ANALIZA SKUPA PODATAKA COFFEE QUALITY DATABASE FROM CQI

ANJA CVETKOVIĆ ISTRAŽIVANJE PODATAKA I

Mere kvaliteta

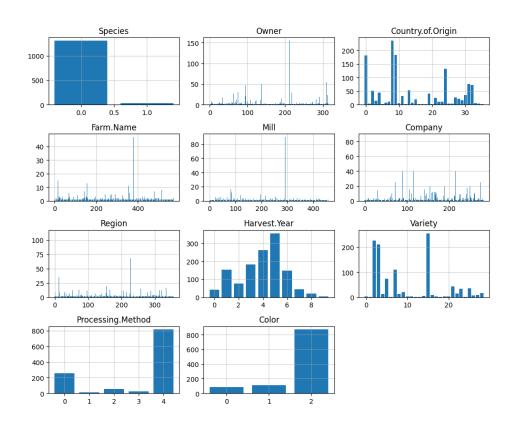
- Aroma miris kafe
- Flavor ukus kafe
- Aftertaste da li kafa ostavlja prijatnu senzaciju na paleti
- Acidity prijatna oštrinu u ukusu kafe
- Body izražava teksturu kafe, za koju je poželjno da ima težinu i da bude kremasta
- Balance ukus se ne lokaliziju na jednom mestu palete
- Uniformity konzistentnost ukusa
- Cup Cleanliness "čista šoljica" označava da se ne javljaju arome koje nisu od kafe a koje su posledica defekata
- Sweetness slatkoća koja se oseti na vrhu jezika, poželjna osobina
- Moisture preporučena vlažnost je oko 11.5% potpuno procesuiranog zrna
- Defects mogu da budu prve i druge kategorije, na osnovu uzorka od 350g, defekti prvi kategorije su skroz crna ili kisela zrna, tragovi rastinja ili kamenčići, dok su defekti druge kategorije insekti u uzorku, oštećenja od vode i slično.
- Total Cup Points zbir ocean prethodnih mera, vrednost na [1,100] i ciljna promenljiva

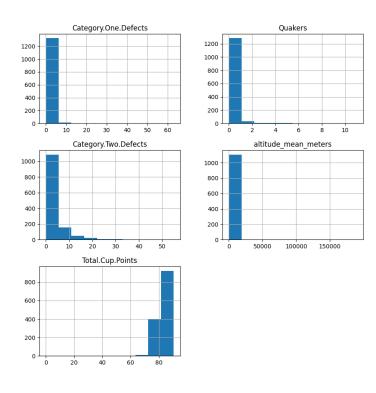
Metapodaci o zrnu

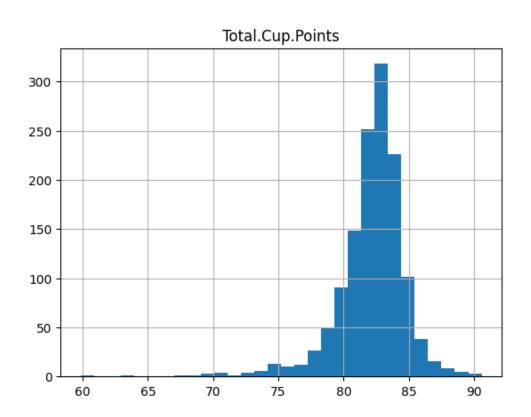
- Processing Method
- Color
- Species (Arabica/Robusta)
- Variety podvrsta
- Quakers broj nezrelih zrna

