

pipeline__dataset__v2

August 24, 2025

1 PIPELINE: model 4, dataset v2(<https://app.roboflow.com/sidsproject/baused-pehn2-8iy1j/generate/preprocessing>), different types of embeddings

1.1 PIPELINE

1.1.1 1. load model face__detection

1.1.2 2. extract embeddings from dataset

1.1.3 3. retrieval to evaluate embeddings goodness

1.1.4 4. train model classification

1.2 1. load the model face__detection

Step 1 is the same for each pipeline, so we do it once at the beginning.

Import dependencies

```
[1]: %load_ext autoreload
      %autoreload 2

      from embeddings import EmbeddingBuilder
      import os
      import ipynbname

      from retrieval import ImageRetrieval
      from classifier import Classifier
      import file_manager

      project_dir = f"{os.getcwd()}.
      ↪split('SIDS_revelation_project')[0]}SIDS_revelation_project"
      image_dataset_path = f"{project_dir}/datasets/onback_onstomach_v2"

      model_path = f"{project_dir}/models/4.weights/best.pt"
```

Choose how to process the dataset: 1. “extract_features”: extracting features and labels
2. “extract_features_imageswithinference”: extracting features and labels and saving predicted images with bboxes
3. “load”: loading features and labels

```
[2]: emb_builder = EmbeddingBuilder(model_path, image_dataset_path, "load")
```

Extracting dataset info from .coco.json

file:-----

Dataset contains 1506 valid samples, and labels are {'baby_on_back': 1,
'baby_on_stomach': 2}

Loading features from

.csv-----

Features loaded succesfully, in particular there are 1506 files in the dataset

Embedding builder initialized

successfully-----

Face detection model: 4 (YOLOv8)

Dataset: /home/terra/Documents/AI_engineering/SIDS-

project/python_project/SIDS_revelation_project/datasets/onback_onstomach_v2

Dataset dimension: 1506

Dataset labels: {'baby_safe': 0, 'baby_unsafe': 1}

1.3 2. Extract embeddings from dataset

Create embeddings

```
[3]: embeddings = emb_builder.embedding_flags()
```

Creation of flags features

embedding-----

1506 embedding created


```
[4]: embeddings.head()
```

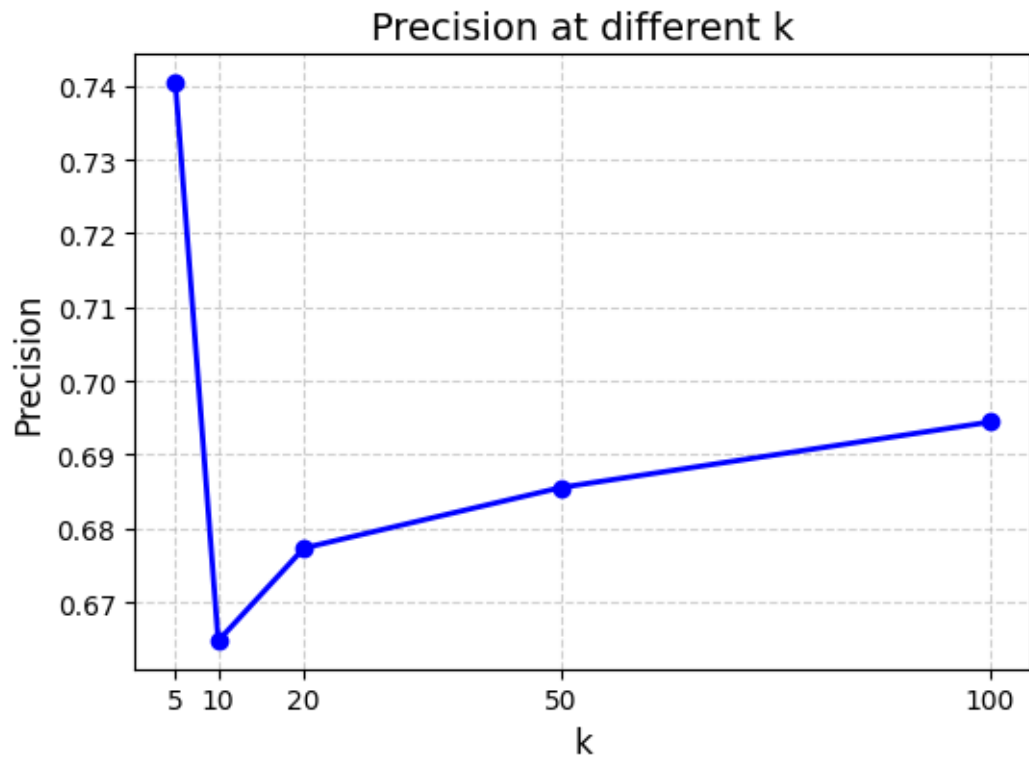
```
[4]:   flag_eye1  flag_eye2  flag_nose  flag_mouth
0         0         0         0         0
1         1         1         1         1
2         0         0         0         0
3         1         1         1         1
4         1         1         1         1
```

1.4 3. Retrieval to evaluate embedding goodness

```
[5]: ret = ImageRetrieval(embeddings, emb_builder.y, emb_builder.image_paths,
    ↪ image_dataset_path, emb_builder.classes_bs)
    ret.report("euclidean")
```

