



Stroke Prediction Dataset

Classification Project



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project goal

Prediction Stroke



what is a stroke?

Stroke is a medical emergency. A stroke occurs when blood flow to a part of your brain is interrupted or reduced, preventing brain tissue from getting oxygen and nutrients. Brain cells begin to die within minutes

Data content

Feature:

1. id
2. gender
3. age
4. hypertension
5. heart_disease
6. ever_married
7. work_type
8. Residence_type
9. avg_glucose_level
10. bmi
11. smoking_status

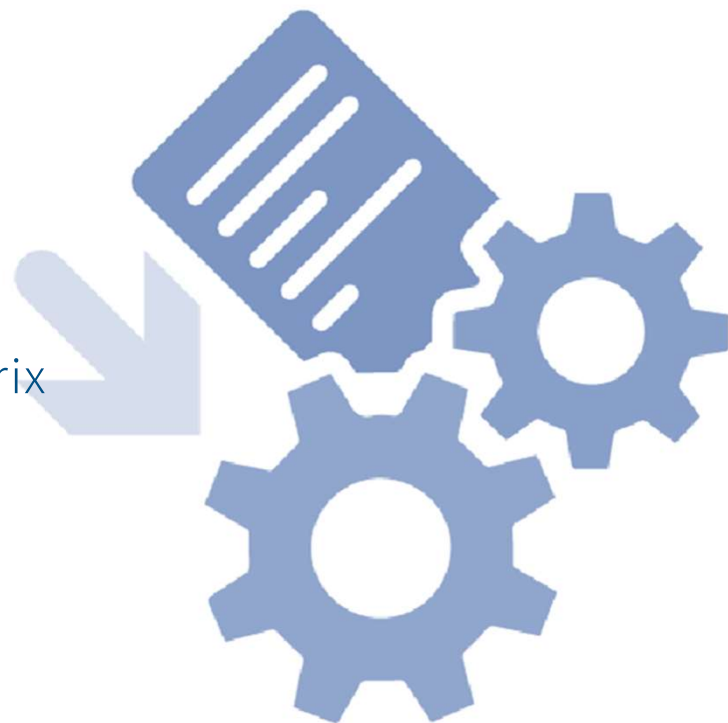
 **12 columns**
5110 row

Target variable – stroke



Exploratory data analysis -EDA

- Missing Values Handling
- Remove duplicate and outlier
- Feature selection
- Compute pairwise correlation of columns – matrix
- Encoding
- Understanding the variables



EDA

1. ID

Feature selection

- 1.Unique value
- 2.Drop

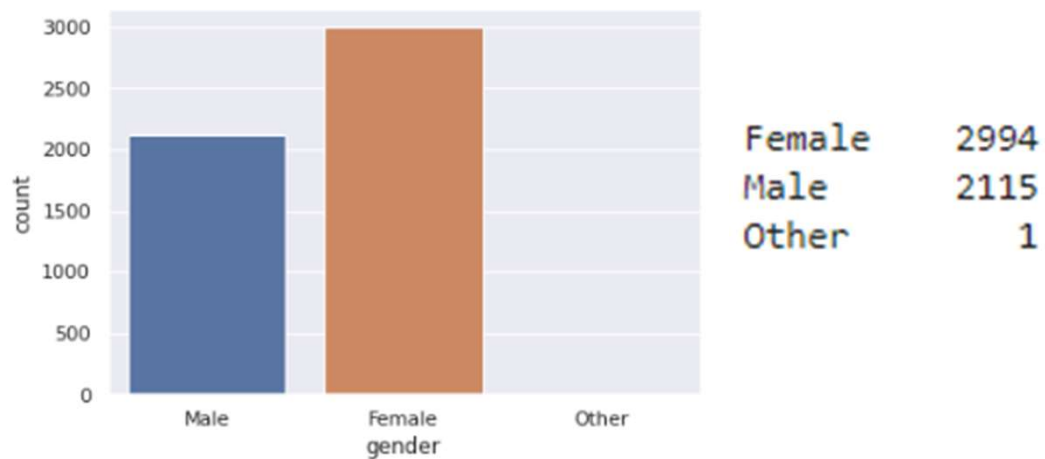
11 columns, 5110 row



EDA

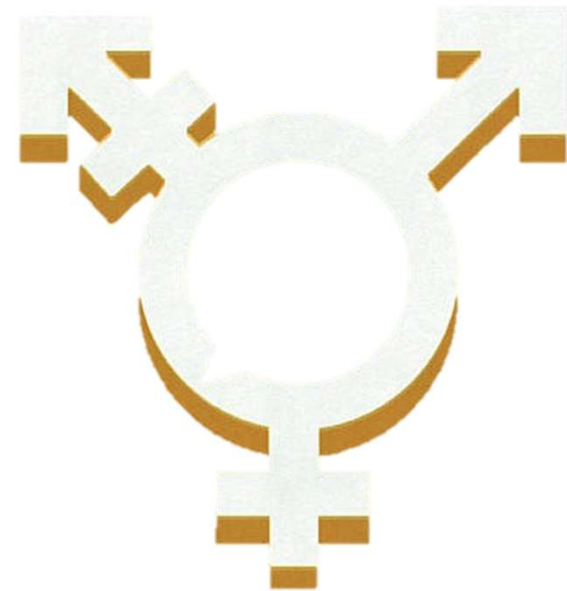
2. Gender

"Male", "Female" or "Other"



		srtoke		
		0	1	
gender	Female	2853	141	4.71%
	Male	2007	108	5.11%
	Other	1	0	0.00%