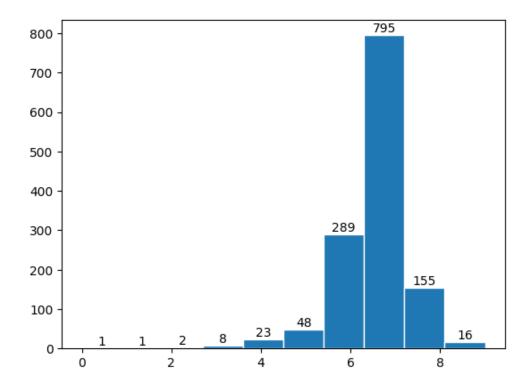
Metapodaci od poreklu

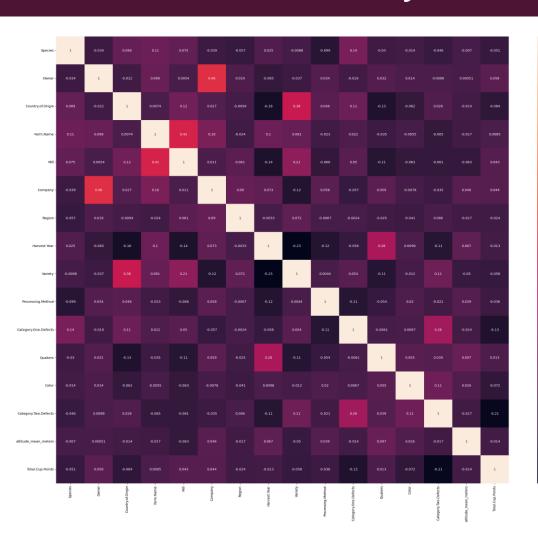
- Owner
- Country of Origin
- Farm Name
- Lot Number
- Mill
- Company
- Altitude
- Region

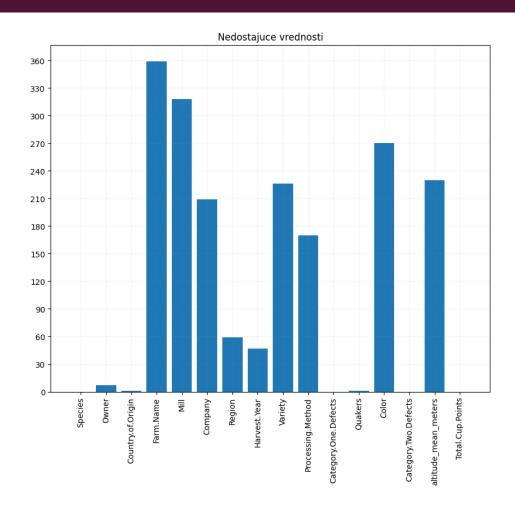


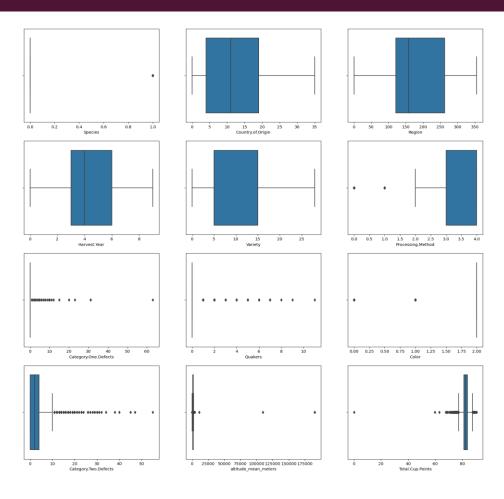












KLASIFIKACIJA-STABLA ODLUČIVANJA

Classification report for model DecisionTreeClassifier on training data ${\tt Accuracy:\ 0.6463963963963963}$

	pre	rec	spe	f1	geo	iba	sup
4.0	0.40	0.13	1.00	0.20	0.36	0.12	15
5.0	0.75	0.09	1.00	0.17	0.31	0.09	32
6.0	0.52	0.12	0.97	0.19	0.34	0.11	194
7.0	0.65	0.97	0.21	0.78	0.45	0.21	532
8.0	0.82	0.26	0.99	0.39	0.51	0.24	104
9.0	1.00	0.27	1.00	0.43	0.52	0.25	11
avg / total	0.64	0.65	0.52	0.57	0.42	0.19	888