Precision at different

k:-----



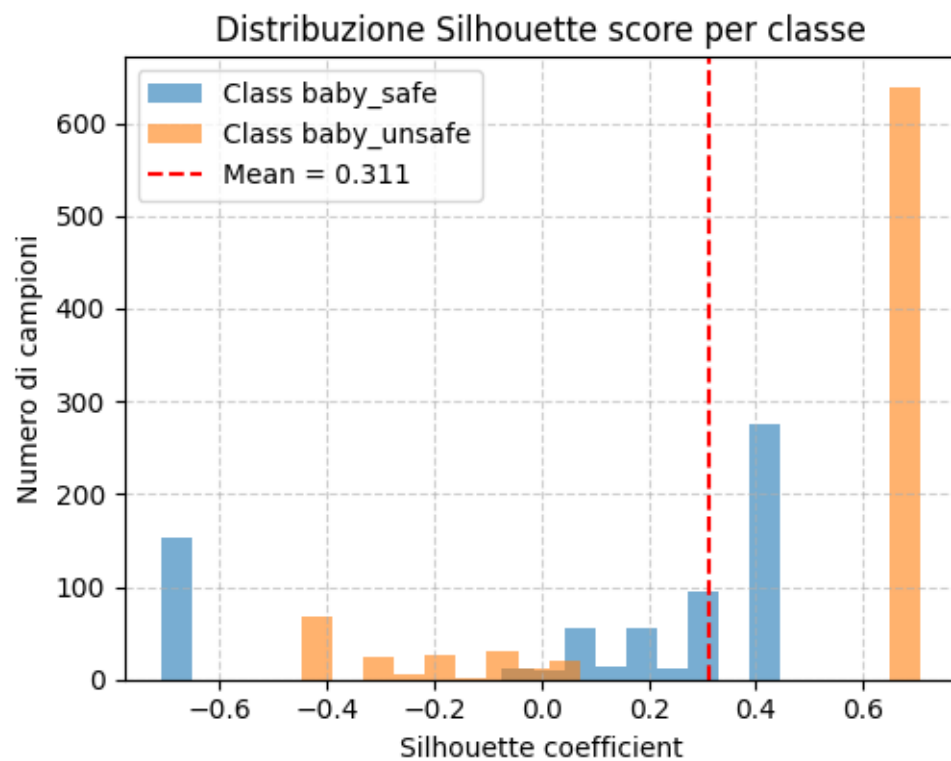
Recall at

R:-----

0.6558580010669358

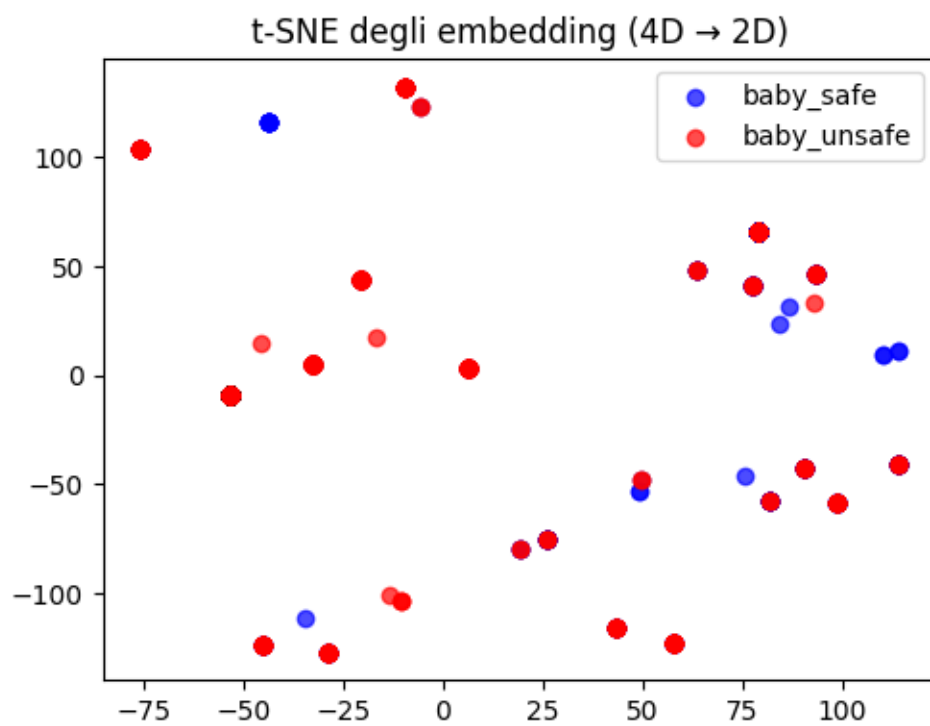
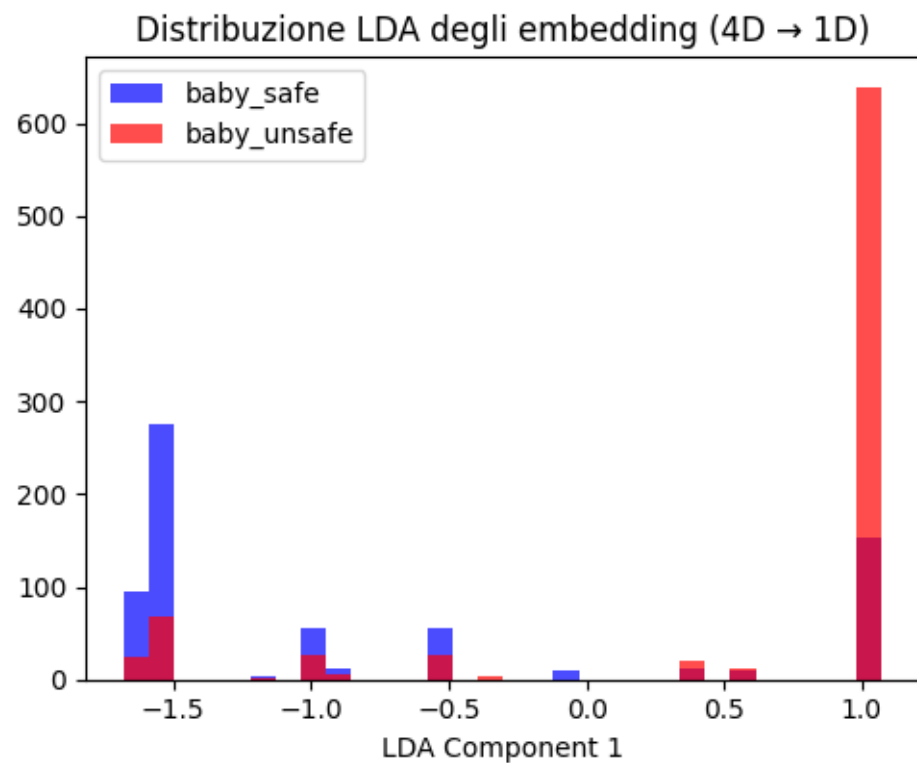
Silhouette

score-----

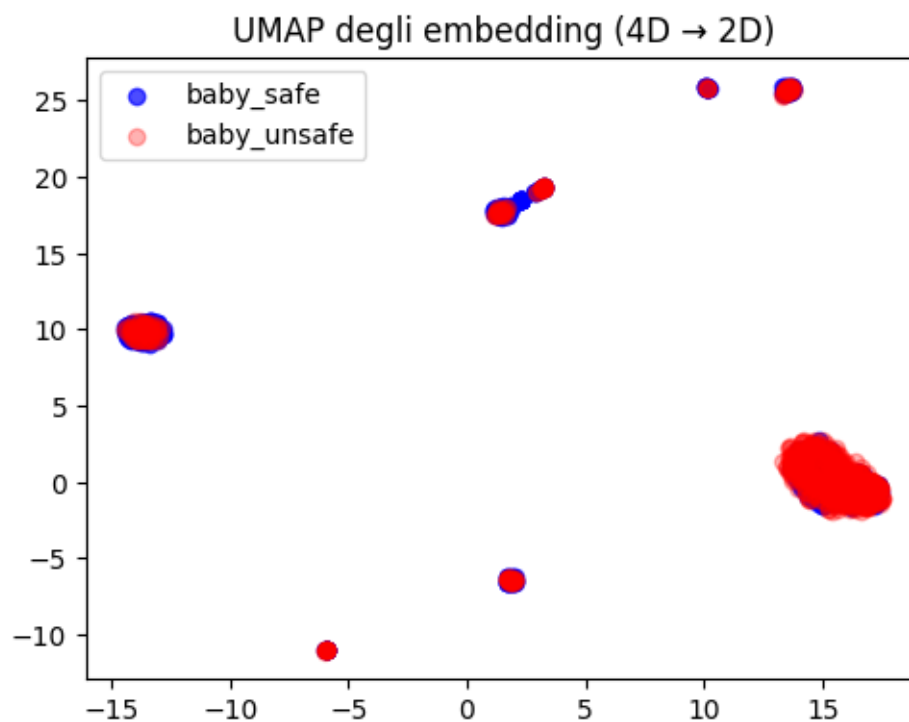


Embeddings

distributions-----



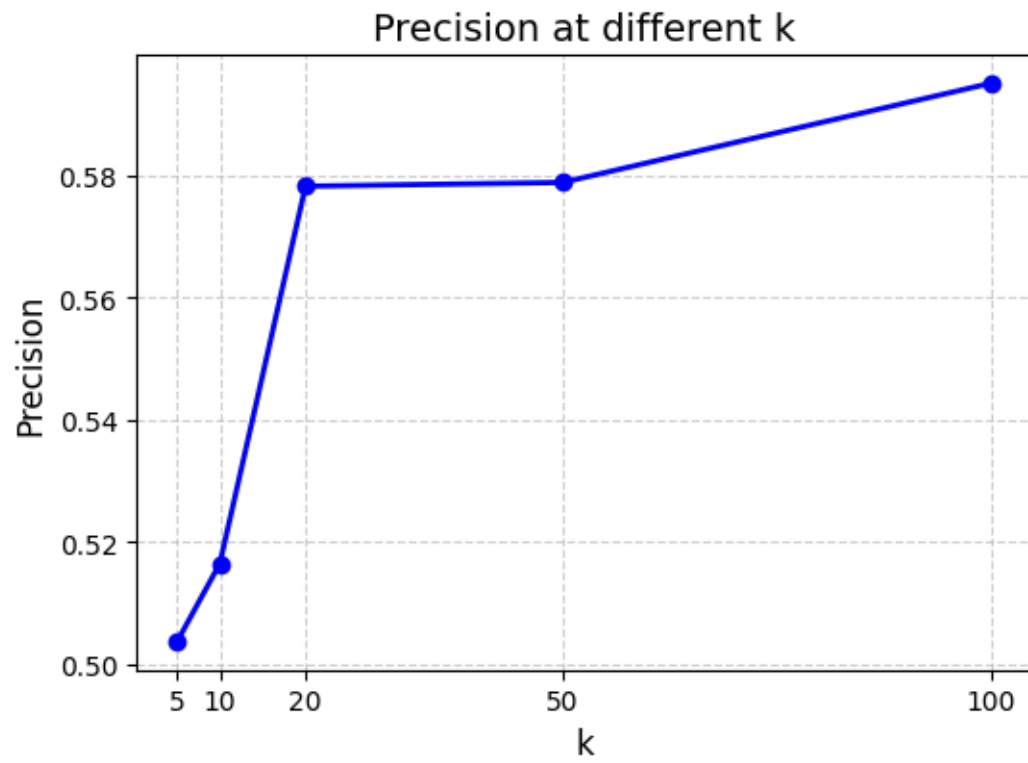
```
/home/terra/anaconda3/envs/SIDS_revelation_project/lib/python3.10/site-  
packages/umap/umap_.py:1952: UserWarning: n_jobs value 1 overridden to 1 by  
setting random_state. Use no seed for parallelism.  
warn(
```



```
[6]: ret.report("cosine")
```