11 columns, 5109 row

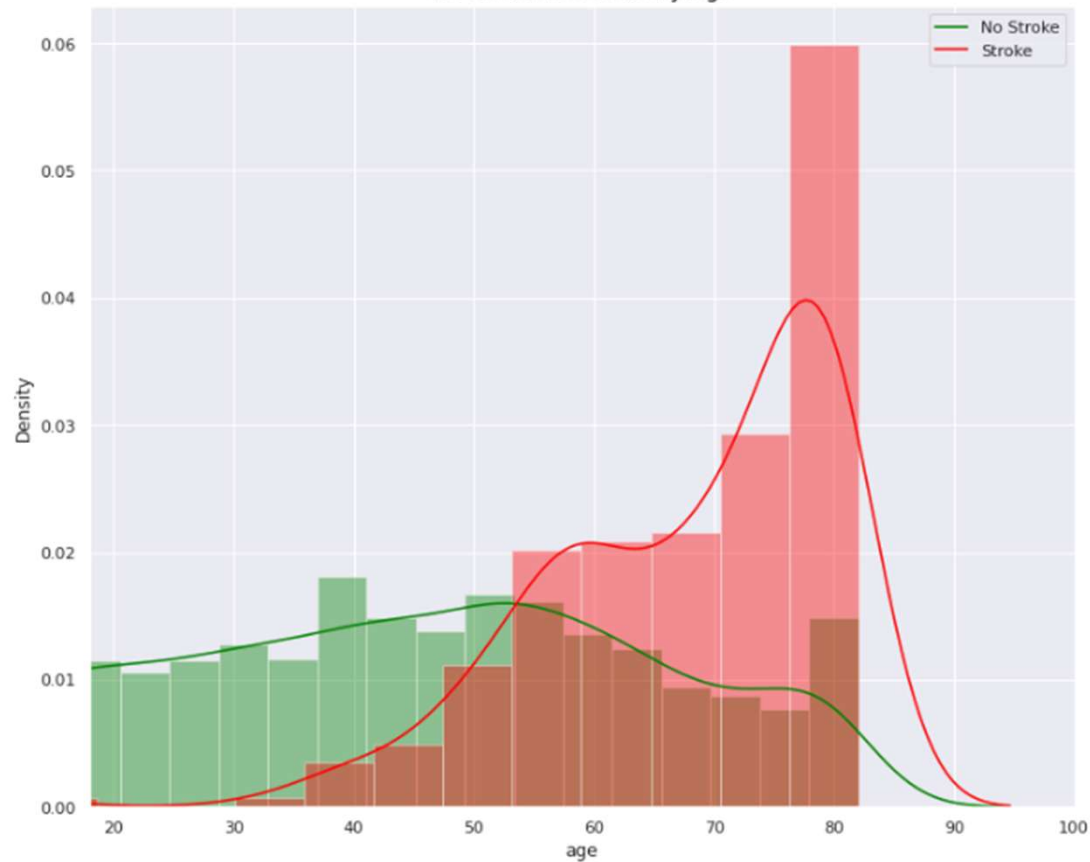


EDA

3. Age:

Age of the patient

No Stroke vs Stroke by Age



11 columns, 5109 row

age	
count	5110.000000
mean	43.226614
std	22.612647
min	0.080000
25%	25.000000
50%	45.000000
75%	61.000000
max	82.000000



EDA

4. Hypertension :

0 - No hypertension

1 - Hypertension

		srtoke		
		0	1	
hypertension	0	4429	183	3.97%
	1	432	66	13.25%

11 columns, 5109 row



EDA

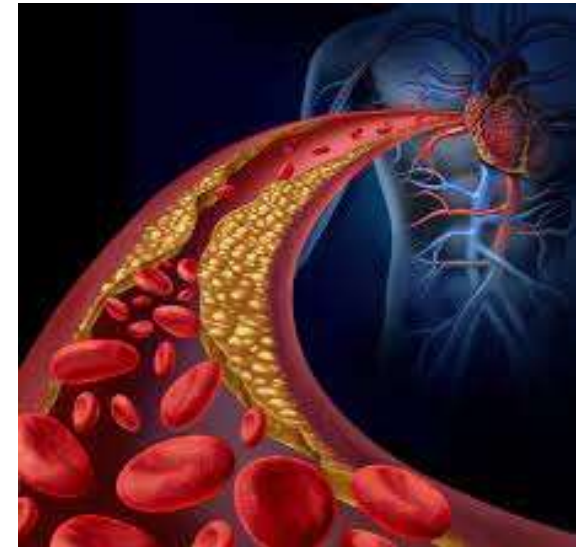
5. Heart disease :

