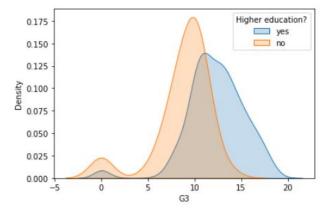
# WUM Projekt 1

Mikołaj Spytek Artur Żółkowski

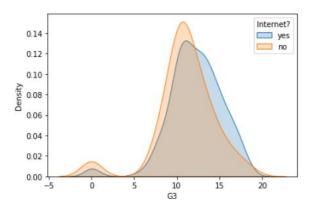
### RangeIndex: 649 entries, 0 to 648 Data columns (total 33 columns):

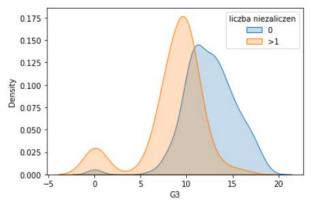
#	Column	Non	-Null Count	Dtype
0	school	649	non-null	objec
1	sex	649	non-null	objec
2	age	649	non-null	int64
3	address	649	non-null	objec
4	famsize	649	non-null	objec
5	Pstatus	649	non-null	objec
6	Medu	649	non-null	int64
7	Fedu	649	non-null	int64
8	Mjob	649	non-null	objec
9	Fjob	649	non-null	objec
10	reason	649	non-null	objec
11	guardian	649	non-null	objec
12	traveltime	649	non-null	int64
13	studytime	649	non-null	int64
14	failures	649	non-null	int64
15	schoolsup	649	non-null	objec
16	famsup	649	non-null	objec
17	paid	649	non-null	objec
18	activities	649	non-null	objec
19	nursery	649	non-null	objec
20	higher	649	non-null	objec
21	internet	649	non-null	objec
22	romantic	649	non-null	objec
23	famrel	649	non-null	int64
24	freetime	649	non-null	int64
25	goout	649	non-null	int64
26	Dalc	649	non-null	int64
27	Walc	649	non-null	int64
28	health	649	non-null	int64
29	absences	649	non-null	int64
30	G1	649	non-null	int64
31	G2	649	non-null	int64
32	G3	649	non-null	int64
1 23				_

### Zbiór danych



### Rozkłady wybranych zmiennych

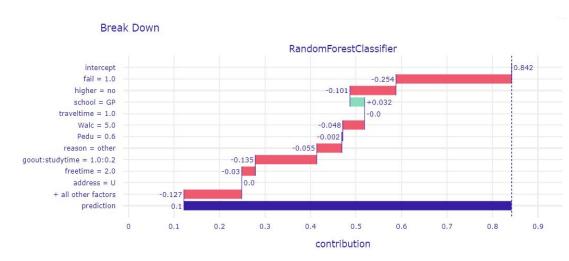




kolumny z ramki danych

### Podejście klasyfikacyjne

		G3	n
		0	15
		1	1
		5	1
count	649.000000	6	3
mean	11.906009	7	10
std	3.230656	8	35
min	0.000000	9	35
25%	10.000000		
50%	12.000000	10	97
75%	14.000000	11	104
max	19.000000	12	72
Name: G3,	dtype: float64	13	82
		14	63
statys	15	49	
zm	ienną celu	16	36
	C	17	29
		18	15
		19	2



Wyjaśnienia rezultatów jednego ze stworzonych modeli

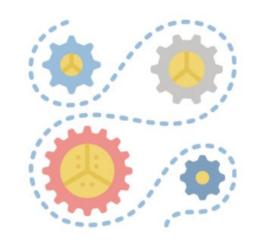
liczba wystąpień poszczególnych ocen

### Preprocessing i automatyczne wybieranie zmiennych

$$I(X;Y) = \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} P(x,y) \log \frac{P(x,y)}{P(x)P(y)}.$$

mutual information

$$\chi_c^2 = \sum \frac{(O_i - E_i)^2}{E_i}$$



statystyka chi^2, podstawa wyboru jednego z selectorów

selectKBest z selectorem chi2		selectKBest z selectorem mutual information		
failures	school_MS Fjob_health	failures	school_MS goout	
address_U failures	Pstatus_T failures	Pstatus_T failures	Fjob_health Walc	
Mjob_at_home Fjob_health	Mjob_services failures	higher_yes studytime	higher_yes Dalc	
Fjob_health reason_home	Fjob_services failures	age failures	failures^2	
reason_other failures	guardian_mother failures	failures Fedu	failures studytime	
internet ves failures	romantic yes failures	failures Dalc	failures Walc	
recursive fea	ture selection	l1 based featu	re selection	
recursive fea	ture selection absences	l1 based featu	re selection failures	
Manual Property of the Control of th	1.0011111111	age	2000	
failures	absences school_MS guardian_father	age	failures	
failures school_MS Pstatus_T	absences school_MS guardian_father	age guardian_mother failures	failures higher_yes Medu	
failures school_MS Pstatus_T schoolsup_yes nursery_yes	absences school_MS guardian_father nursery_yes failures	age guardian_mother failures failures^2	failures higher_yes Medu failures Medu	

# recznie wybrane zmienne fail reason higher school age goout Pedu Gender Relations address internet Mjob studytime

### Wybór modeli i hiperparametrów

- Logistic Regression
- SVR
- Random Forest
- Gradient Boosting



```
GradientBoostingRegressor(learning_rate=0.045, n_estimators=100, criterion='mse', random_state=0)
RandomForestRegressor(n_estimators=20, max_features=0.5, min_samples_split=3, n_jobs=-1, random_state=0)
LogisticRegression(max_iter=1000)
= SVR(C=1.5)
```

## Wyniki

	Linear Regression	SVR	Random Forest	Gradient Boosting
SelectKBest (chi2)	3.139358	2.884703	2.926594	2.875707
SelectKBest (mutual information)	2.647433	2.655581	2.603689	2.632258
RFE	2.949454	2.758008	2.716849	2.791437
L1 Based Model Selection	2.801049	2.813967	2.905404	2.834317
Hand-prepared features	2.639725	2.680089	2.770634	2.591747

RMSE baseline'u: 3.070836358822292

### Wyjaśnienie modelu

#### Break Down