Precision at different

k:-----

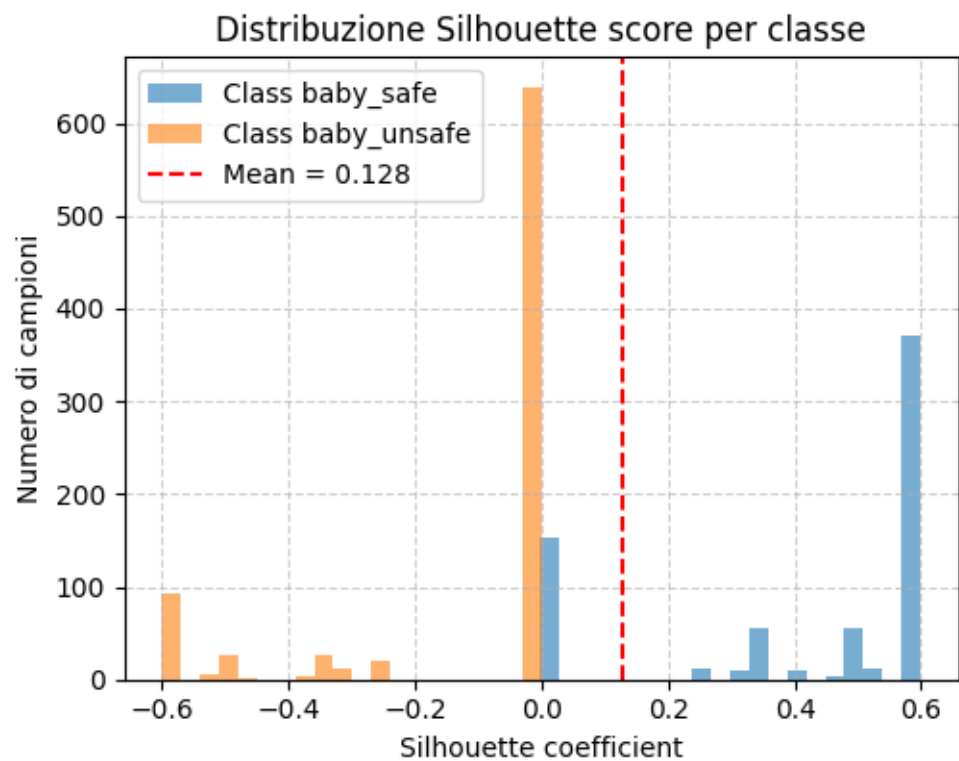


Recall at

R-----
0.5754122273017442

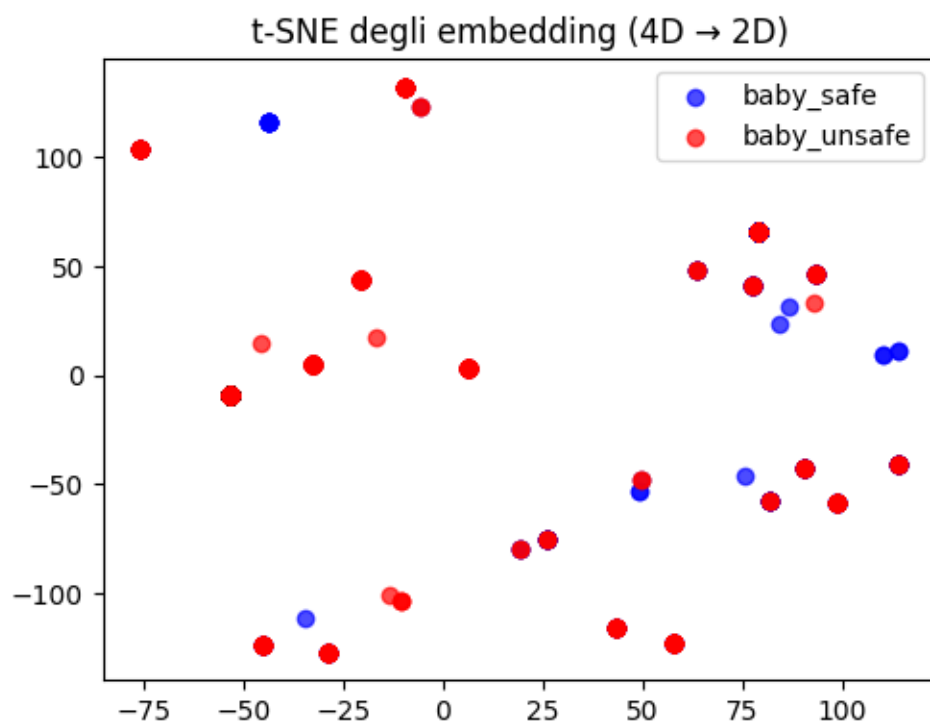
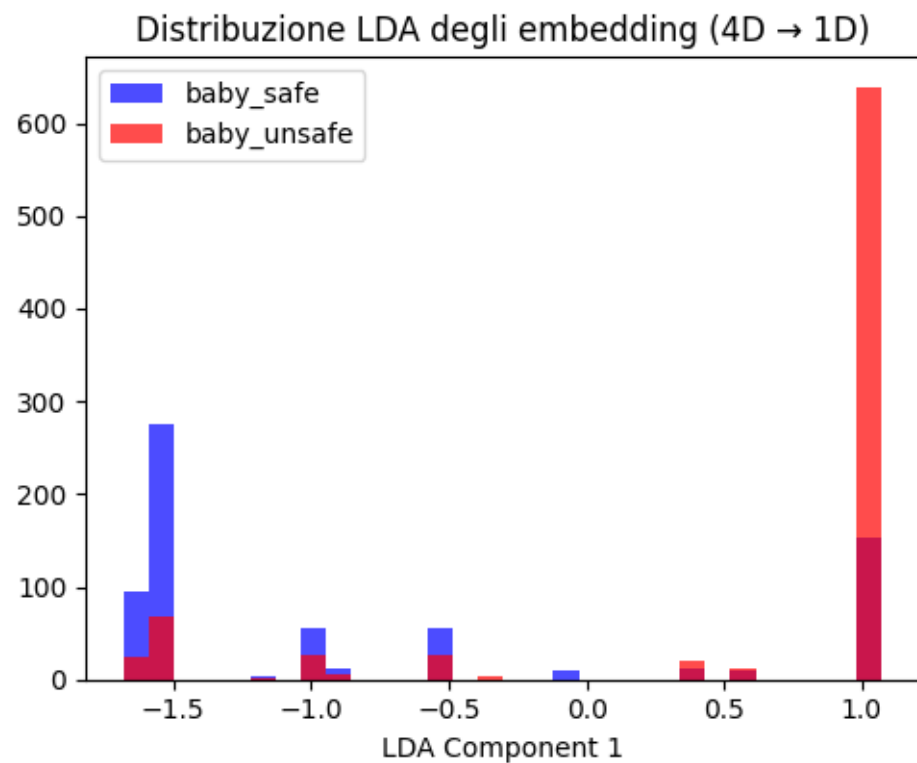
Silhouette

score-----



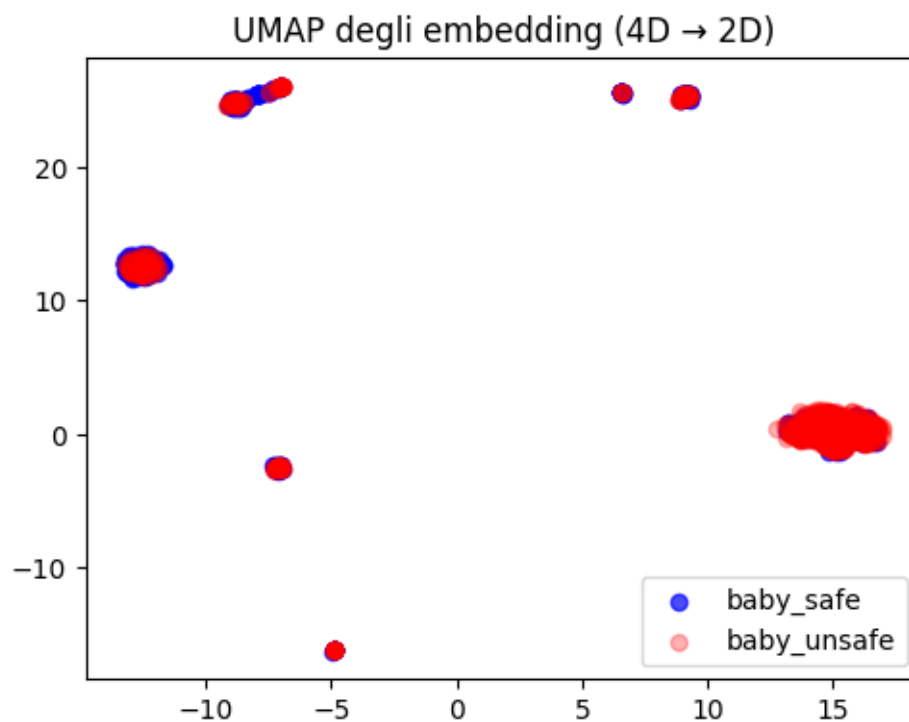
Embeddings

distributions-----



```
/home/terra/anaconda3/envs/SIDS_revelation_project/lib/python3.10/site-  
packages/umap/umap_.py:1952: UserWarning: n_jobs value 1 overridden to 1 by  
setting random_state. Use no seed for parallelism.
```

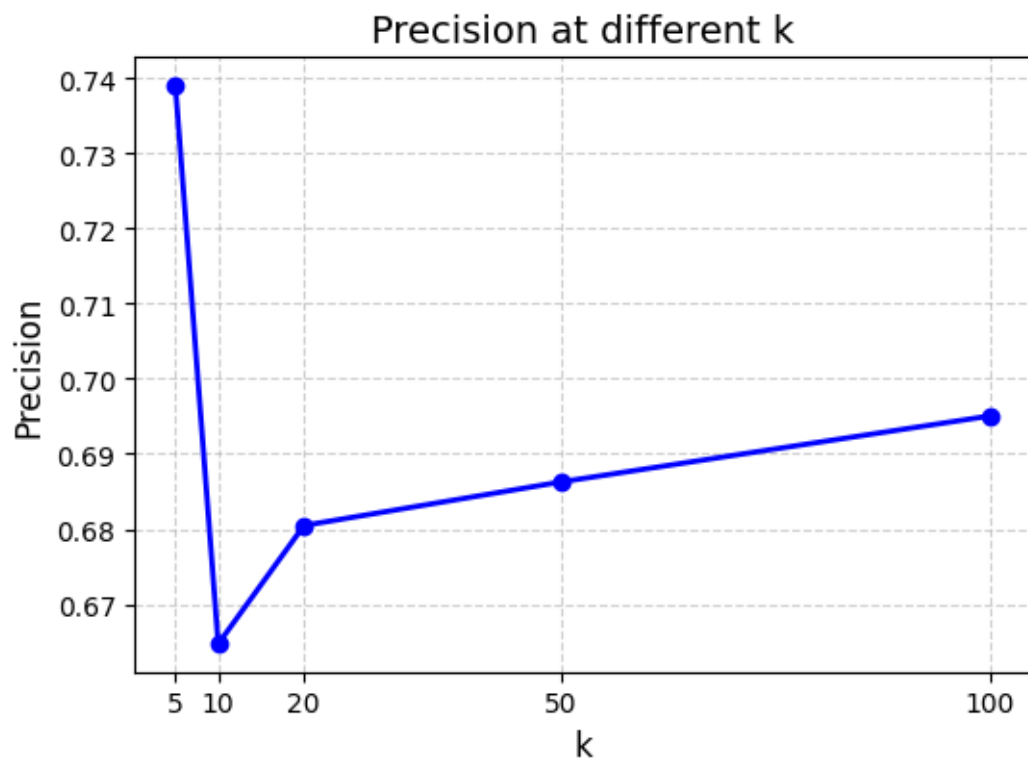
```
warn(
```



```
[7]: ret.report("minkowski")
```