0 - No heart diseases

1 - Heart disease

		srtoke		
		0	1	
heart_disease	0	4632	202	4.18%
	1	229	47	17.03%

11 columns, 5109 row



EDA

6. Ever married :

"No" or "Yes"

		srtoke		0	1	
ever_married	No	1728	29	1.65%		
	Yes	3133	220	6.56%		

11 columns, 5109 row



EDA

7. Work type :

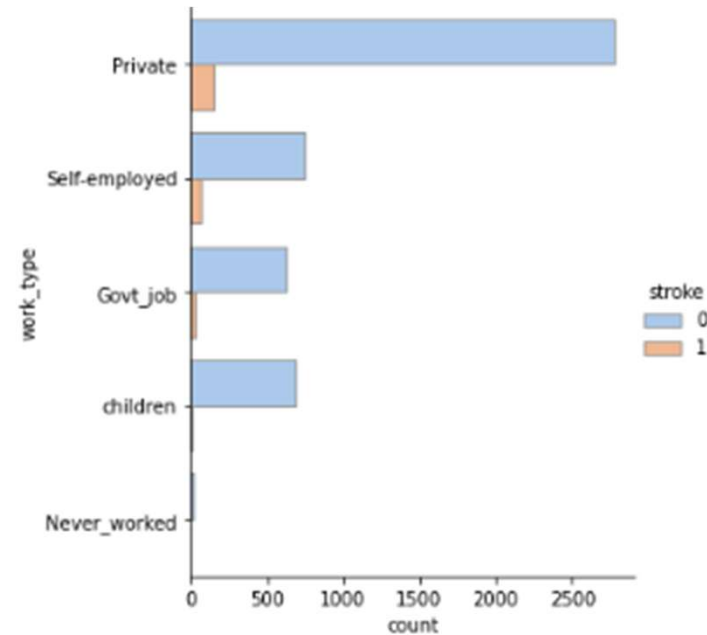
"Private"

"Self-employed"

"Govt_jov"

"Children"

"Never_worked"



srtoke		0	1	
work_type	Govt_jov	624	33	5.02%
	Never_worked	22	0	0.00%
	Private	2776	149	5.09%
	Self-employed	754	65	7.94%
	children	685	2	0.29%