 ${\tt Confusion} \ {\tt matrix} \ {\tt for} \ {\tt model} \ {\tt DecisionTreeClassifier} \ {\tt on} \ {\tt training} \ {\tt data}$

4 5 6 7 8 9 4 2 0 1 12 0 0 5 0 3 4 25 0 0 6 1 1 23 167 2 0 7 1 0 12 516 3 0 8 1 0 4 72 27 0

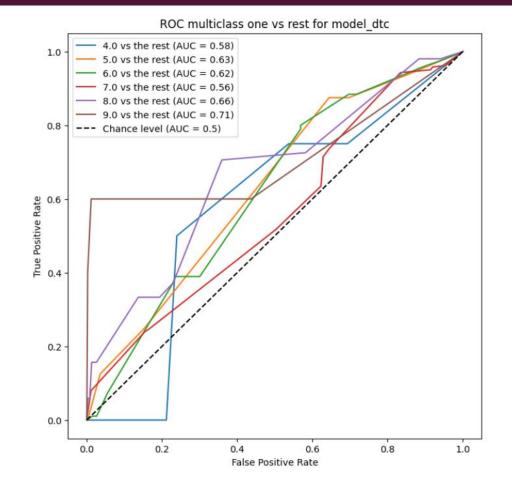
Classification report for model DecisionTreeClassifier on test data Accuracy: 0.6050228310502284

_							
	pre	rec	spe	f1	geo	iba	sup
4.0	0.00	0.00	0.99	1.00	0.00	0.00	8
5.0	0.00	0.00	1.00	1.00	0.00	0.00	16
6.0	0.33	0.08	0.95	0.13	0.28	0.07	95
7.0	0.63	0.94	0.17	0.75	0.40	0.17	263
8.0	0.57	0.16	0.98	0.25	0.39	0.14	51
9.0	0.67	0.40	1.00	0.50	0.63	0.38	5
vg / total	0.52	0.61	0.49	0.57	0.36	0.14	438

.....

Confusion matrix for model DecisionTreeClassifier on test data

	4	5	6	7	8	9
4	0	0	4	4	0	0
5	0	0	3	13	0	0
6	1	1	8	85	0	0
7	3	0	7	247	5	1
8	0	0	2	41	8	0
9	0	0	0	2	1	2



Stablo odlučivanja sa nasumičnim hiper-parametrima i nebalansiranim podacima

KLASIFIKACIJA-STABLA ODLUČIVANJA

Classification report for model DecisionTreeClassifier on training data Accuracy: 0.9326971371170266

	pre	rec	spe	f1	geo	iba	sup
4.0	1.00	0.99	1.00	0.99	0.99	0.98	467
5.0	0.93	0.96	0.98	0.94	0.97	0.94	425
6.0	0.91	0.73	0.99	0.81	0.85	0.71	192
7.0	0.94	0.78	1.00	0.85	0.88	0.76	123
8.0	0.88	0.90	0.98	0.89	0.94	0.87	296
9.0	0.91	1.00	0.97	0.95	0.98	0.97	488
avg / total	0.93	0.93	0.98	0.93	0.96	0.91	1991

Confusion matrix for model DecisionTreeClassifier on training data

	4	5	6	7	8	9	
4	460	6	0	0	1	0	
5	0	408	5	2	8	2	
6	0	13	141	4	17	17	
7	1	5	5	96	8	8	
8	0	8	4	0	265	19	
9	0	0	0	0	1	487	