Precision at different

k:-----



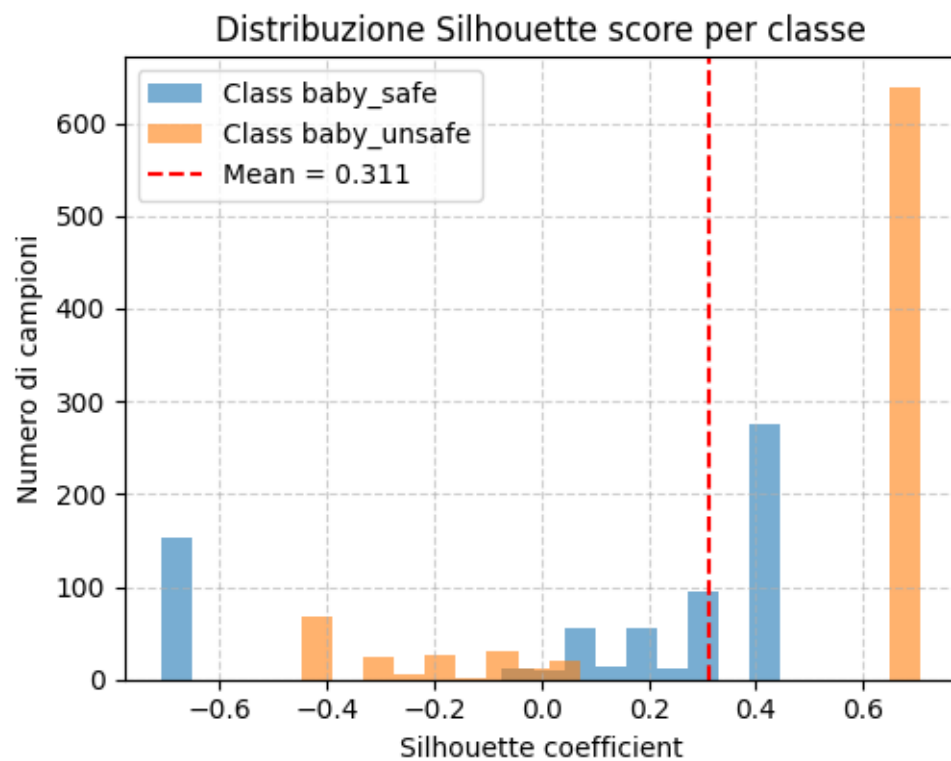
Recall at

R-----

0.6502013355033254

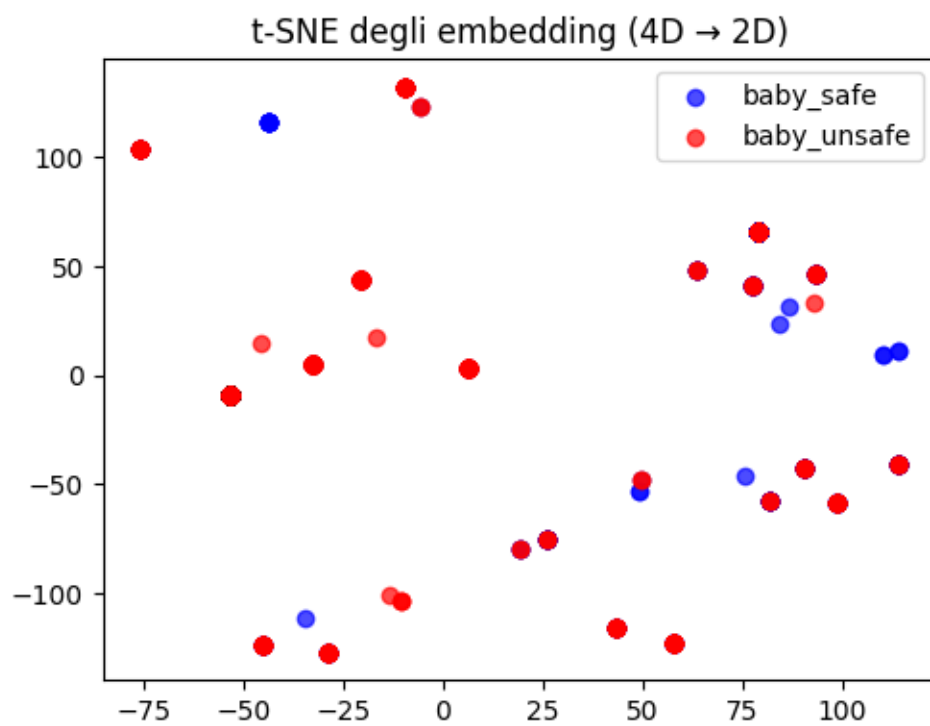
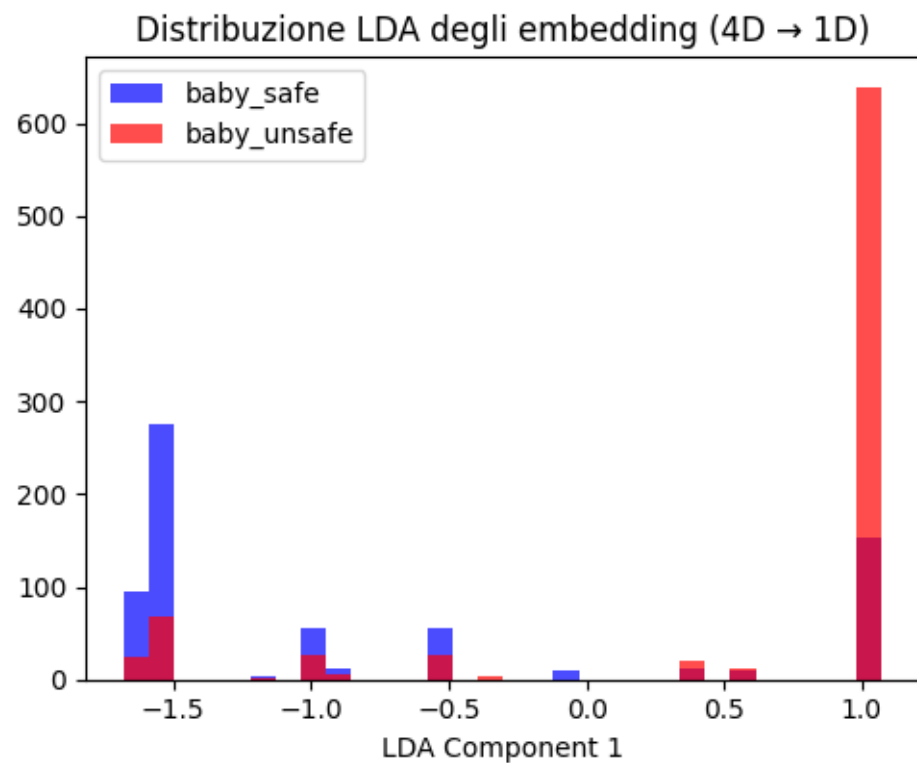
Silhouette