11 columns, 5109 row



EDA

9. Residence type : "Rural" or "Urban"

srtoke		0	1	
Residence_type	Rural	2400	114	4.53%
	Urban	2461	135	5.20%

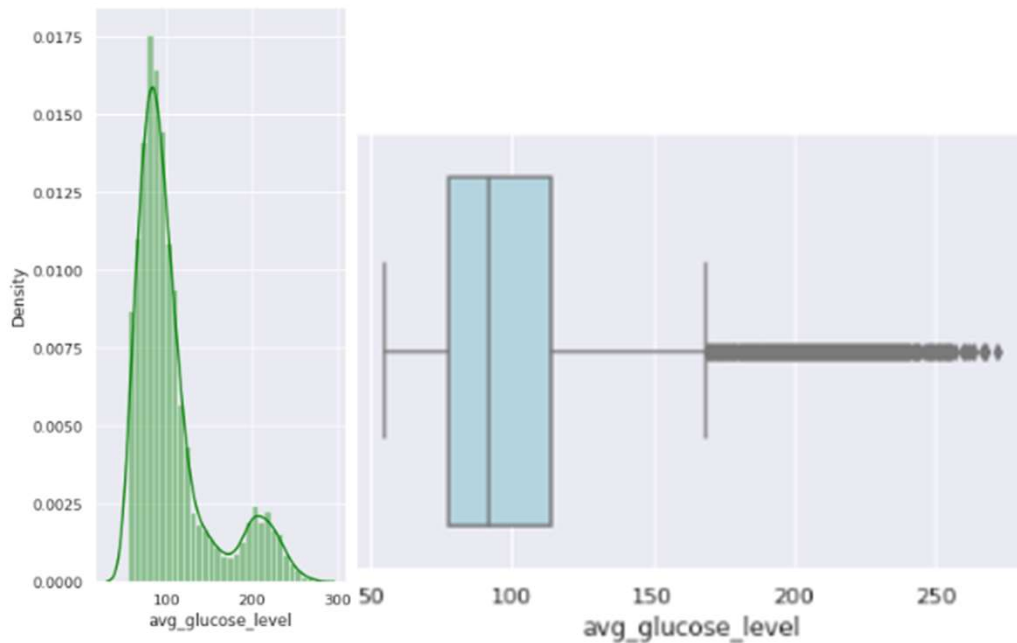
11 columns, 5109 row



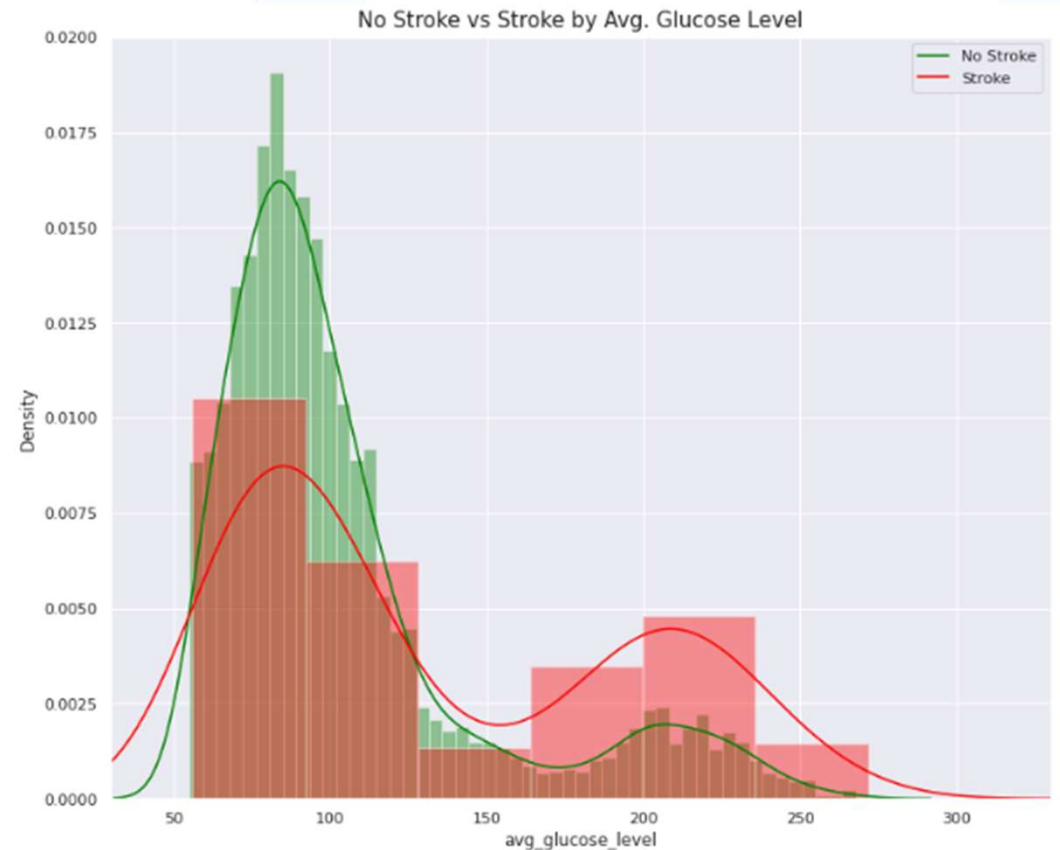
EDA

9. Avg Glucose level:

Average glucose level in blood



11 columns, 5109 row

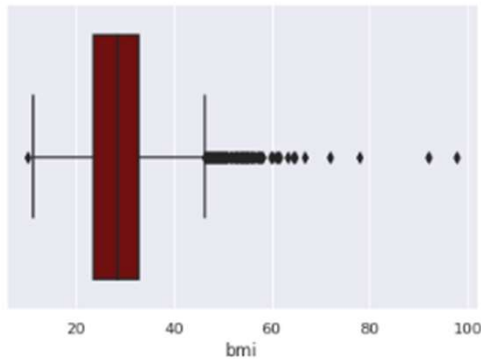


EDA

10. BMI:

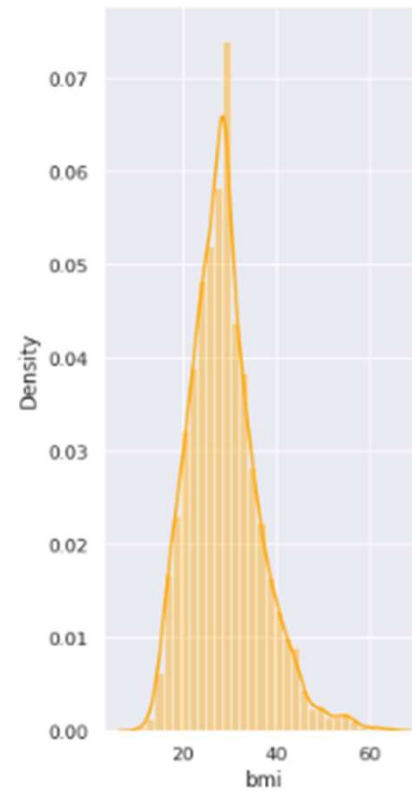
Body Mass Index

201 null - Replacing the missing values with mean



Outlier:

Decision - Drop 5 max bmi



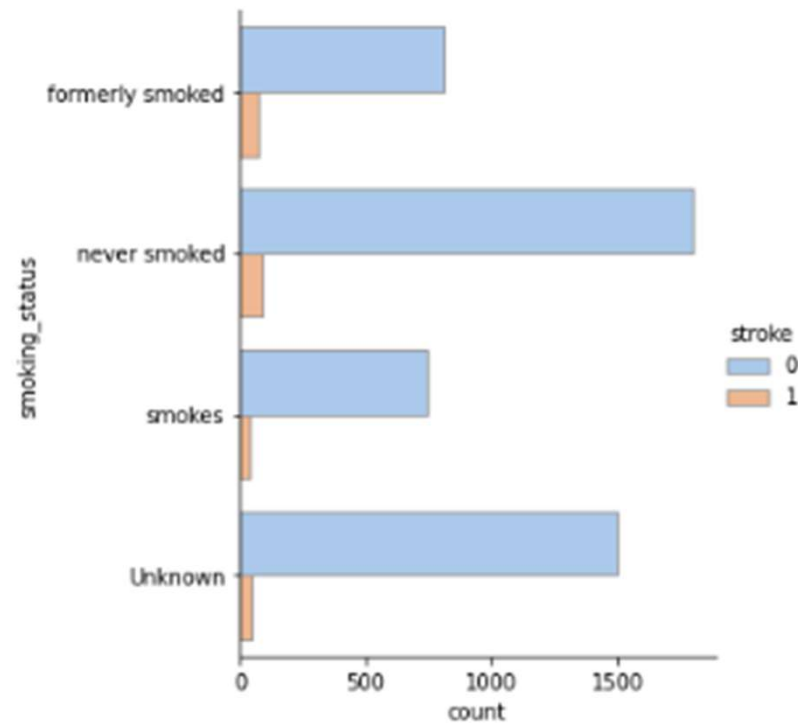
11 columns, 5104 row



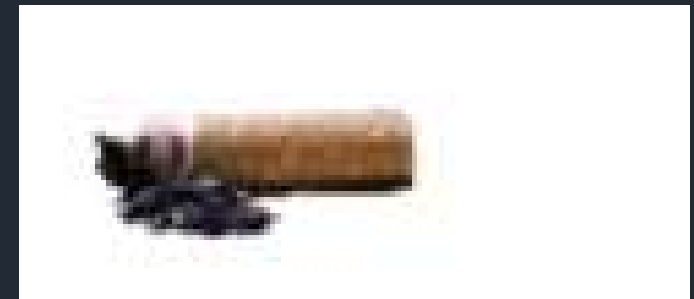
EDA

10. Smoking Status:

"Formerly smoked"
"Never smoked"
"Smokes"
"Unknown"



		stroke		
		0	1	
smoking_status	Unknown	1497	47	3.04%
	formerly smoked	815	70	7.91%
	never smoked	1802	90	4.76%
	smokes	747	42	5.32%

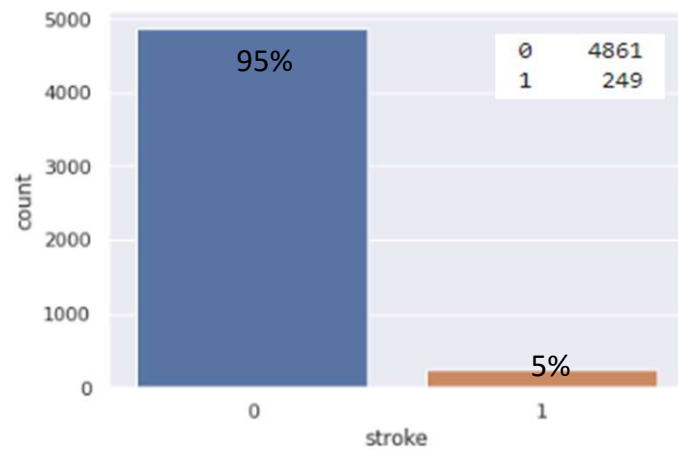


Target variable

12. Stroke:

0 = No stroke

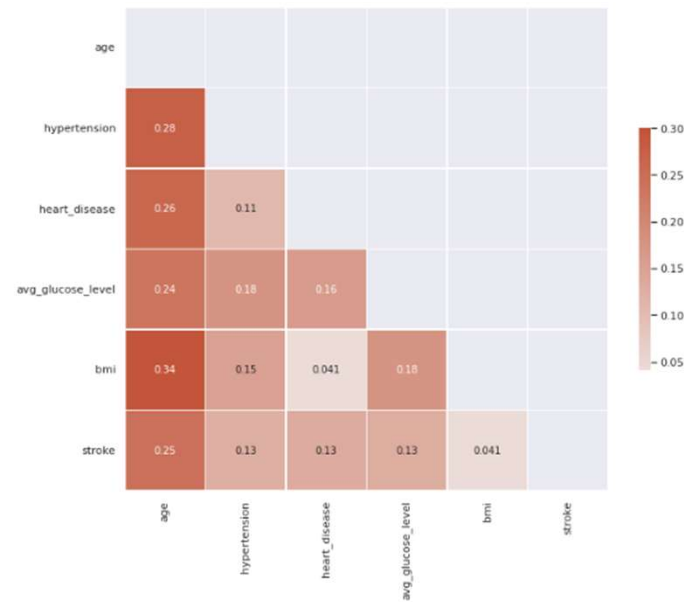
1 = Stroke



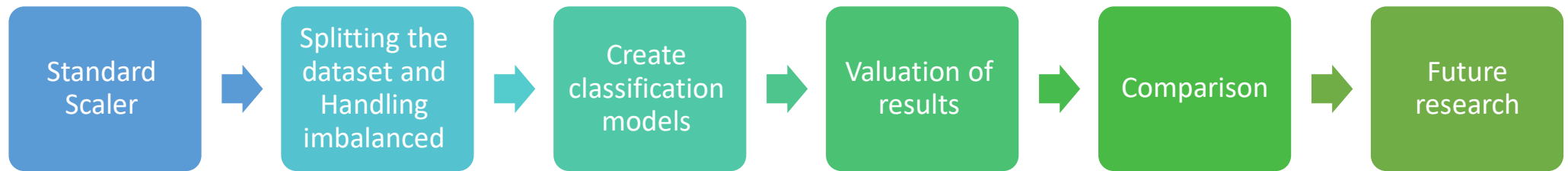
Conclusion and the next steps

Age has the highest impact on stroke, even though the stroke also depends on the other variables,

