runtime:	9.	21	.8	82	5	70	0	0	00	0	2	5s	e	c	on	d	S

Experiment: 20.0% Unlabelled Data
Accuracy: 82.698026493646948
F1 Score: 0.602484720968924
Classification Report
precision recall f1-score support

accu	iracy			0.83
macro	avg	0.72	0.83	0.75
weighted	i avg	0.89	0.83	0.84
ROC Curv	re			
1.0				
£				
E 0.8				
9 0.0				
2				



Runtime: 43.97701110000003seconds

Experiment: 1 Accuracy: 83. F1 Score: 0.6 Classificatio	022438496891 075			
	precision	recall	f1-score	support
0	0.96	0.83	0.89	3114
1	0.48	0.83	0.61	585
accuracy			0.83	3699
macro avg	0.72	0.83	0.75	3699
weighted avg	0.89	0.83	0.85	3699

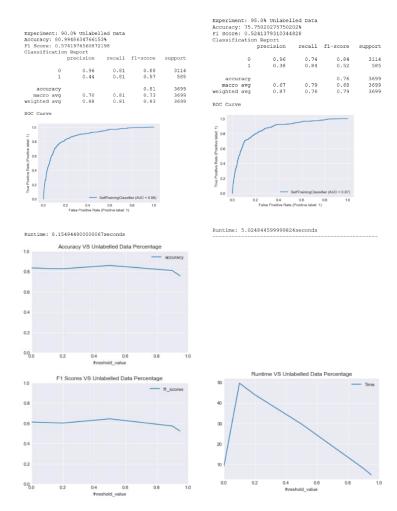
1.0						_
0.8	1					
0.6						
0.8 0.6 0.4						
0.2						
0.0			-	SelfTraining	Classifier (AUC	= 0.90
	0.0	0.2	0.4	0.6 ate (Positive la	0.8	1.0

|--|--|--|--|

Experiment: 5	0.0% Unlabel	led Data		
Accuracy: 85.	996215193295	48%		
Fl Score: 0.6	447187928669	409		
Classificatio	n Report			
	precision	recall	fl-score	support
0	0.96	0.87	0.91	3114
1	0.54	0.80	0.64	585
accuracy			0.86	3699
macro avg	0.75	0.84	0.78	3699
weighted avg	0.89	0.86	0.87	3699

1.0						
Irue Positive Kate (Positive label: 1)	,					
0.6						
0.4						
0.2						
0.0			_	- SelfTrai	ningClassifi	er (AUC = 0.9
	0.0	0.2	0.4	0.6 tate (Positi		.8 1

Runtime: 29.771075699999983seconds



Conclusion:

We can say that even with less amount of labeled data and more amount of unlabeled data the model's performance is remaining similar. The task of labelling is expensive in real world and thus getting higher performances from model even with large amount of unlabeled data is a boon.

2. A Semi Supervised Ensemble Approach

In statistics and machine learning, ensemble methods use multiple learning algorithms to obtain better predictive performance than could be obtained from any of the constituent learning algorithms alone.

STEPS:

- 1. The base classifiers are trained on the unlabeled data using pseudo labelling approach (self-training) and the majority voting is used to make predictions
- 2. The above step is continued until all the data becomes labelled data.
- 3. The model trained using the final trained data will be the resultant model.

The above experiment is done with different amount of unlabeled data (0%, 10%, 20%, 50%, 90%, 95%) and we analyze the interplay between the amount of unlabeled data and the performance of the resultant model. Also, the same experiment is done on the three datasets.

Experiment Setup

The base models are wrapped under self-training classifier module and these base models are used to form an ensemble which is then run on the data with different percentages of unlabeled records.

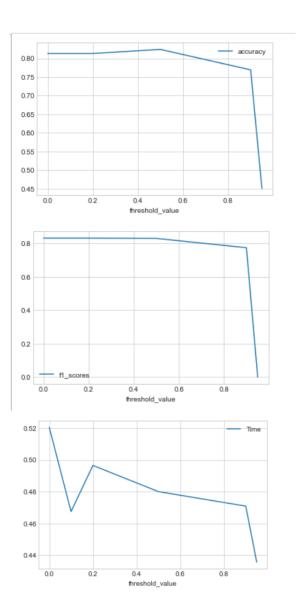
The random forest classifier is used as the based classifier.

Results Obtained Heart Disease Dataset.