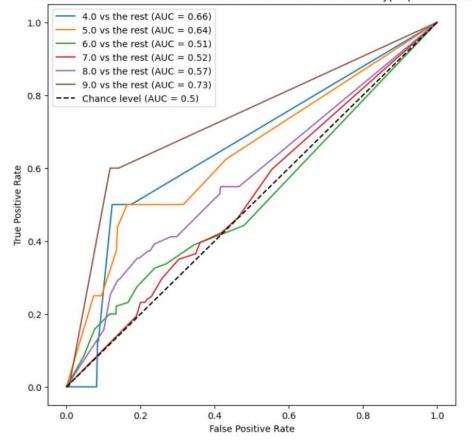
Classification report for model DecisionTreeClassifier on test data Accuracy: 0.2694063926940639

	pre	rec	spe	f1	geo	iba	sup
4.	0.04	0.12	0.94	0.06	0.34	0.11	8
5.	0.09	0.56	0.80	0.16	0.67	0.44	16
6.	0.30	0.19	0.87	0.23	0.41	0.15	95
7.	0.67	0.25	0.81	0.36	0.45	0.19	263
8.	0.20	0.41	0.78	0.27	0.57	0.31	51
9.	0.06	0.60	0.89	0.11	0.73	0.52	5
avg / tota	1 0.49	0.27	0.82	0.31	0.46	0.21	438

Confusion matrix for model DecisionTreeClassifier on test data

	4	5	6	7	8	9
4	1	6	1	0	0	0
5	2	9	2	1	0	2
6	10	25	18	20	17	5
7	12	46	35	66	68	36
8	0	8	5	11	21	6
9	0	1	0	1	0	3

ROC multiclass one vs rest for decision tree model wiht tuned hyperparameters



Stablo odlučivanja sa podešenim hiper-parametrima i balansiranim podacima

KLASIFIKACIJA-RANDOM FOREST

Classification report for model RandomForestClassifier on training data Accuracy: 0.9974886991461577

	pre	rec	spe	f1	geo	iba	sup
4.0	1.00	1.00	1.00	1.00	1.00	1.00	467
5.0	1.00	1.00	1.00	1.00	1.00	1.00	425
6.0	1.00	1.00	1.00	1.00	1.00	1.00	192
7.0	1.00	0.98	1.00	0.99	0.99	0.98	123
8.0	0.99	0.99	1.00	0.99	1.00	0.99	296
9.0	0.99	1.00	1.00	1.00	1.00	1.00	488
avg / total	1.00	1.00	1.00	1.00	1.00	1.00	1991

Confusion matrix for model RandomForestClassifier on training data

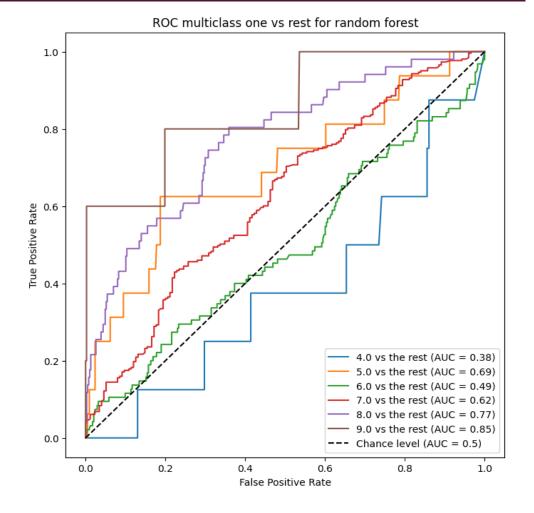
4 5 6 7 8 9 4 466 0 0 0 1 0 5 0 425 0 0 0 0 6 0 0 192 0 0 0 7 0 0 0 121 1 1 8 0 0 0 0 294 2

Classification report for model RandomForestClassifier on training data Accuracy: 0.3401826484018265