Such as: glucose level, heart disease, blood pressure, smoking and even type of work and area of residence

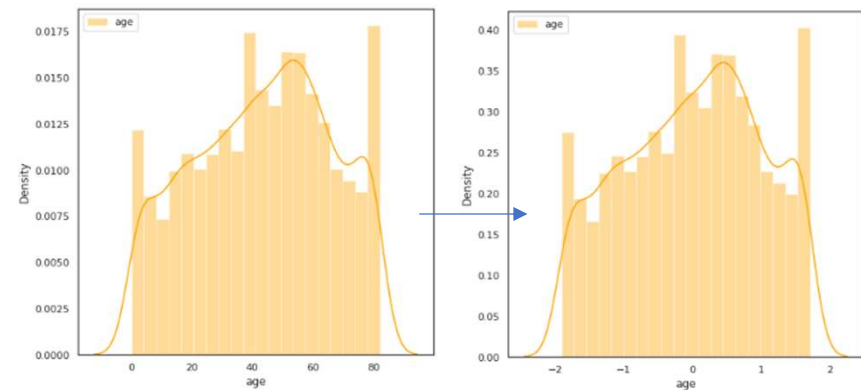
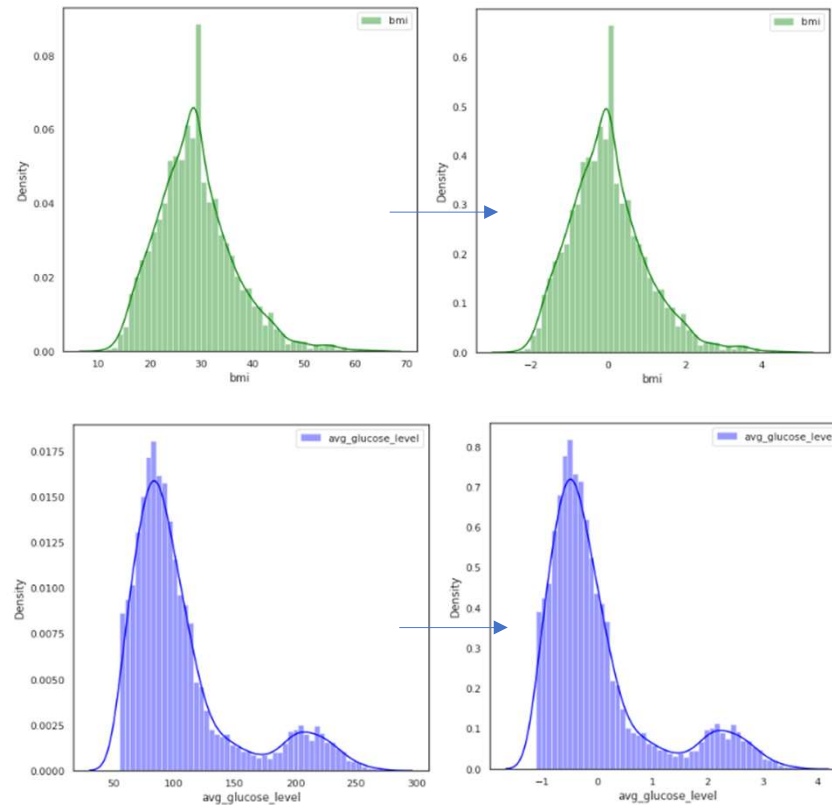


Conclusion and the next steps



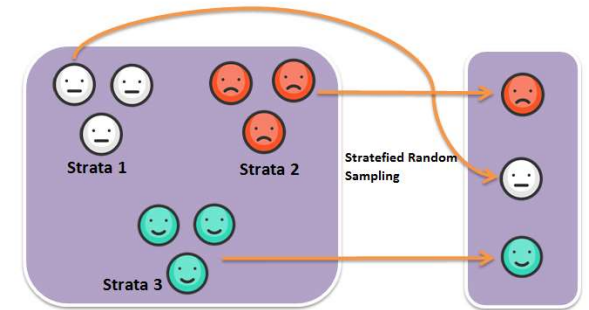
Standard Scaler

➤ The models With and Without Standard Scaler



Splitting data and Handling imbalanced

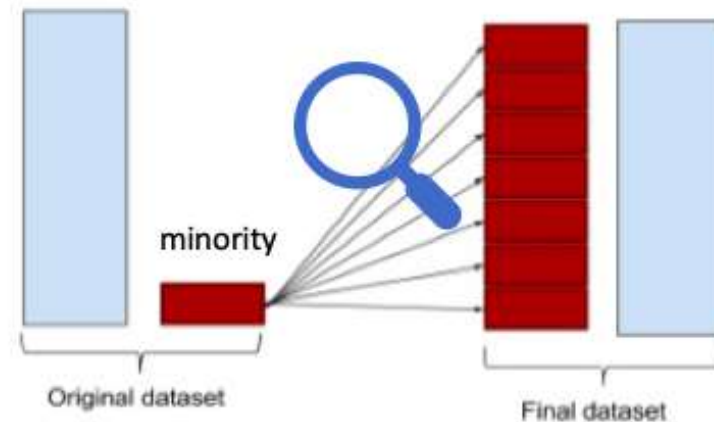
- Splitting dataset 30-70
- stratify (95-5)
- Over Sampling using SMOTE (50-50 train data)