score-----

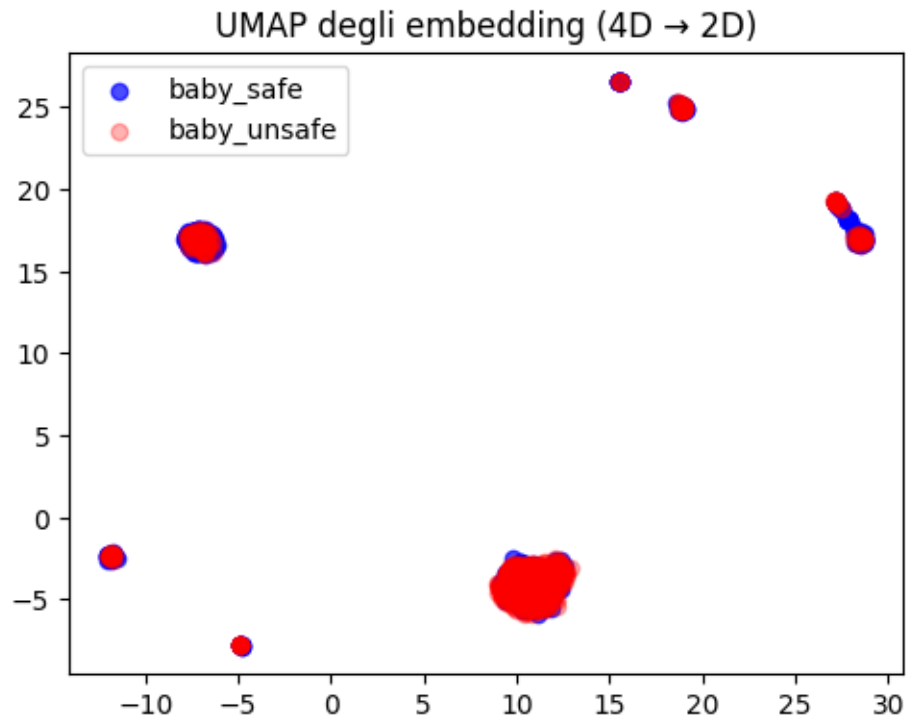


Embeddings

distributions-----



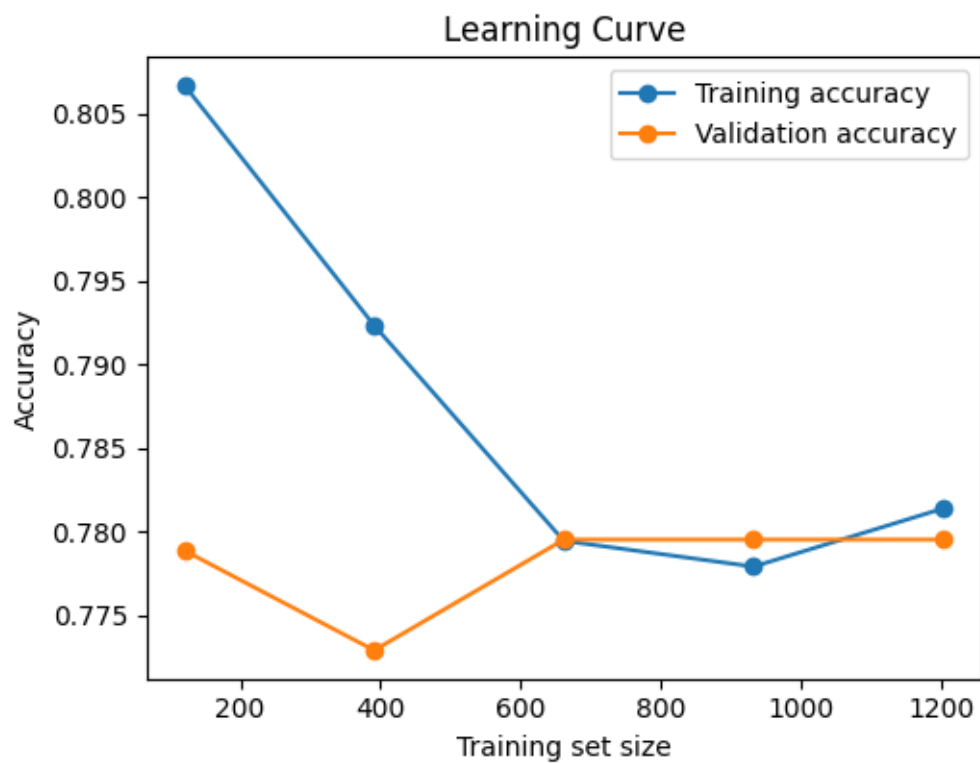
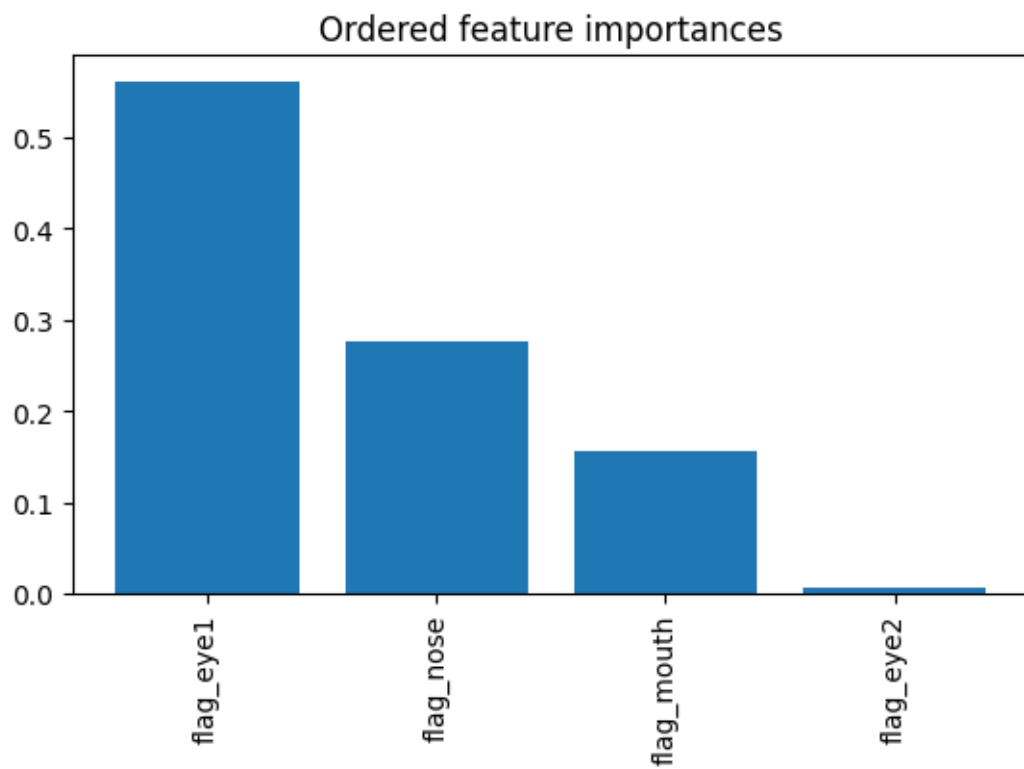
```
/home/terra/anaconda3/envs/SIDS_revelation_project/lib/python3.10/site-
packages/umap/umap_.py:1952: UserWarning: n_jobs value 1 overridden to 1 by
setting random_state. Use no seed for parallelism.
warn(
```



1.5 4. Train model classification

```
[8]: clf = Classifier(embeddings, emb_builder.y, emb_builder.classes_bs)
      clf.random_forest()
```

```
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```

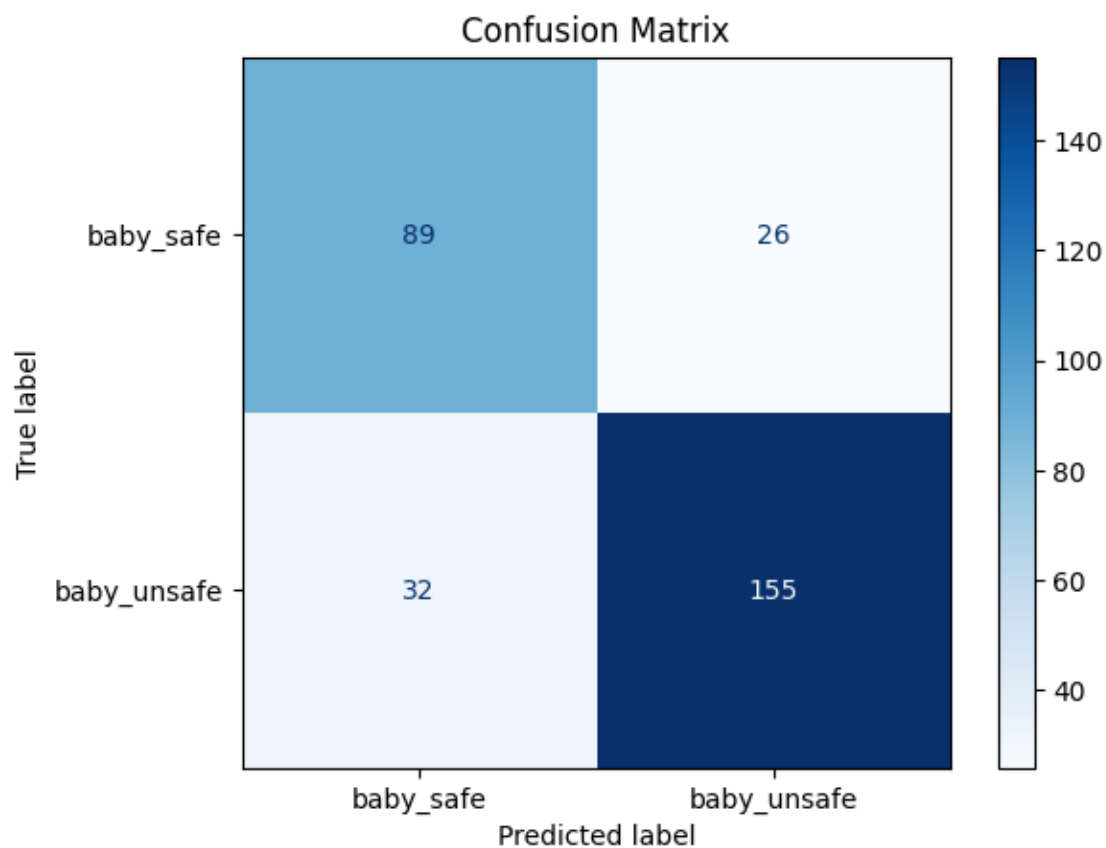


Dataset labels:-----
{'baby_safe': 0, 'baby_unsafe': 1}

Report-----

	precision	recall	f1-score	support
baby_safe	0.74	0.77	0.75	115
baby_unsafe	0.86	0.83	0.84	187
accuracy			0.81	302
macro avg	0.80	0.80	0.80	302
weighted avg	0.81	0.81	0.81	302

Confusion matrix-----



1.6 2. Extract embeddings from dataset

Create embeddings

```
[9]: embeddings = emb_builder.embedding_all_features()
```

Creation of all features

embedding-----

1506 embedding created