		3681318681 5831683168			
Classific	cation H	Report			
	pı	recision	recall	f1-score	suppor
	0	0.80	0.78	0.79	4
	1	0.82	0.84	0.83	5
accu	racy			0.81	9
macro	avg	0.81	0.81	0.81	9
weighted	avg	0.81	0.81	0.81	9
Experimen	nt. 10 (
F1 Score	: 81.318 : 0.831	3681318681 5831683168	31%		
	: 81.318 : 0.8316 cation H	8681318681 5831683168 Report	31% 315	f1-ggoro	gunnar
F1 Score	: 81.318 : 0.8316 cation H	8681318681 5831683168 Report	31% 315	f1-score	suppor
F1 Score	: 81.318 : 0.8316 cation H	8681318681 5831683168 Report recision	31% 315 recall	f1-score	
F1 Score	: 81.318 : 0.8316 cation H	8681318681 5831683168 Report recision 0.80	31% 315 recall 0.78		4
F1 Score Classifi	: 81.318 : 0.8318 cation F	8681318681 5831683168 Report recision 0.80	31% 315 recall 0.78	0.79	4
F1 Score Classific accu:	: 81.318 : 0.8316 cation F	8681318681 8831683168 Report recision 0.80 0.82	31% 315 recall 0.78 0.84	0.79 0.83	4

Experiment: 2 Accuracy: 81. F1 Score: 0.8 Classificatio	318681318681 316831683168	31%		
	precision	recall	f1-score	support
0 1	0.80 0.82	0.78 0.84		41 50
accuracy macro avg weighted avg	0.81			
ROC Curve				
Experiment: 5 Accuracy: 82. F1 Score: 0.8 Classificatio	417582417582 297872340425	41%		
	precision	recall	f1-score	support
0		0.88 0.78	0.82 0.83	41 50
accuracy macro avg weighted avg		0.83 0.82		91 91 91
ROC Curve				

Experiment: 90.	O& Unlabel	led Data		
Accuracy: 76.92				
F1 Score: 0.774				
Classification				
	recision	recall	f1-score	support
0	0.71	0.83		41
1	0.84	0.72	0.77	50
accuracy			0.77	91
macro avg	0.77	0.77	0.77	91
weighted avg	0.78	0.77	0.77	91
Experiment: 95. Accuracy: 45.05 F1 Score: 0.0	54945054945			
Classification				
I	orecision	recall	II-score	support
0	0.45	1.00	0.62	41
1	0.00	0.00	0.00	50
accuracy			0.45	91
macro avq	0.23	0.50	0.31	
weighted avg	0.20	0.45		91
ROC Curve				



Marketing Dataset.

Experiment: 0.0% Unlabelled Data Accuracy: 95.68452380952381% F1 Score: 0.9567809239940388 Classification Report

	precision	recall	f1-score	support
0	0.96	0.96	0.96	337
1	0.96	0.96	0.96	335
accuracy			0.96	672
macro avg	0.96	0.96	0.96	672
weighted avg	0.96	0.96	0.96	672

ROC Curve

Experiment: 10.0% Unlabelled Data Accuracy: 95.08928571428571% F1 Score: 0.9512555391432792

Classificati	on Report			
	precision	recall	f1-score	support
0	0.96	0.94	0.95	337
1	0.94	0.96	0.95	335
accuracy			0.95	672
macro avg	0.95	0.95	0.95	672
weighted avg	0.95	0.95	0.95	672

ROC Curve

Experiment: 90.0% Unlabelled Data Accuracy: 92.261904761904778 F1 Score: 0.9235294117647059 Classification Report

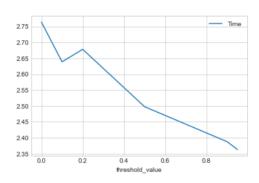
Classificatio	precision	recall	f1-score	support
0	0.94	0.91	0.92	337
1	0.91	0.94	0.92	335
accuracy macro avg weighted avg	0.92 0.92	0.92 0.92	0.92 0.92 0.92	672 672 672

ROC Curve

Experiment: 95.0% Unlabelled Data Accuracy: 87.20238095238095% F1 Score: 0.8693009118541033 Classification Report

Classificat		Report precision	recall	f1-score	support
	0	0.86	0.89	0.87	337
	1	0.89	0.85	0.87	335
accurac	У			0.87	672
macro av	g	0.87	0.87	0.87	672
weighted av	g	0.87	0.87	0.87	672