Before OverSampling, counts of label '1': 174
Before OverSampling, counts of label '0': 3398

After OverSampling, the shape of train_X: (6796, 17)
After OverSampling, the shape of train_y: (6796,)

After OverSampling, counts of label '1': 3398
After OverSampling, counts of label '0': 3398



classification models

- Logistic Regression
- SVM
- Random Forest



valuation of results

➤ Precision

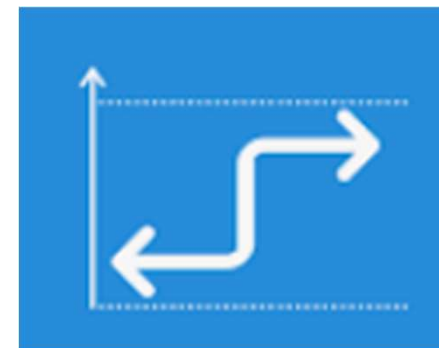
➤ Recall

➤ F1

and connection



Logistic Regression



➤ With and without Over Sampling

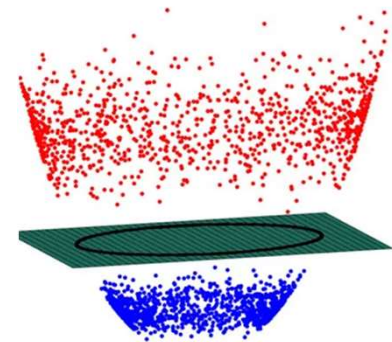
Testing Score 0.9510443864229765				
	precision	recall	f1-score	support
0	0.95	1.00	0.97	1457
1	0.00	0.00	0.00	75
accuracy			0.95	1532
macro avg	0.48	0.50	0.49	1532
weighted avg	0.90	0.95	0.93	1532

	0	1
0	1457	0
1	75	0

Testing Score 0.7271540469973891				
	precision	recall	f1-score	support
0	0.99	0.72	0.83	1457
1	0.13	0.79	0.22	75
accuracy			0.73	1532
macro avg	0.56	0.76	0.53	1532
weighted avg	0.94	0.73	0.80	1532

	0	1
0	1055	402
1	16	59

SVM



- With and without Over Sampling
- With and without Standard Scaler

➤ Cross Validation

Scores : 0.802 0.781 0.790 0.776 0.797 0.780 0.795

➤ Grid Search

Best parameters set found on development set:

`{'C': 1000, 'gamma': 0.001, 'kernel': 'rbf'}`

	precision	recall	f1-score	support
0	0.98	0.76	0.85	1457
1	0.13	0.68	0.21	75
accuracy			0.75	1532
macro avg	0.55	0.72	0.53	1532
weighted avg	0.94	0.75	0.82	1532

Training Score 0.796203649205415

Testing Score 0.7088772845953003

	0	1
0	1026	431
1	15	60

Random Forest

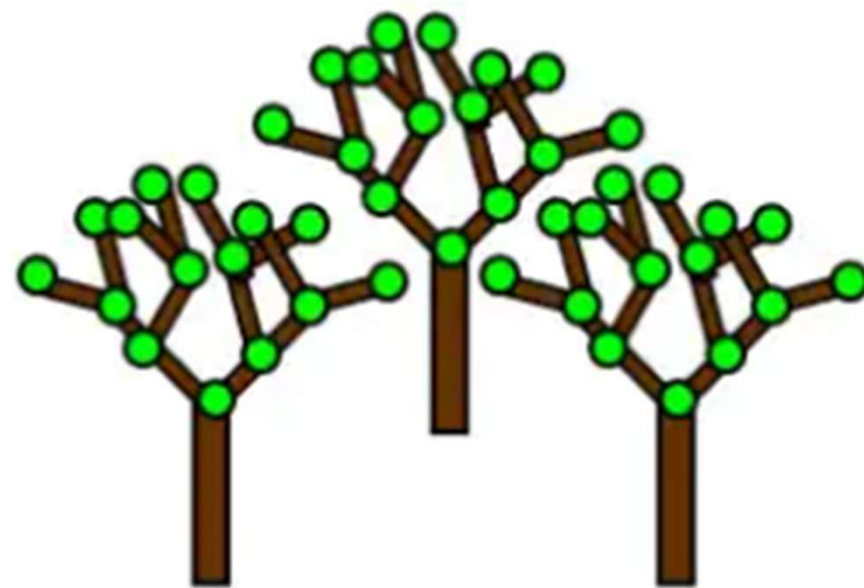
➤ With and without Over Sampling

➤ Grid Search

Best Parameters : {'criterion': 'gini', 'n_estimators': 150, 'random_state': 0}

	precision	recall	f1-score	support
0	0.95	0.99	0.97	1457
1	0.09	0.01	0.02	75
accuracy			0.95	1532
macro avg	0.52	0.50	0.50	1532
weighted avg	0.91	0.95	0.93	1532