```
[10]: embeddings.head()
```

```
[10]:
```

	flag_eye1	flag_eye2	flag_nose	flag_mouth	x_eye1	y_eye1	x_eye2	\
0	0	0	0	0	-1.000000	-1.000000	-1.000000	
1	1	1	1	1	0.747866	0.955937	0.746319	
2	0	0	0	0	-1.000000	-1.000000	-1.000000	
3	1	1	1	1	0.533123	0.143157	0.374687	
4	1	1	1	1	0.859706	0.598094	0.846316	

	y_eye2	x_nose	y_nose	x_mouth	y_mouth	eye_distance	\
0	-1.000000	-1.000000	-1.000000	-1.000000	-1.000000	-1.000000	
1	0.680452	0.709361	0.853581	0.650598	0.840860	0.275490	
2	-1.000000	-1.000000	-1.000000	-1.000000	-1.000000	-1.000000	
3	0.148591	0.457184	0.163462	0.469082	0.201713	0.158529	
4	0.728694	0.827010	0.658031	0.788980	0.648129	0.131284	

	face_vertical_length	face_angle_vertical	face_angle_horizontal	\
0	-1.000000	-1.000000	-1.000000	
1	0.060125	122.830696	147.334481	
2	-1.000000	-1.000000	-1.000000	
3	0.040059	87.692268	154.811363	
4	0.039298	104.018161	136.106213	

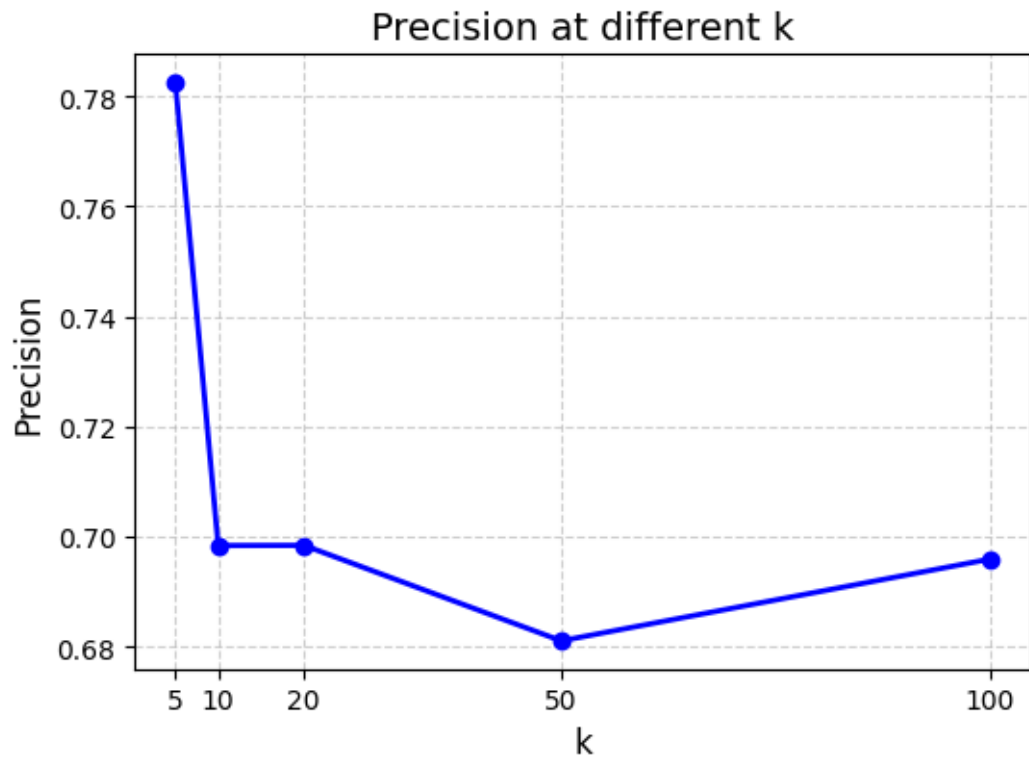
	symmetry_diff
0	0.000000
1	0.085138
2	0.000000
3	0.004185
4	0.002724

1.7 3. Retrieval to evaluate embedding goodness

```
[11]: ret = ImageRetrieval( embeddings, emb_builder.y, emb_builder.image_paths,
    ↪ image_dataset_path, emb_builder.classes_bs)
ret.report("euclidean")
```

Precision at different

k:-----



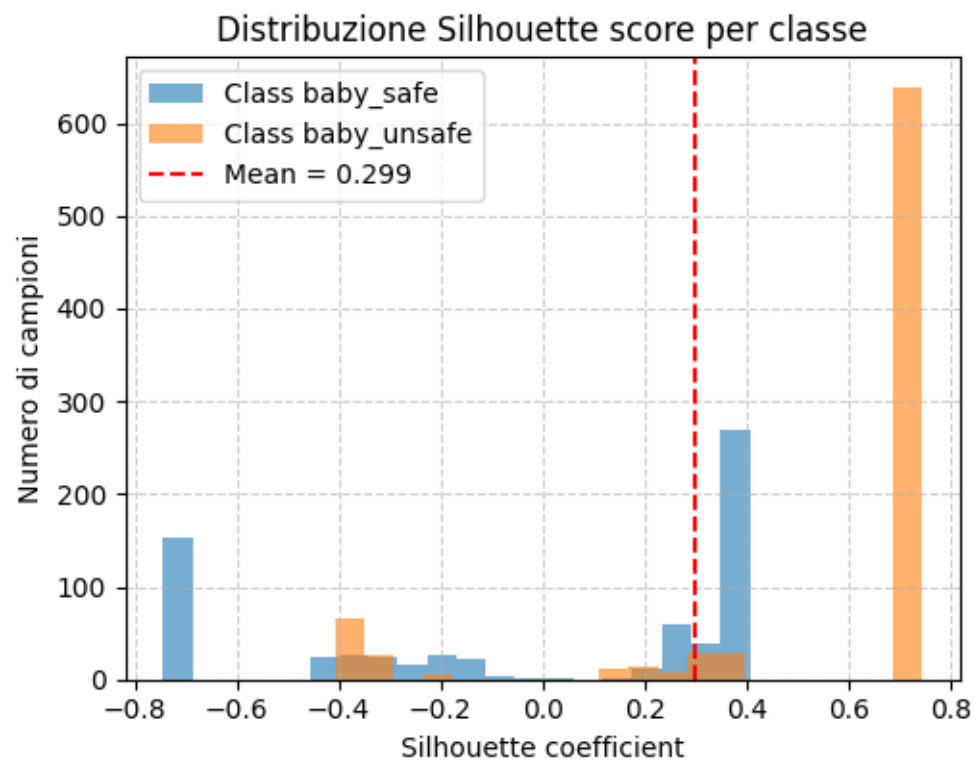
Recall at

R:-----

0.6126016574337287

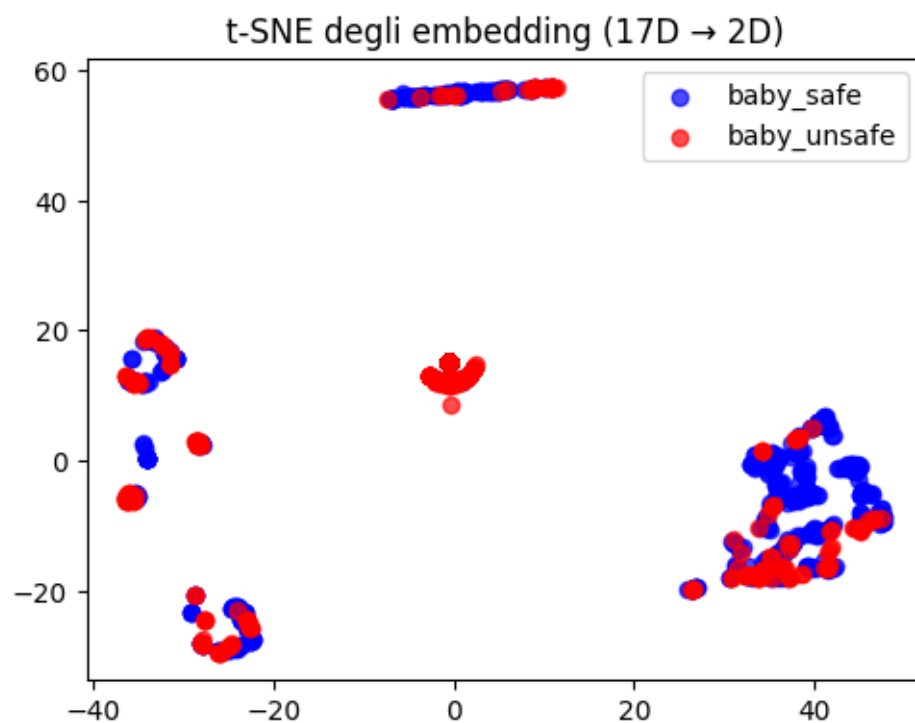
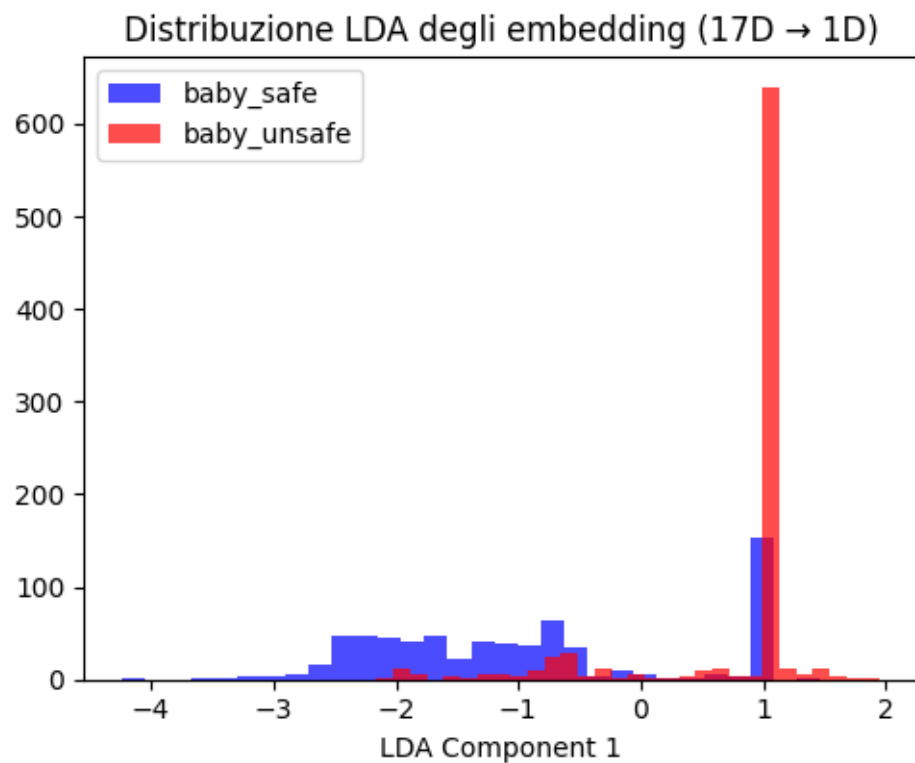
Silhouette

score-----



Embeddings

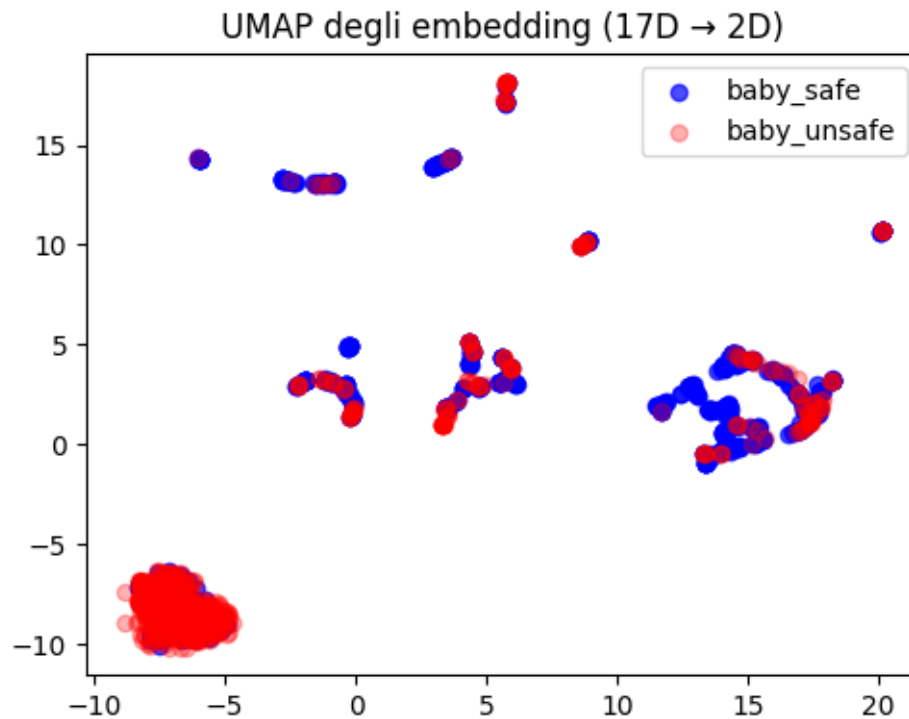
distributions-----