ROC Curve



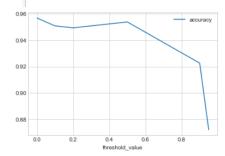
Experiment: 20.0% Unlabelled Data Accuracy: 94.94047619047619% F1 Score: 0.9492537313432836 Classification Report precision recall f1-score support

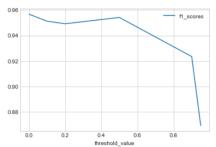
	-			
0	0.95	0.95	0.95	337
1	0.95	0.95	0.95	335
accuracy			0.95	672
macro avg	0.95	0.95	0.95	672
weighted avg	0.95	0.95	0.95	672

ROC Curve

Experiment: 50.0% Unlabelled Data Accuracy: 95.38690476190477% F1 Score: 0.9542097488921715 Classification Report precision recall recall f1-score support

	processin			010010
0	0.96	0.94	0.95	337
1	0.94	0.96	0.95	335
accuracy			0.95	672
macro avg	0.95	0.95	0.95	672
weighted avg	0.95	0.95	0.95	672





Online Shopper Intention - imbalanced

Experiment: 0.0% Unlabelled Data Accuracy: 89.05109489051095% F1 Score: 0.6033300685602352 Classification Report

	precision	recall	f1-score	support
0 1	0.92 0.71	0.96 0.53	0.94 0.60	3114 585
accuracy macro avg weighted avg	0.81 0.88	0.74	0.89 0.77 0.88	3699 3699 3699

ROC Curve

Experiment: 10.0% Unlabelled Data

Accuracy: 89.18626655852934% F1 Score: 0.6146435452793834

assification. p	Report recision	recall	f1-score	support
0	0.92	0.96	0.94	3114
1	0.70	0.55	0.61	585
accuracy			0.89	3699
macro avg	0.81	0.75	0.78	3699
ighted awa	0.88	0.89	0.89	3699

ROC Curve

Experiment: 90.0% Unlabelled Data Accuracy: 89.07812922411462% F1 Score: 0.6085271317829457 Classification Report

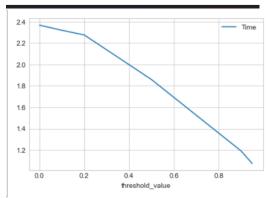
	precision	recall	f1-score	support
0 1	0.92 0.70	0.96 0.54	0.94 0.61	3114 585
accuracy macro avg weighted avg	0.81 0.88	0.75 0.89	0.89 0.77 0.88	3699 3699 3699

ROC Curve

Experiment: 95.0% Unlabelled Data Accuracy: 89.05109489051095% F1 Score: 0.5888324873096447 Classification Report

	precision	recall	f1-score	support
0	0.91	0.96	0.94	3114
1	0.72	0.50	0.59	585
accuracy			0.89	3699
macro avg	0.82	0.73	0.76	3699
weighted avg	0.88	0.89	0.88	3699

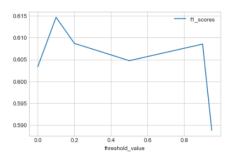
ROC Curve



Experiment: 20.0% Unlabelled Data Accuracy: 89.29440389294405% F1 Score: 0.6086956521739131 Classification Report precision recall f1-score support accuracy macro avg weighted avg 0.89 0.77 0.89 3699 3699 3699 0.88 0.89 Experiment: 50.0% Unlabelled Data Accuracy: 89.18626655852934% F1 Score: 0.6047430830039526 Classification Report recall f1-score precision support 0.91 0.60 0.52 585 0.89 0.77 0.88 3699 accuracy 0.82 macro avg weighted avg 3699 0.89 3699

0.8930		_			accuracy
					,
0.8925					
0.8920					
0.8915					
0.8910	_/_			`	
	/				
0.8905	-/				
	0.0		0.4	0.6	0.8
			hreshold va	lue	