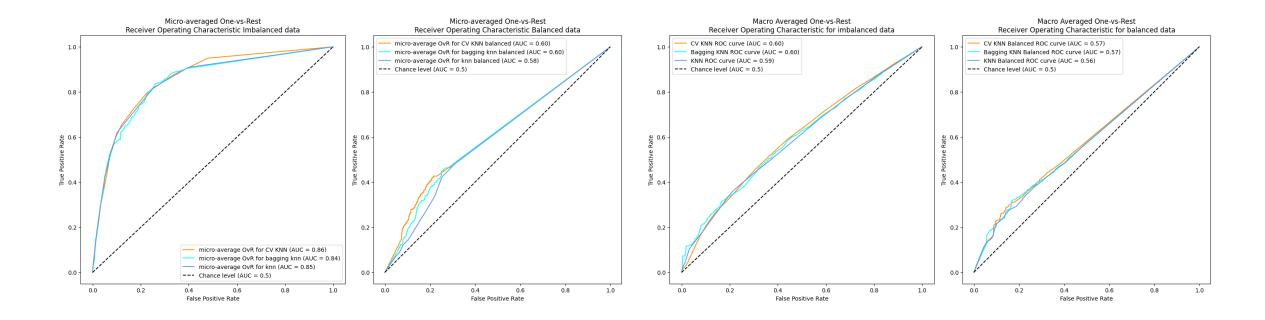
	pre	rec	spe	f1	geo	iba	sup
4.0	0.00	0.00	0.92	1.00	0.00	0.00	8
5.0	0.08	0.31	0.87	0.13	0.52	0.26	16
6.0	0.28	0.26	0.81	0.27	0.46	0.20	95
7.0	0.70	0.32	0.79	0.44	0.51	0.24	263
8.0	0.29	0.61	0.80	0.39	0.70	0.48	51
9.0	0.12	0.60	0.95	0.19	0.75	0.55	5
avg / total	0.52	0.34	0.81	0.40	0.51	0.26	438

Confusion matrix for model RandomForestClassifier on training data

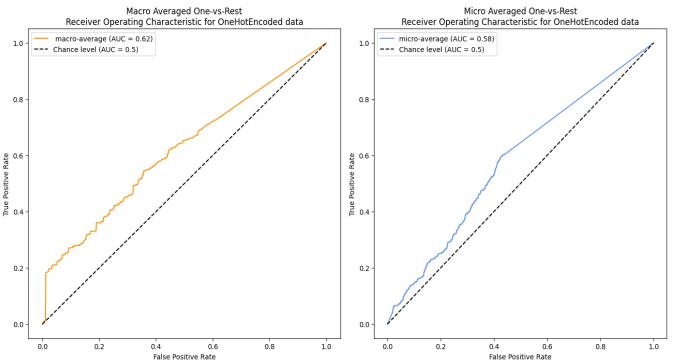
4 5 6 7 8 9 4 0 6 1 1 0 0 5 4 5 4 2 1 0 6 18 13 25 23 15 1 7 12 32 57 85 59 18 8 1 3 3 9 31 4 9 0 0 0 1 1 3

