



Master in Computer Vision *Barcelona*

Module 3: Machine learning for computer vision

Project: Bag of Visual Words Image Classification

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Credit to Marçal Rossinyol

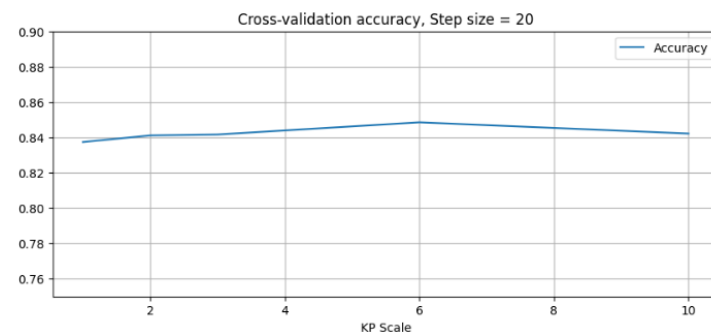
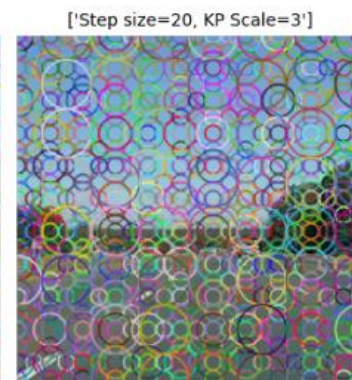
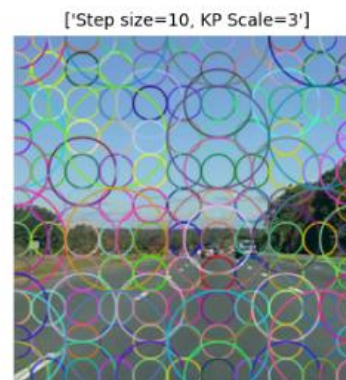
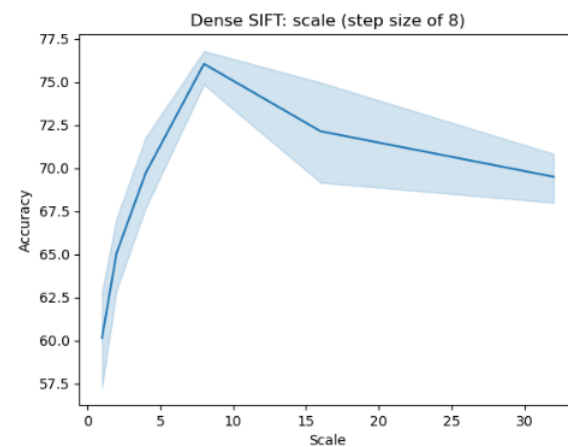
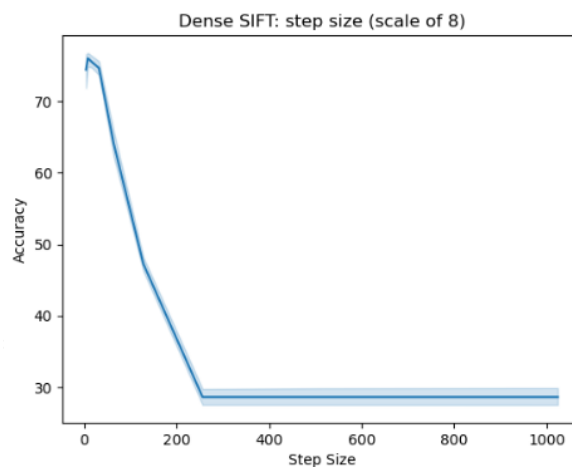
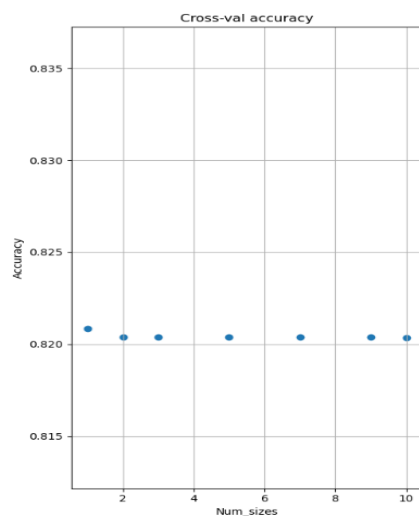
Tasks to do