	0	1
0	1447	10
1	74	1



Present main results and comparison of methods

Grid Search

scoring = 'accuracy'.

```
LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
                    intercept_scaling=1, l1_ratio=None, max_iter=100,
                    multi_class='auto', n_jobs=None, penalty='l2',
                    random_state=None, solver='lbfgs', tol=0.0001, verbose=0,
                    warm_start=False):
```

Best Accuracy : 77.27%

Best Parameters : {'C': 0.5, 'random_state': 0}

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False):
```

Best Accuracy : 78.13%

Best Parameters : {'C': 0.5, 'kernel': 'linear', 'random_state': 0}

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None,
                       criterion='gini', max_depth=None, max_features='auto',
                       max_leaf_nodes=None, max_samples=None,
                       min_impurity_decrease=0.0, min_impurity_split=None,
                       min_samples_leaf=1, min_samples_split=2,
                       min_weight_fraction_leaf=0.0, n_estimators=100,
                       n_jobs=None, oob_score=False, random_state=None,
                       verbose=0, warm_start=False):
```

Best Accuracy : 97.29%

Best Parameters : {'criterion': 'gini', 'n_estimators': 100, 'random_state': 0}

scoring = 'recall'

```
LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
                    intercept_scaling=1, l1_ratio=None, max_iter=100,
                    multi_class='auto', n_jobs=None, penalty='l2',
                    random_state=None, solver='lbfgs', tol=0.0001, verbose=0,
                    warm_start=False):
```

Best Accuracy : 81.64%

Best Parameters : {'C': 0.5, 'random_state': 0}

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False):
```

Best Accuracy : 85.17%

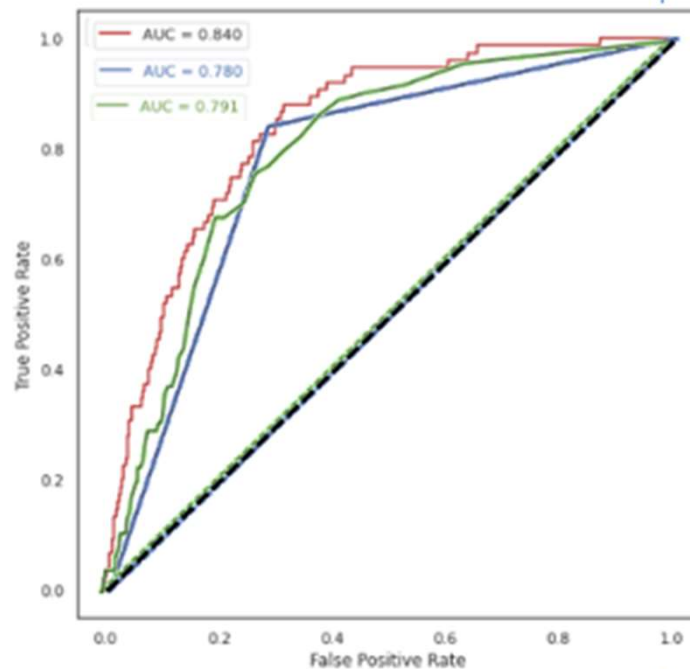
Best Parameters : {'C': 1, 'kernel': 'linear', 'random_state': 0}

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None,
                       criterion='gini', max_depth=None, max_features='auto',
                       max_leaf_nodes=None, max_samples=None,
                       min_impurity_decrease=0.0, min_impurity_split=None,
                       min_samples_leaf=1, min_samples_split=2,
                       min_weight_fraction_leaf=0.0, n_estimators=100,
                       n_jobs=None, oob_score=False, random_state=None,
                       verbose=0, warm_start=False):
```

Best Accuracy : 95.29%

Best Parameters : {'criterion': 'gini', 'n_estimators': 150, 'random_state': 0}

Present main results and comparison of methods



scoring = 'recall'

Logistic Regression

p	1073	384		precision	recall	f1-score	support
N	14	61	0	0.99	0.74	0.84	1457
	P	N	1	0.14	0.81	0.23	75
						0.74	1532
				0.56	0.77	0.54	1532
				0.95	0.74	0.81	1532

SVM

P	1050	407		precision	recall	f1-score	support
N	12	63	0	0.99	0.72	0.83	1457
	P	N	1	0.13	0.84	0.23	75
						0.73	1532
				0.56	0.78	0.53	1532
				0.95	0.73	0.80	1532

random forest

	P	1448	9		precision	recall	f1-score	support
N		74	1	0	0.95	0.99	0.97	1457
	P			N	1	0.10	0.01	0.02
								75
							0.95	1532
					0.53	0.50	0.50	1532
					0.91	0.95	0.93	1532

summary and Future research

- Scaling VS Threshold
- Defining additional / other variables in Grid Search
- Examination of additional models
- Combine scores and determining weight for each of them
- ensemble methods



Thanks

Tomer badug
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Judi Eliya