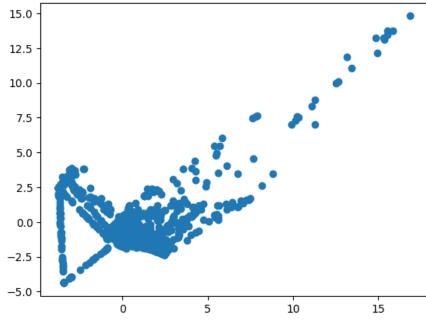


KLASIFIKACIJA-KNN



KLASIFIKACIJA-KNN



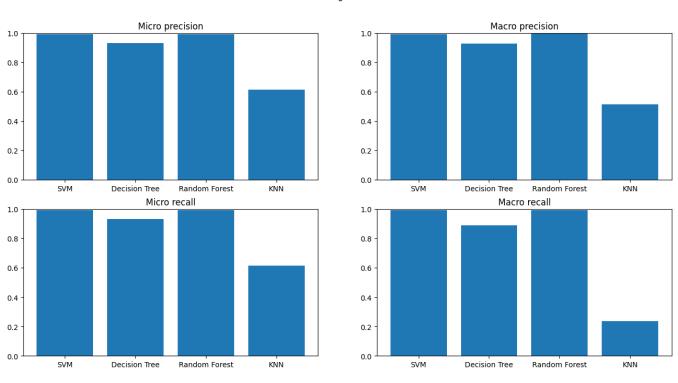


KLASIFIKACIJA-SVM

lassification report for model SVC on training data					Classification report for model SVC on test data										
					f1-score								f1-score		
4.0		1.00	0	1.00	1.00	467			4.0		0.03	0.12	0.05	8	
5.0		1.00	0	1.00	1.00	425			5.0		0.15	0.50	0.23	16	
6.0		1.00	0	0.99	1.00	192			6.0		0.23	0.20	0.21	95	
7.0		0.99	9	0.99	0.99	123			7.0		0.68	0.38	0.49	263	
8.0		0.99	9	0.98	0.98	296			8.0		0.22	0.39	0.28	51	
9.0		0.99	9	1.00	0.99	488			9.0		0.04	0.20	0.06	5	
accuracy					0.99	1991	ā	accu	racy				0.34	438	
macro avg		1.00	0	0.99	0.99	1991	ma	acro	avg		0.22	0.30	0.22	438	
eighted avg		0.99	9	0.99	0.99	1991	weigh	nted	avg		0.49	0.34	0.38	438	
onfusion mat	rix f	or mo	odel		training d	ata	Confi								
4 5	6	7	8	9						7					
467 0	0	0	0	0			4 1	1 4	4 1	2	0	0			
1 424	0	0	0	0			5 1	1 :	3 1	4	2	0			
0 0	191	0	1	0			6 19	5 1	3 19	28	17	3			
	0	122	1	0			7 17	7 2	5 56	100	48	16			
0 0								_							
	0	1	289	6			8 2	2 :	25	14	20	8			