Improve the BoVW code with:

- Dense SIFT (with tiny steps and different scales!)
- L2-norm - power norm
- SVM classifier
- Scalers (norms)
- Cross-validation
- Linear, RBF and histogram intersection kernels
- Spatial Pyramids
- Fisher Vectors (OPTIONAL)

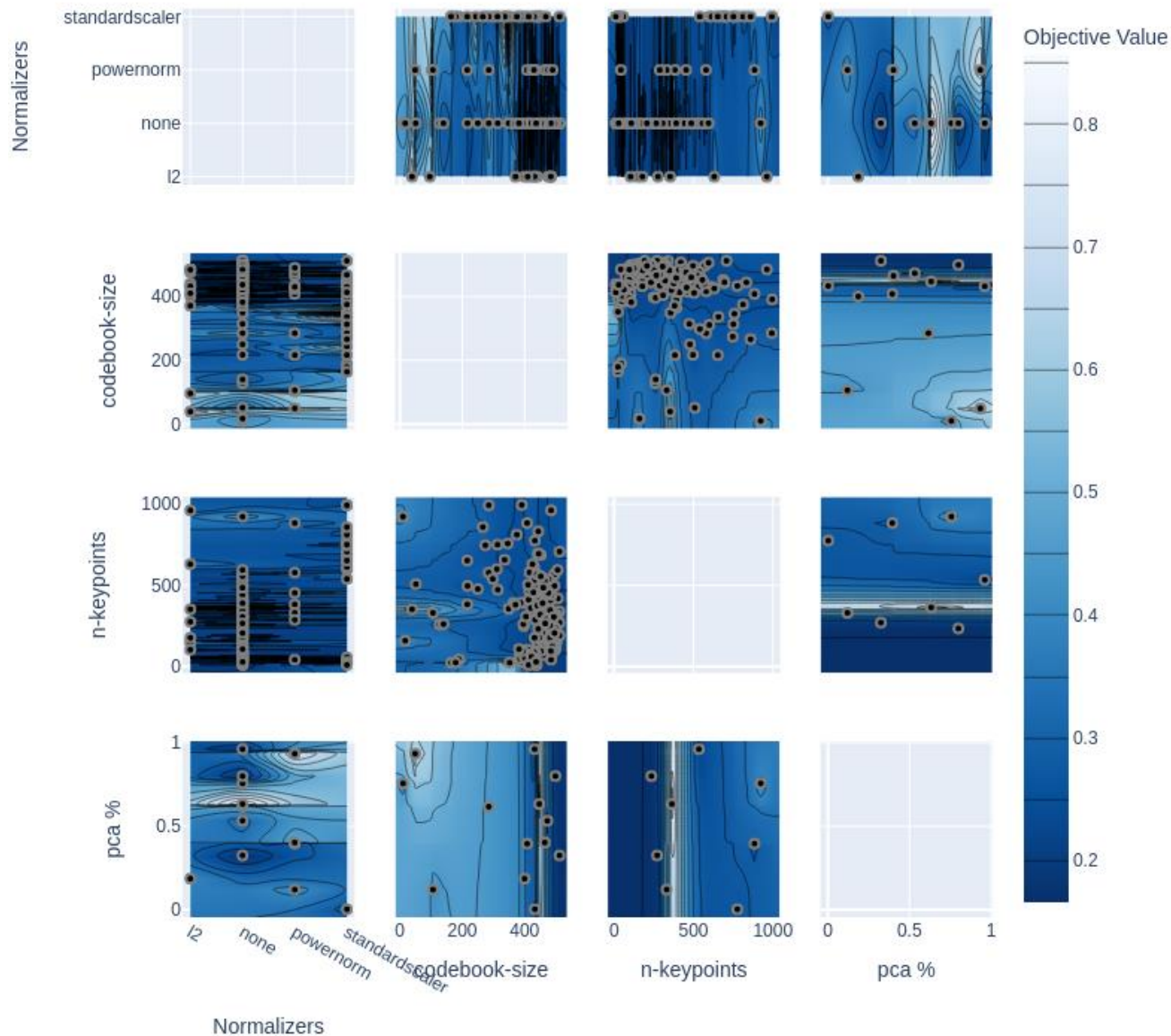
Step size & scale

Small up to a certain limit

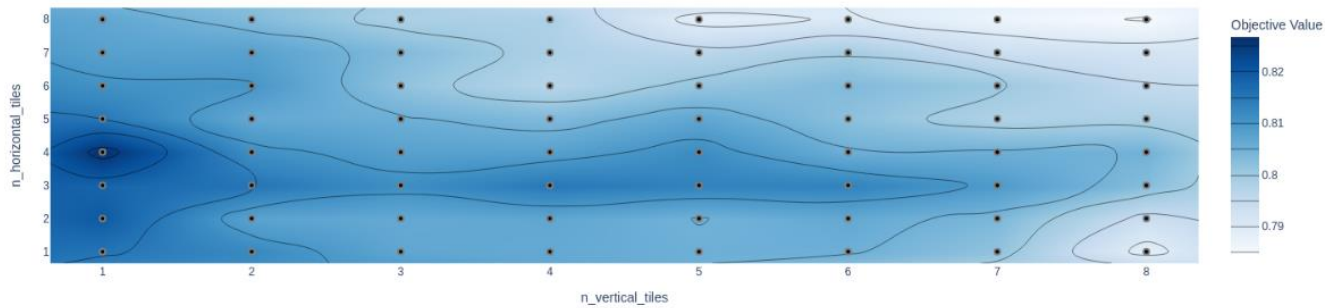
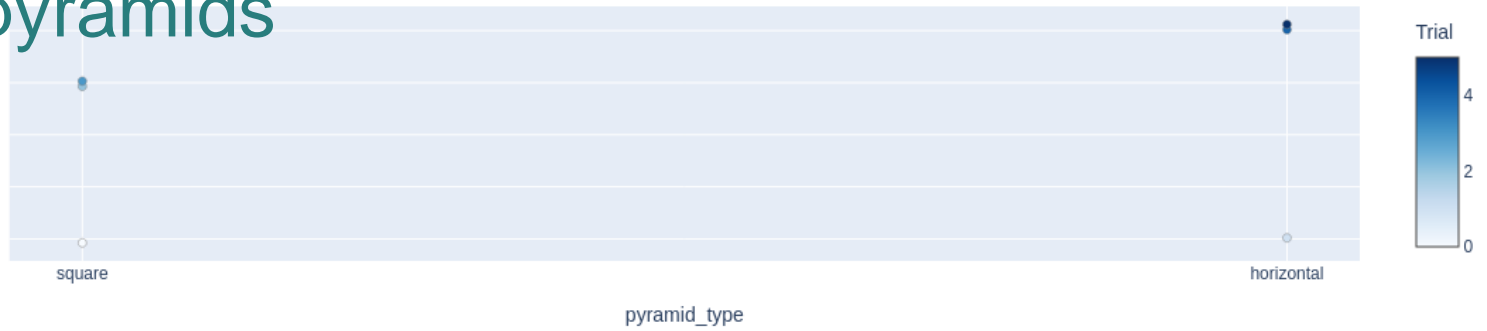