```

/home/terra/anaconda3/envs/SIDS_revelation_project/lib/python3.10/site-
packages/umap/umap_.py:1952: UserWarning: n_jobs value 1 overridden to 1 by
setting random_state. Use no seed for parallelism.
  warn(
/home/terra/anaconda3/envs/SIDS_revelation_project/lib/python3.10/site-
packages/sklearn/manifold/_spectral_embedding.py:328: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(

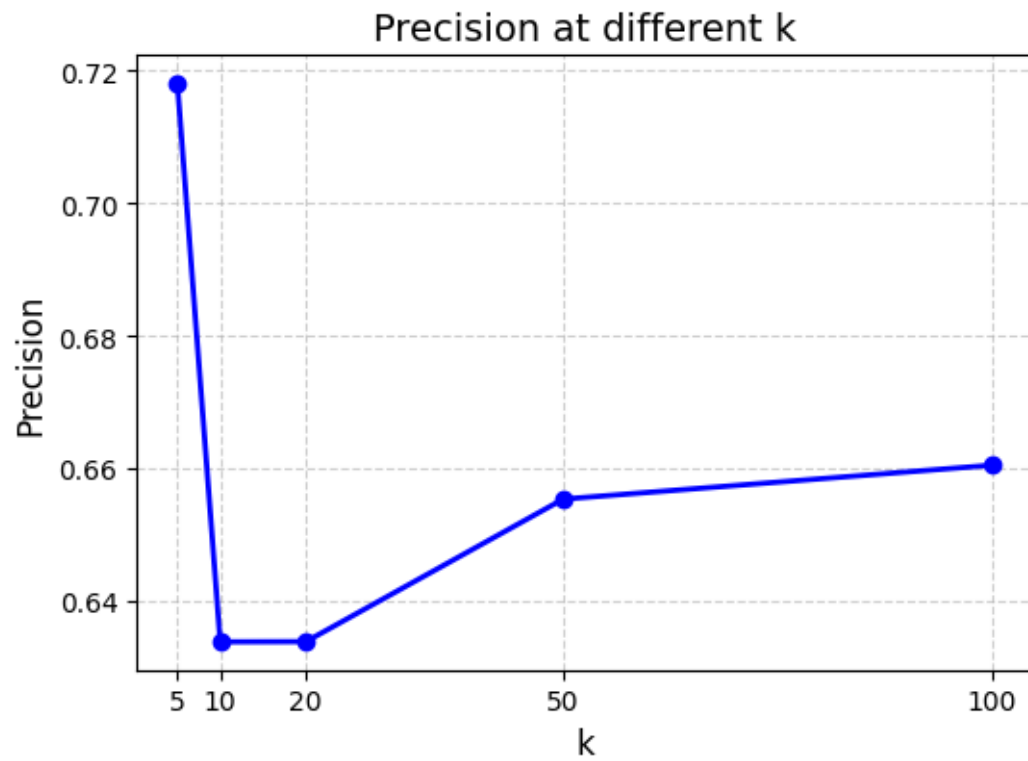
```



```
[12]: ret.report("cosine")
```

Precision at different

k:-----

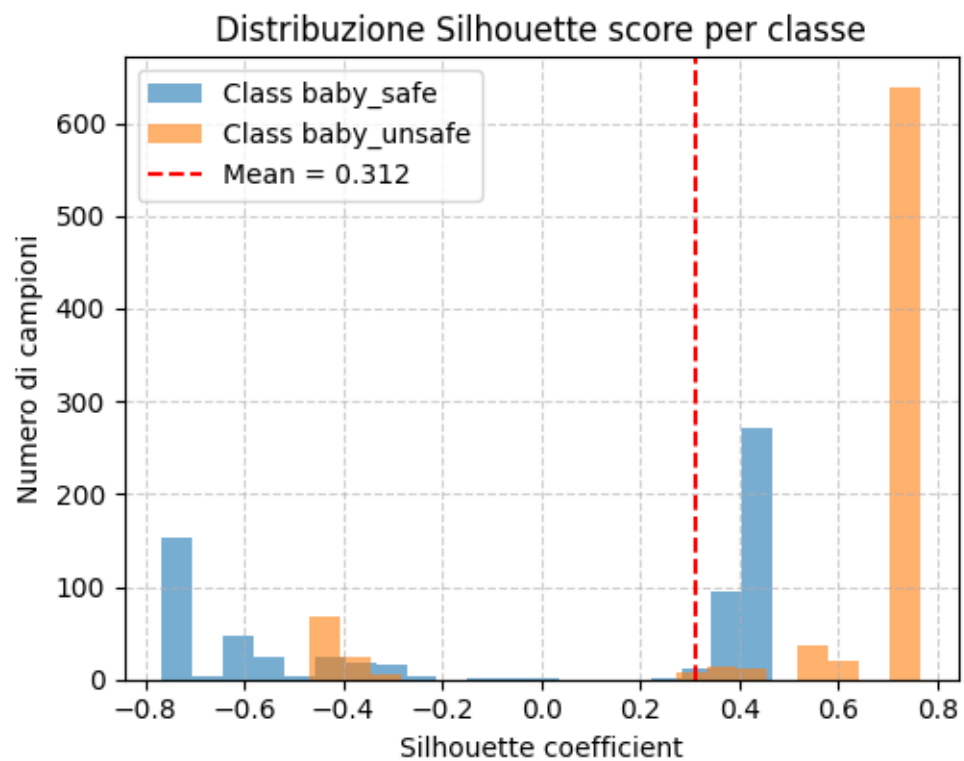


Recall at

R-----
0.6158959434877811

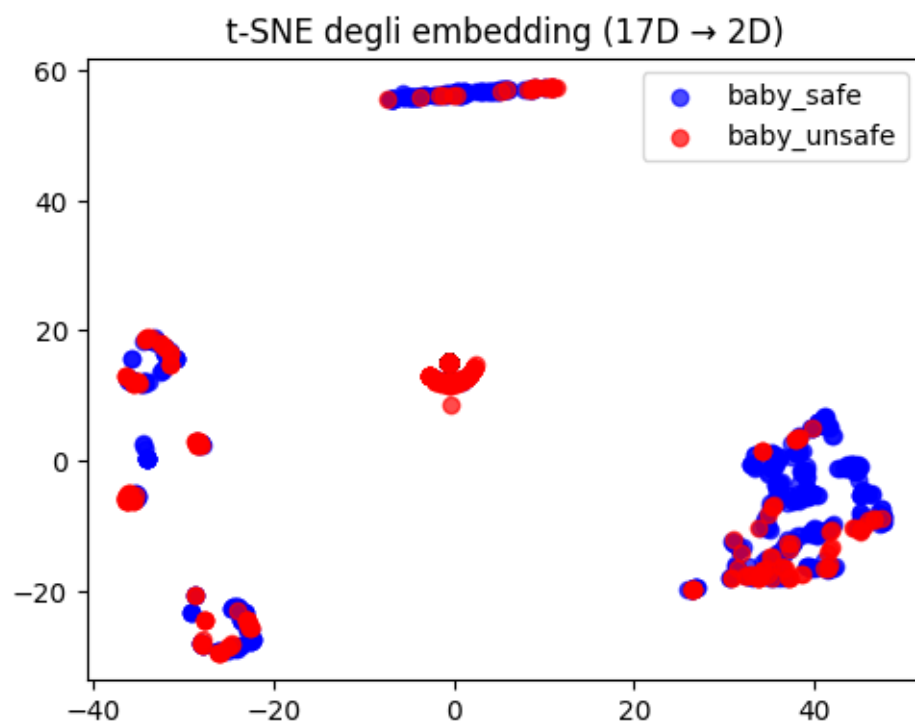