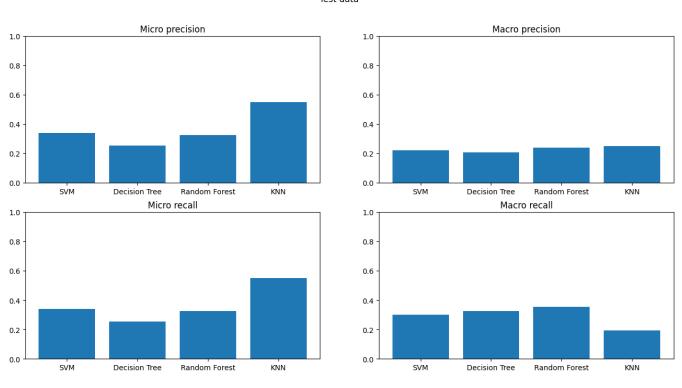
KLASIFIKACIJA



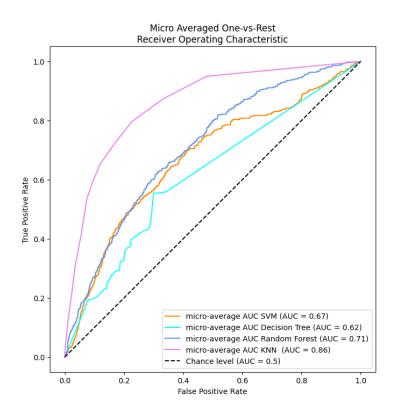


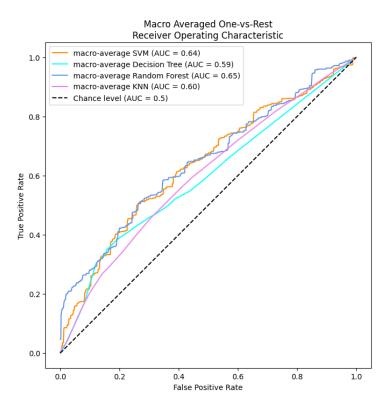
KLASIFIKACIJA



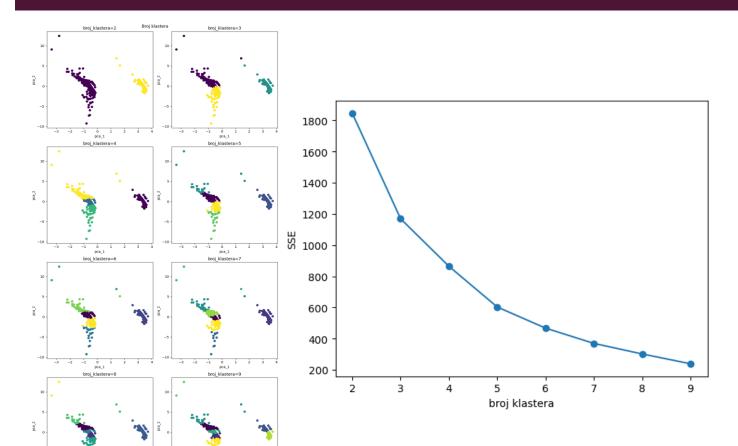


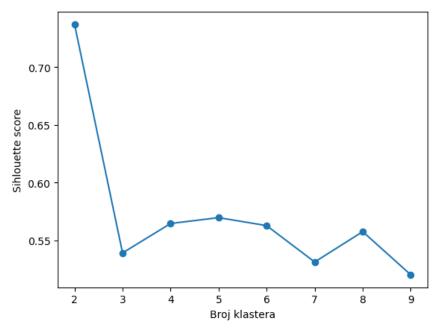
KLASIFIKACIJA



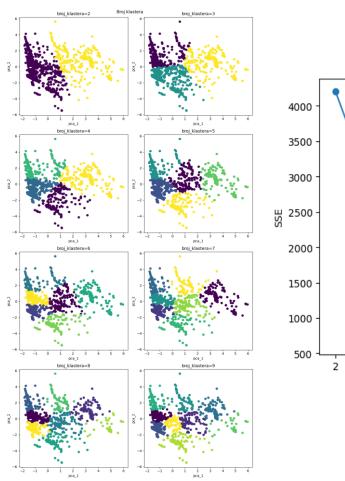


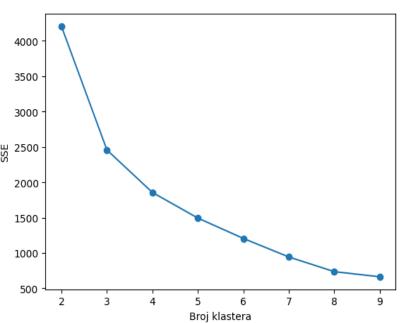
KLASTEROVANJE-KMEANS

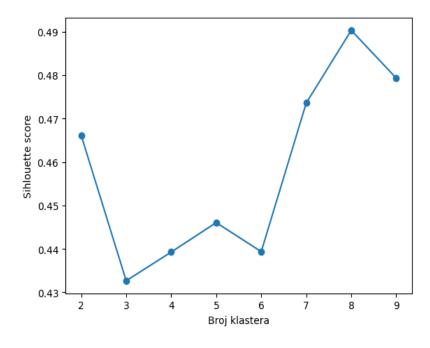




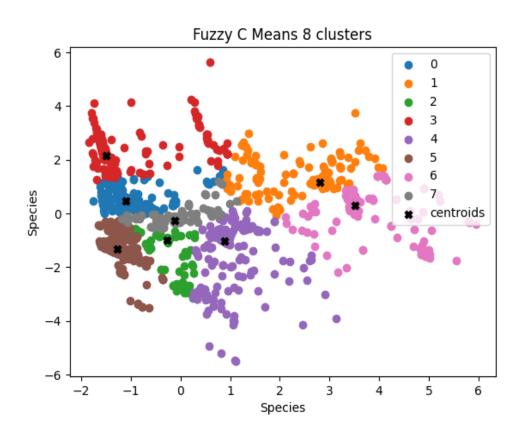
KLASTEROVANJE-KMEANS





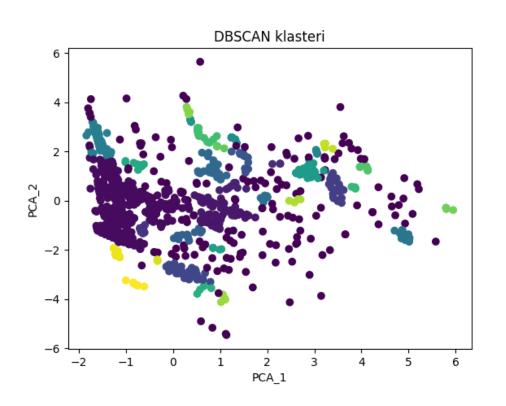


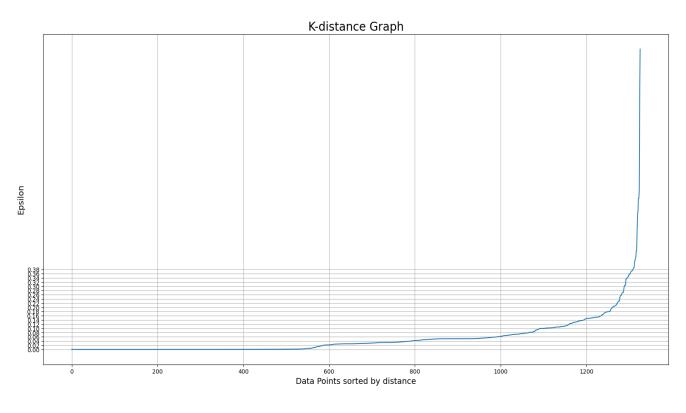
KLASTEROVANJE-FUZZY CMEANS





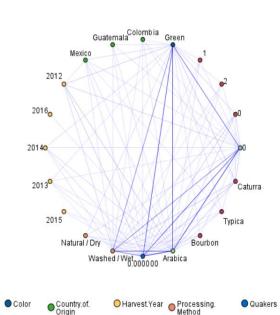
KLASTEROVANJE-DBSCAN





PRAVILA PRIDRUŽIVANJA-APRIORI

Consequent	Antecedent	Support %	Confidence %	Lift	Γ
Harvest.Year = 2012	Variety = Typica	10.256	82.353	2.676	Ì.
	Country.of.Origin = Mexico	10.250	02.333	2.070	ľ
Harvest.Year = 2012	Variety = Typica				
	Country.of.Origin = Mexico	10.256	82.353	2.676	
I.	Quakers = 0.000000				
Harvest.Year = 2012	Variety = Typica				
	Country.of.Origin = Mexico	10.256	82.353	2.676	
	Species = Arabica				