Parameter selection

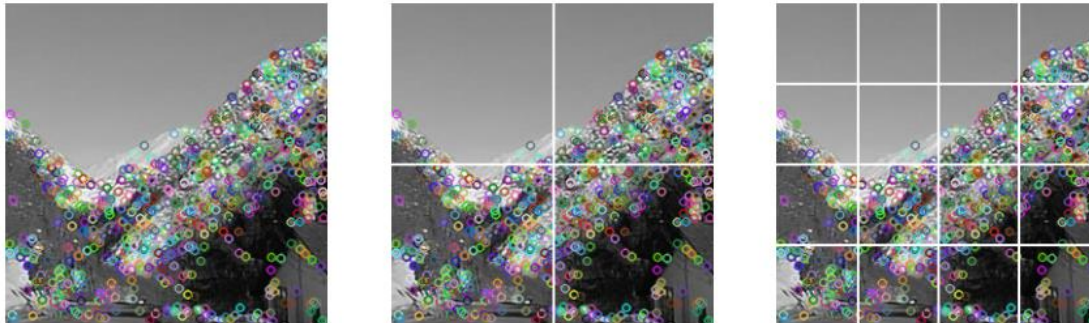
Contour Plot



Spatial pyramids

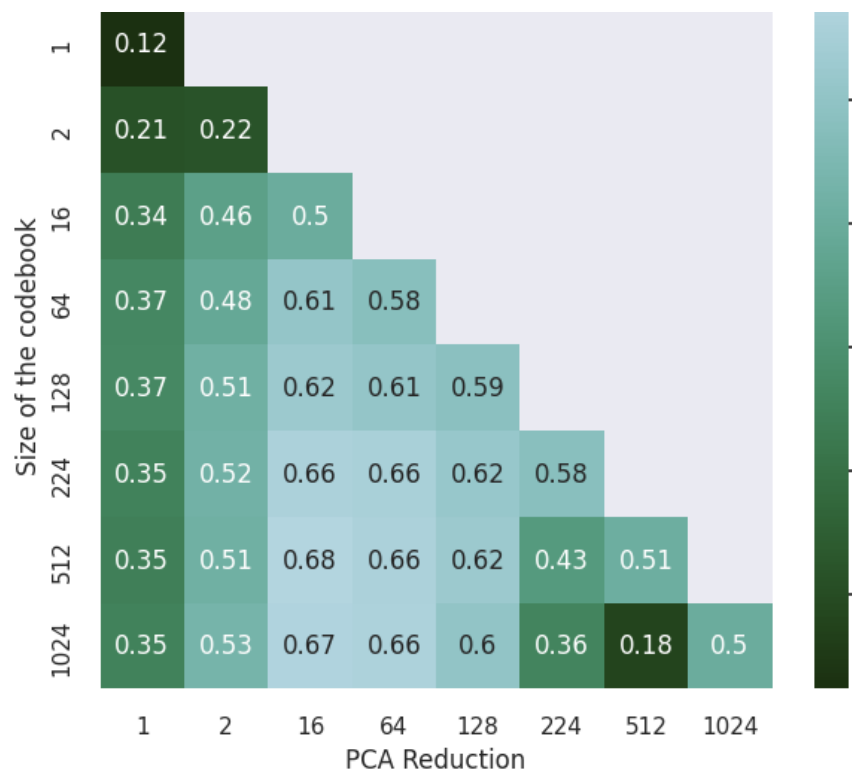


Spatial pyramid divisions with keypoints

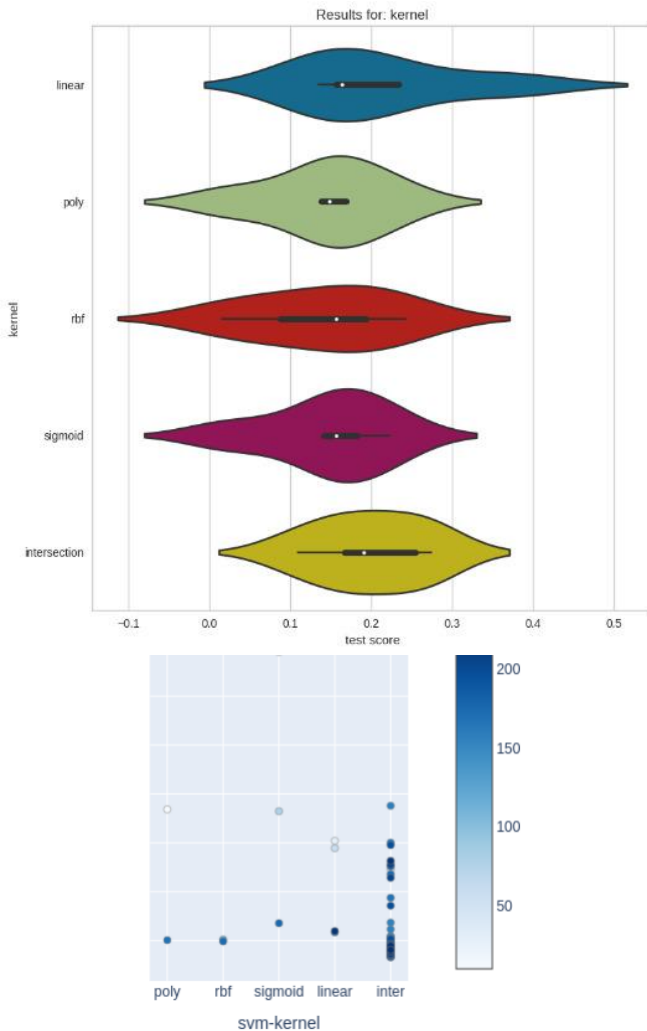


SP => dense

Pca