ROC Curve



Online shopper intention - balanced up sampled

F1 Score: 0.644					Accuracy: 86.72614 F1 Score: 0.650533
Classification 1		recall	f1-score	support	Classification Rep
0	0.95	0.89	0.92	3114	0
1			0.64		1
accuracy			0.87	3699	accuracy
macro avg	0.76	0.82	0.78	3699	macro avg
weighted avg	0.89	0.87	0.88	3699	weighted avg
ROC Curve					ROC Curve
Experiment: 10.					
Accuracy: 86.56 F1 Score: 0.644 Classification	3936198972 7462473195 Report	7 % 139	f1-score	support	Experiment: 50.0% Accuracy: 86.64503 F1 Score: 0.648648 Classification Rep
Accuracy: 86.56 F1 Score: 0.644 Classification	3936198972 7462473195 Report recision	7% 139 recall	f1-score	• •	Accuracy: 86.64503 F1 Score: 0.648648 Classification Rep
Accuracy: 86.56 F1 Score: 0.644 Classification p	3936198972 7462473195 Report recision 0.95	7% 139 recall 0.88		3114	Accuracy: 86.64503 F1 Score: 0.648648 Classification Rep
Accuracy: 86.56 F1 Score: 0.644 Classification p	3936198972 7462473195 Report recision 0.95	7% 139 recall 0.88	0.92 0.64	3114	Accuracy: 86.64503 F1 Score: 0.648648 Classification Rep prec
Accuracy: 86.56 F1 Score: 0.644 Classification: p.	3936198972 7462473195 Report recision 0.95 0.55	7% 139 recall 0.88 0.77	0.92 0.64 0.87	3114 585 3699	Accuracy: 86.6450: F1 Score: 0.64864! Classification Rep prec 0 1
Accuracy: 86.56 F1 Score: 0.644 Classification: p 0 1 accuracy	3936198972 7462473195 Report recision 0.95 0.55	7% 139 recall 0.88 0.77	0.92 0.64 0.87 0.78	3114 585 3699 3699	Accuracy: 86.6450: F1 Score: 0.64864! Classification Rep prec 0 1
Accuracy: 86.56 F1 Score: 0.644 Classification in 0 1 accuracy macro avg	3936198972 7462473195 Report recision 0.95 0.55	7% 139 recall 0.88 0.77	0.92 0.64 0.87 0.78	3114 585 3699 3699	Accuracy: 86.6450: F1 Score: 0.64864! Classification Rep pre

Experiment: 90.0% Unlabelled Data Accuracy: 86.26655852933224% F1 Score: 0.6496551724137932 Classification Report

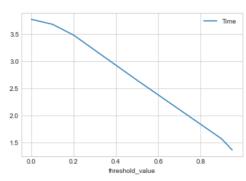
CIASSIII	dulo	precision	recall	f1-score	support
	0	0.96	0.87	0.91	3114
	1	0.54	0.81	0.65	585
accui	cacy			0.86	3699
macro	avg	0.75	0.84	0.78	3699
weighted	avg	0.89	0.86	0.87	3699

ROC Curve

Experiment: 95.0% Unlabelled Data Accuracy: 86.18545552852123% F1 Score: 0.6463667820069203

Classifica	atio	n Report precision	recall	f1-score	support
	0	0.96	0.87	0.91	3114
	1	0.54	0.80	0.65	585
accura	асу			0.86	3699
macro a	_	0.75	0.84	0.78	3699
weighted a	avg	0.89	0.86	0.87	3699

ROC Curve



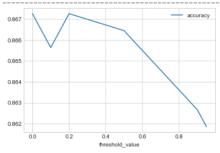
% Unlabelled Data 14220059475% 338078291815

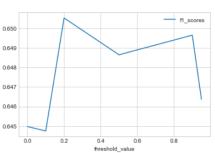
Classificatio	precision	recall	f1-score	support
0	0.96 0.56	0.88	0.92 0.65	3114 585
accuracy	0.50	0.76	0.03	3699
macro avg	0.76	0.83	0.87	3699
weighted avg	0.89	0.87	0.88	3699

Unlabelled Data

03919978372% 486486486488 eport

support	f1-score	recall	precision	
3114	0.92	0.88	0.96	0
585	0.65	0.78	0.56	1
3699	0.87			accuracy
3699	0.78	0.83	0.76	macro avg
3699	0.88	0.87	0.89	weighted avg





Online shopper intention – balanced down sampled

Experiment: 0.0% Unlabelled Data Accuracy: 84.18491484184915% F1 Score: 0.6257197696737044

Classification Report

	precision	recall	f1-score	support
0	0.96	0.84	0.90	3114
1	0.50	0.84	0.63	585
accuracy			0.84	3699
macro avg	0.73	0.84	0.76	3699
weighted avg	0.89	0.84	0.86	3699