Harvest.Year = 2012	Variety = Typica				
	Country.of.Origin = Mexico	10.256	82.353	2.676	
	Quakers = 0.000000	10.230	62.555	2.070	
	Species = Arabica				
Variety = Caturra	Country.of.Origin = Colombia	10.935	95.172	2.64	
	Category.One.Defects = 0	10.935	55.172	2.04	
Variety = Caturra	Country.of.Origin = Colombia				
	Category.One.Defects = 0	10.935	95.172	2.64	
I.	Species = Arabica				
Variety = Caturra	Country.of.Origin = Colombia				
	Category.One.Defects = 0	10.256	94.853	2.631	
	Quakers = 0.000000				
Variety = Caturra	Country.of.Origin = Colombia				
	Category.One.Defects = 0	10.256	94.853	2.631	
	Quakers = 0.000000	10.250	04.000	2.001	
	Species = Arabica				
Variety = Caturra	Country.of.Origin = Colombia				
	Processing.Method = Washed / Wet	10.709	93.662	2.598	
	Color = Green				
Variety = Caturra	Country.of.Origin = Colombia				
	Processing.Method = Washed / Wet	10.709	93.662	2.598	
	Color = Croon	10.700	0.002	2.000	\perp



Species

Variety

Category.One. Category.Two.
Defects

ZAKLJUČAK

- Poteškoće i izazovi
- Skup podataka ili tehnika?
- Pitanja?