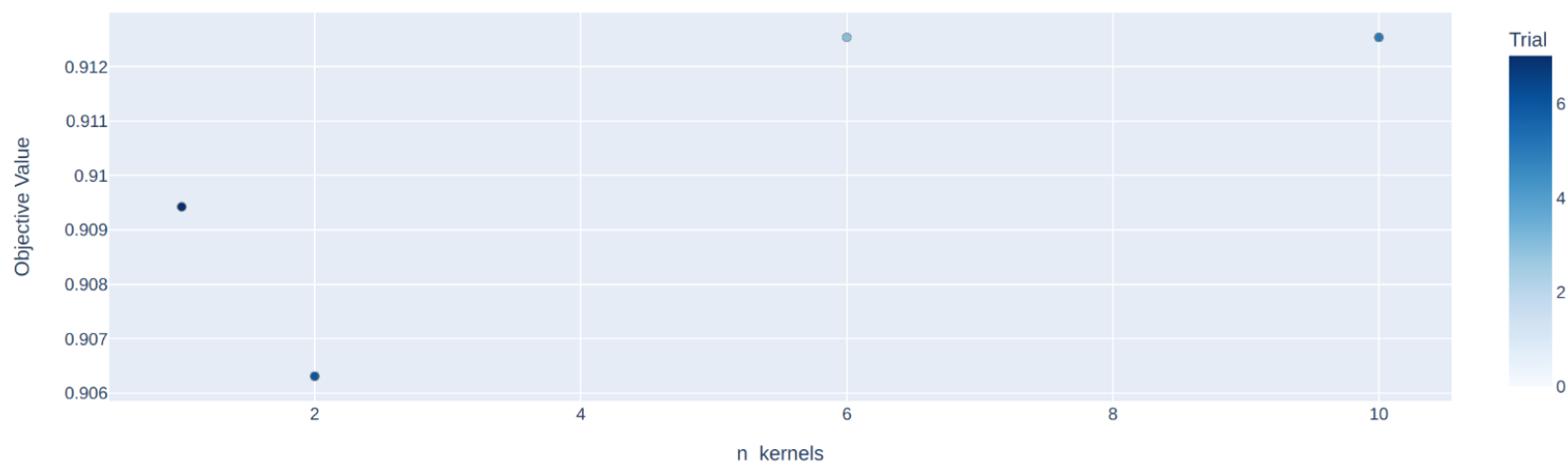


Svm kernel & others

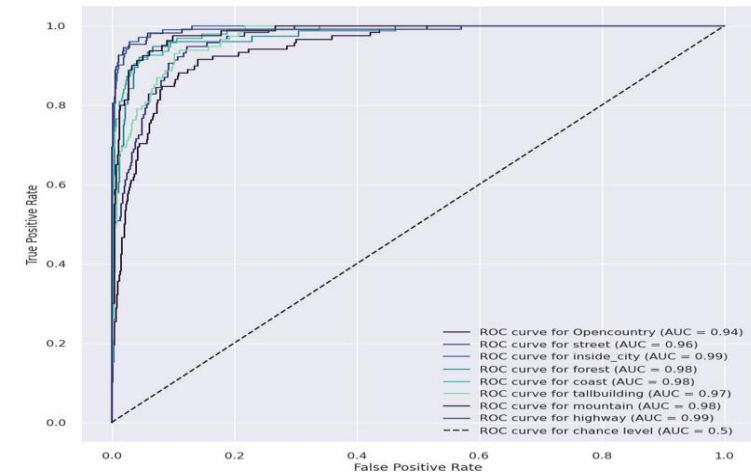
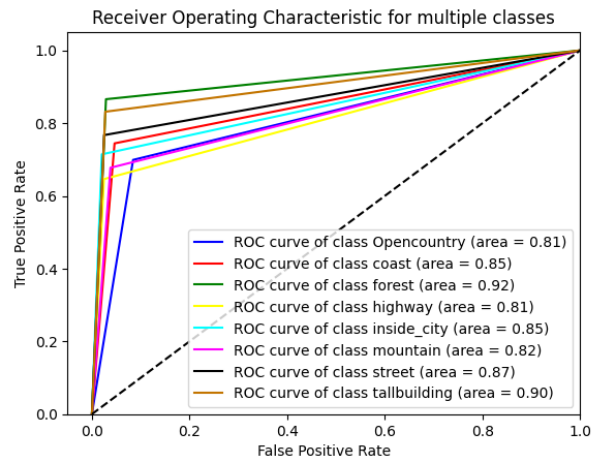
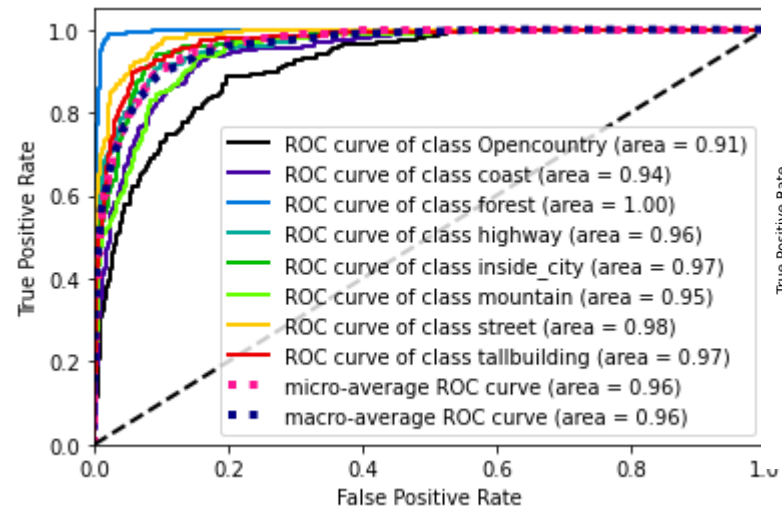


Classifier	Test Accuracy	Accuracy STD	Fit Time
SVM	0.738969	0.008270	16.453848
KNN	0.677299	0.012100	16.475282
Logistic	0.689527	0.021052	16.384468
Random Forest	0.717172	0.014285	16.590544
Naive Bayes	0.620415	0.027716	17.176783

Fisher vectors



On accuracy some hot 90ies but reported 83.5



Other issues

representation??



No way



Hard cases ...