Experiment: 10.0% Unlabelled Data Accuracy: 84.13084617464179% F1 Score: 0.6244401791426745 Classification Report

Classifica	atlo:	precision	recall	f1-score	support
	0	0.96	0.84	0.90	3114
	1	0.50	0.83	0.62	585
accura	асу			0.84	3699
macro a	avg	0.73	0.84	0.76	3699
weighted a	avg	0.89	0.84	0.86	3699

ROC Curve

Experiment: 20.0% Unlabelled Data Accuracy: 82.9683698296837% F1 Score: 0.6091811414392059 Classification Report

CIASSILIC	Latit			£1	
		precision	recall	f1-score	support
	0	0.96	0.83	0.89	3114
	1	0.48	0.84	0.61	585
accur	racy			0.83	3699
macro	avg	0.72	0.83	0.75	3699
weighted	avg	0.89	0.83	0.85	3699

ROC Curve

Experiment: 50.0% Unlabelled Data Accuracy: 83.86050283860503% F1 Score: 0.6199872692552515 Classification Report precision recall

	precision	recarr	11-30016	Support
0	0.96	0.84	0.90	3114
1	0.49	0.83	0.62	585
accuracy			0.84	3699
macro avg	0.73	0.84	0.76	3699
weighted avg	0.89	0.84	0.85	3699

ROC Curve

Experiment: 90.0% Unlabelled Data Accuracy: 79.99459313327927% F1 Score: 0.5756880733944955

support	f1-score	recall	Report precision	Classification I
3114	0.87	0.79	0.97	0
585	0.58	0.86	0.43	1
3699	0.80			accuracy
3699	0.72	0.82	0.70	macro avg
3699	0.82	0.80	0.88	weighted avg

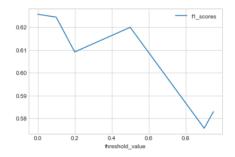
ROC Curve

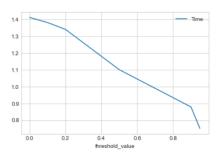
Experiment: 95.0% Unlabelled Data Accuracy: 81.04893214382265% F1 Score: 0.5829863176680548

Classification	n Report precision	recall	f1-score	support
0 1	0.96 0.45	0.81 0.84	0.88 0.58	3114 585
accuracy macro avg weighted avg	0.71 0.88	0.82 0.81	0.81 0.73 0.83	3699 3699 3699

ROC Curve

0.84					accuracy
	\				
0.83					
0.00					
0.82					
0.62					
0.81					\ ,
0.61					
0.80					
0.80	0.0	0.2	0.4	0.6	0.8
			threshold_v		





3. Intrinsically semi supervised learning method

These methods, which we call intrinsically semi-supervised, do not rely on any intermediate steps, or supervised base learners. Usually, they are extensions of existing supervised methods to include unlabeled samples in the objective function.

STEPS:

Pseudo-labelling as a form of margin maximization

when using self-training with supervised SVMs, the decision boundary is iteratively pushed away from the unlabeled samples. Even though the unlabeled data are not explicitly incorporated into the loss function, this amounts to exploiting the low-density assumption, as done in the case of S3VMs.

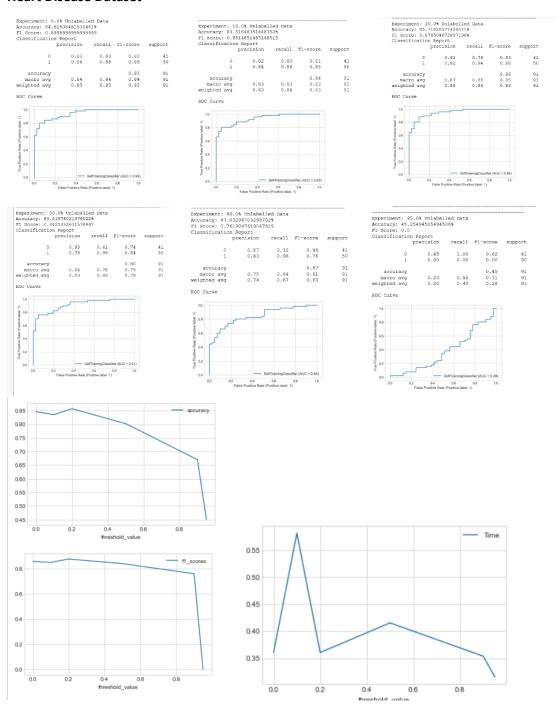
The above experiment is done with different amount of unlabeled data (0%, 10%, 20%, 50%, 90%, 95%) and we analyze the interplay between the amount of unlabeled data and the performance of the resultant model. Also, the same experiment is done on the three datasets.

Experiment Setup

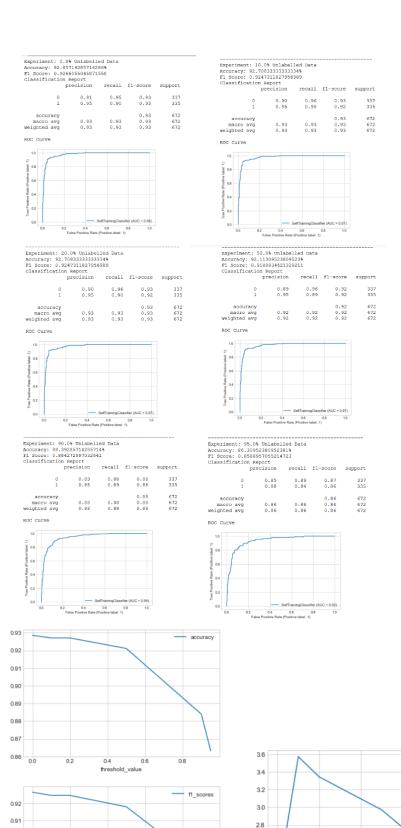
The SVM classifier is used as the base classifier.

Results Obtained

Heart Disease Dataset



Marketing Dataset



2.6

2.4

22

0.0

0.2

0.6

fhreshold_value

0.8

0.90

0.89

0.88

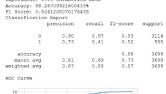
0.87

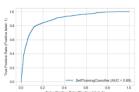
0.0

fhreshold_value

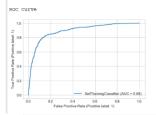
— Time

Online Shopper Intention – imbalanced

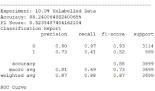


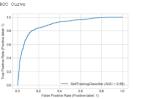


Accuracy: 88. F1 Score: 0.5	158961881589	62%		
Classificatio		1716		
	precision	recall	f1-score	support
0	0.90	0.97	0.93	3114
1	0.73	0.40	0.51	58
accuracy			0.88	369
macro avg	0.81	0.68	0.72	369



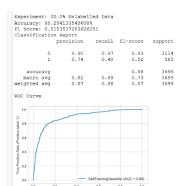
0.886





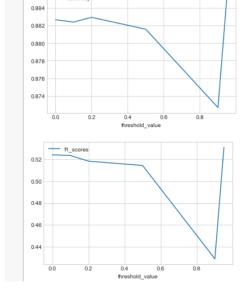
PyDetruet	IL. S	O.OS UNLADEL	Ted Data		
Accuracy:	87.	266828872668	28%		
F1 Score:	0.4	290909090909	091		
Classific	catio	n Report			
		precision	recall	f1-score	support
	0	0.88	0.98	0.93	3114
	1	0.74	0.30	0.43	585
accur	асу			0.87	3699
macro	avg	0.81	0.64	0.68	3699
weighted	avg	0.86	0.87	0.85	3699

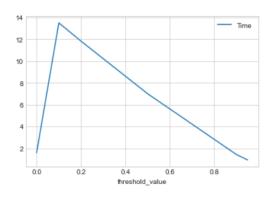
1.0				_	_
£					
True Positive Rate (Positive label: 1)					
2					
0.6					
8	/				
0.4					
ě					
0.2					
2					
0.0		_	SelfTraining	Classifier (AUC	= 0.88)



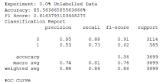
Experiment: 95.0	0% Unlabel	led Data		
Accuracy: 88.53	7442552041	09%		
F1 Score: 0.530	9734513274	337		
Classification	Report			
p	recision	recal1	f1-score	suppor
0	0.90	0.97	0.93	311
1	0.75	0.41	0.53	58
accuracy			0.89	369
macro avg	0.83	0.69	0.73	369
weighted avg	0.87	0.89	0.87	369