mountain
ClassAccuracy: 88.60
GT: mountain,
Pred: Opencountry



GT: mountain,
Pred: Opencountry



street
ClassAccuracy: 83.75
GT: street,
Pred: tallbuilding



GT: street,
Pred: inside_city



tallbuilding
ClassAccuracy: 92.59
GT: tallbuilding,
Pred: mountain



GT: tallbuilding,
Pred: forest



But, not so hard ☹

> wrongly predicted as Opencountry

131
coast



163
coast



215
coast

146
coast

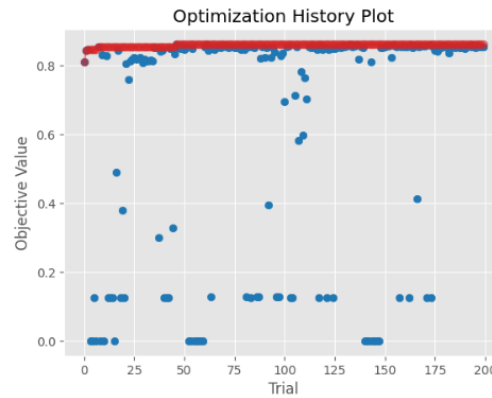


166
coast



224
coast

History (slice) not very informative

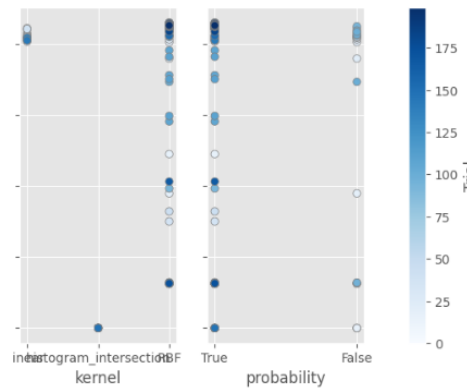


No need to get all the performance in all cases ALWAYS

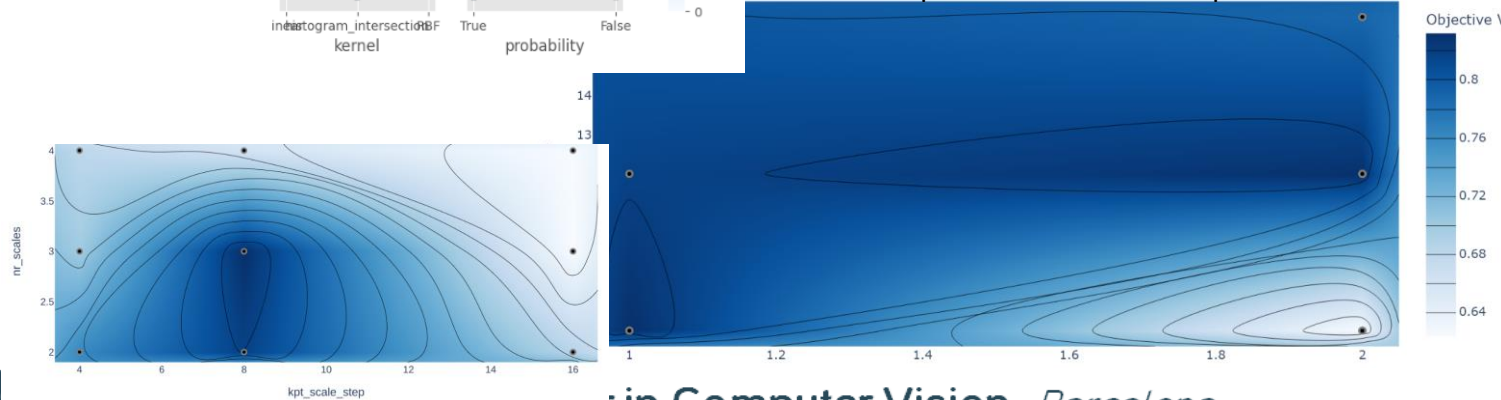
Distance metric	Base Accuracy	PCA Accuracy	LDA Accuracy	Computing time
Euclidean	78.69%	76.46%	84.01%	30.26s
Manhattan	80.05%	75.34%	84.01%	31.18s
Chebyshev	59.85%	73.73%	84.39%	31.32s
Minowski	78.69%	78.07%	84.01%	31.51s
Hamming	37.30%	14.13%	14.13%	31.35s
Canberra	62.70%	75.84%	82.65%	32.19s
Braycurtis	80.05%	76.70%	84.51%	31.35s

	label	accuracy	f1_score
0	Opencountry	0.912020	0.687225
1	coast	0.925651	0.767442
2	forest	0.966543	0.877828
3	highway	0.951673	0.715328
4	inside_city	0.962825	0.836957
5	mountain	0.944238	0.792627
6	street	0.975217	0.868421
7	tallbuilding	0.957869	0.844037
8	OVERALL	0.796248	0.796536

Slice Plot



Samples to create heat map



Group	grade
1	10
2	8
3	5
4	6
5	*
6	10
7	8
8	7
9	9
10	10

* No figures shown