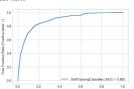
1.0					_	_
		 _				
0.8						
	1					
0.6						
04						
0.4						
0.2						
0.0			SelfTra	siningClass	ifier (AUC =	0.89)



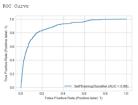


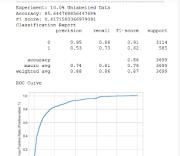
Online shopper intention – balanced up sampled





Experiment: 50. Accuracy: 86.45 F1 Score: 0.627	5798864557 5092936802	99%		
Classification 1	Report			
P	recision	recall	fl-score	support
0	0.94	0.89	0.92	3114
1	0.56	0.72	0.63	585
accuracy			0.86	3699
macro avq	0.75	0.81	0.77	3699
weighted avg	0.88	0.86	0.87	3699



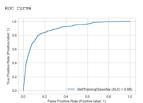


Experiment: 20.0% Unlabelled Data Accuracy: 85.77994052446607% Fl Score: 0.6193921852387844 Classification Report precision recall

ROC Curve

recall fi-score support

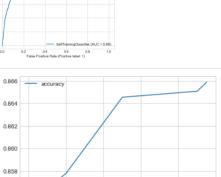




0.2

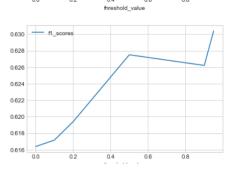
0.856

0.0

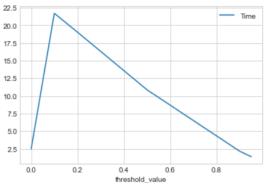


0 1	0.94	0.89 0.72	0.92 0.63	311 58
accuracy macro avg weighted avg	0.75 0.88	0.81 0.87	0.87 0.77 0.87	369 369 369
ROC Curve				
10 08 08 (1) 08 08 00 00 00 00 00 00 00 00 00 00 00				
02 02	Se	#TrainingClassifier	AUC = 0.891	
0.0 0.2	0.4 False Positive Rate (F	0.6 0.8	1.0	

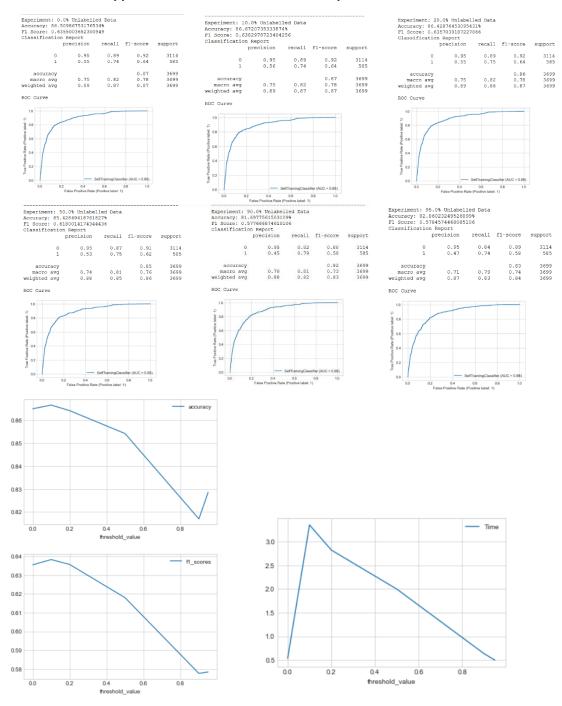
Experiment: 95.0% Unlabelled Data
Accuracy: 86.59007053257638
F1 Soore: 0.5004023845007451
Classification Report
precision recall f1-score support



0.8



Online shopper intention – balanced down sampled



Interplay between class imbalance and SSL.

When the class is balanced using up sampling/down sampling it is observed that there is a gain in F1 score overall all the experiments (with different amounts of unlabeled data) and since the classes are imbalanced the more precise measure of the performance of the model will be f1 score and not the accuracy.

Best SSL algorithm

The overall best performance is obtained using the Self Training approach, considering both the accuracy and F1 score. although we are including our 'pseudo-labeled' data with labeled training data, some of the 'pseudo-labeled' data is certainly going to be incorrect. When enough of the 'pseudo-labels' are incorrect, the self-training algorithm can reinforce poor classification decisions, and classifier performance can actually get worse.