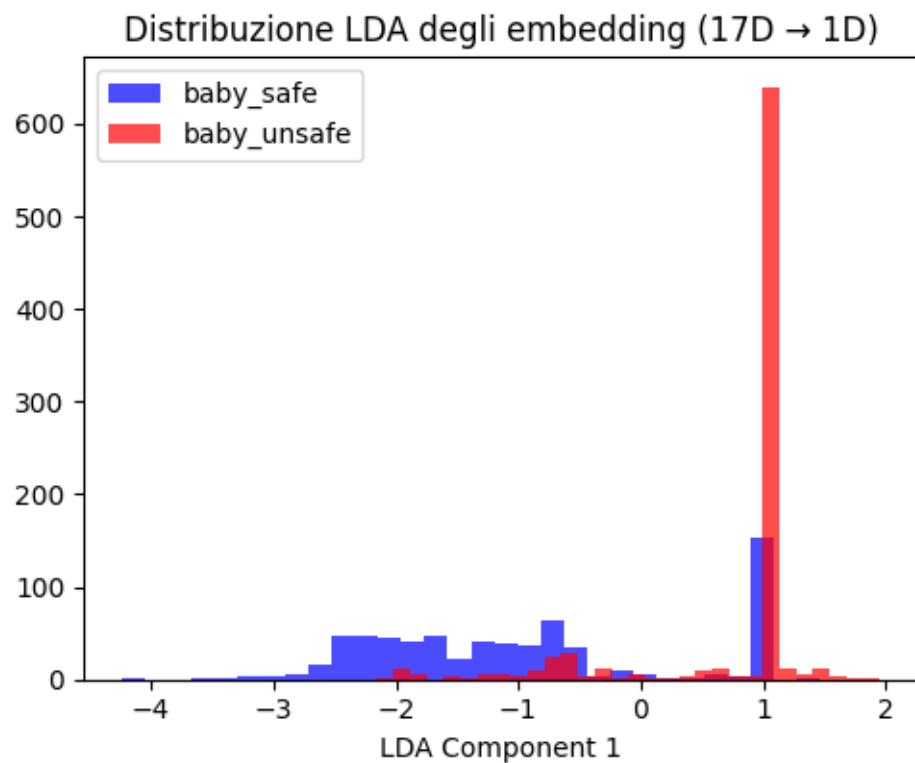
Silhouette

score-----



Embeddings

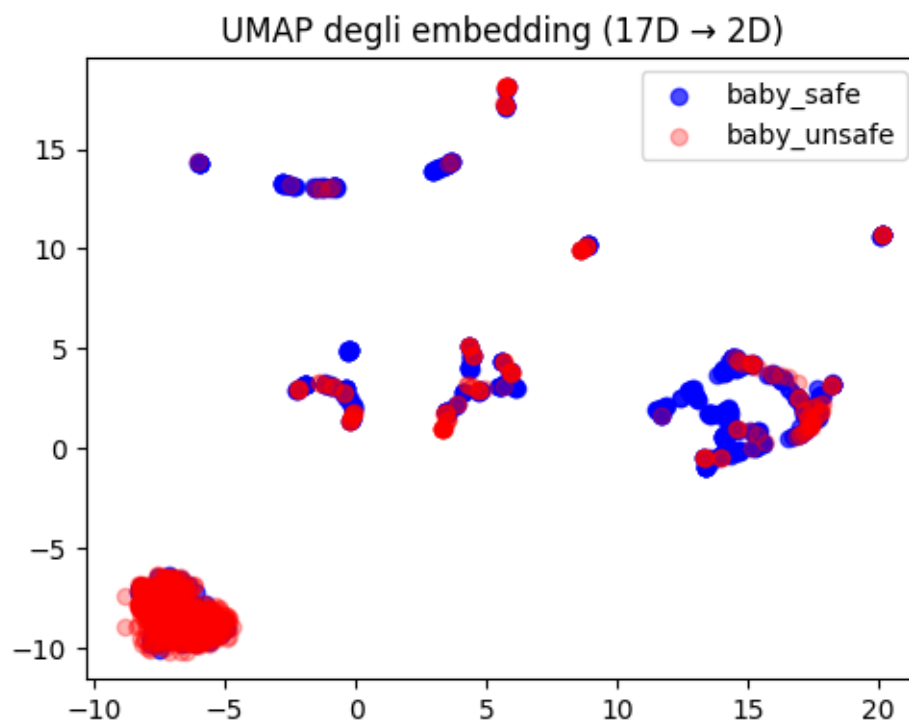
distributions-----




```

/home/terra/anaconda3/envs/SIDS_revelation_project/lib/python3.10/site-
packages/umap/umap_.py:1952: UserWarning: n_jobs value 1 overridden to 1 by
setting random_state. Use no seed for parallelism.
  warn(
/home/terra/anaconda3/envs/SIDS_revelation_project/lib/python3.10/site-
packages/sklearn/manifold/_spectral_embedding.py:328: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(

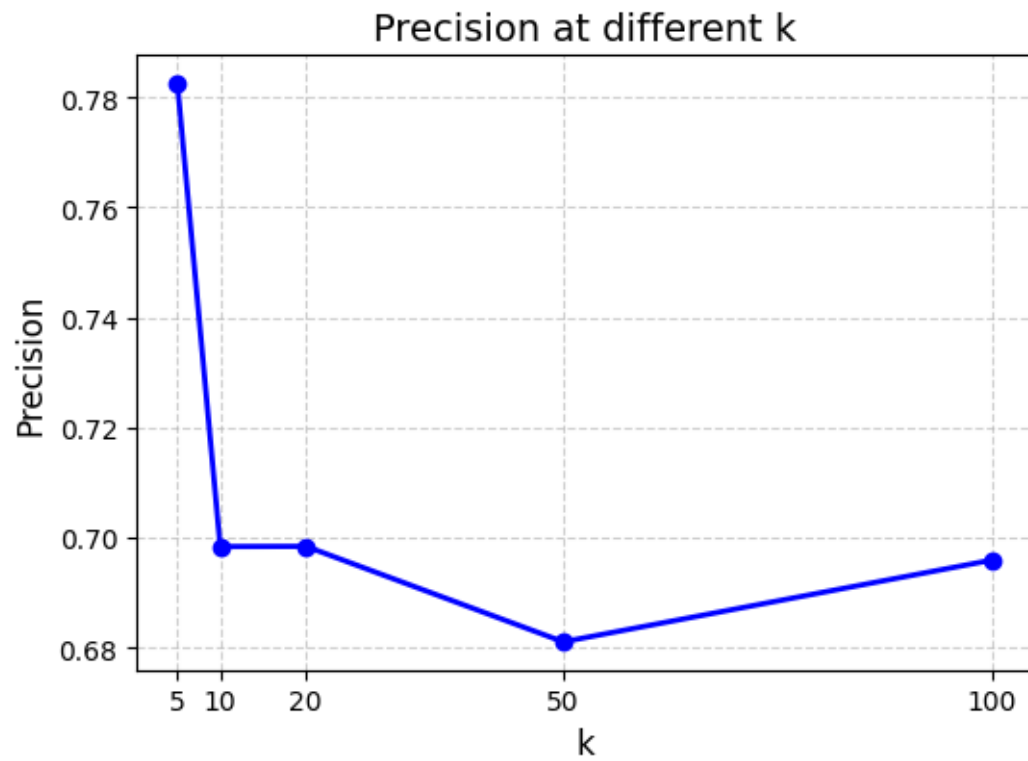
```



```
[13]: ret.report("minkowski")
```

Precision at different

k:-----

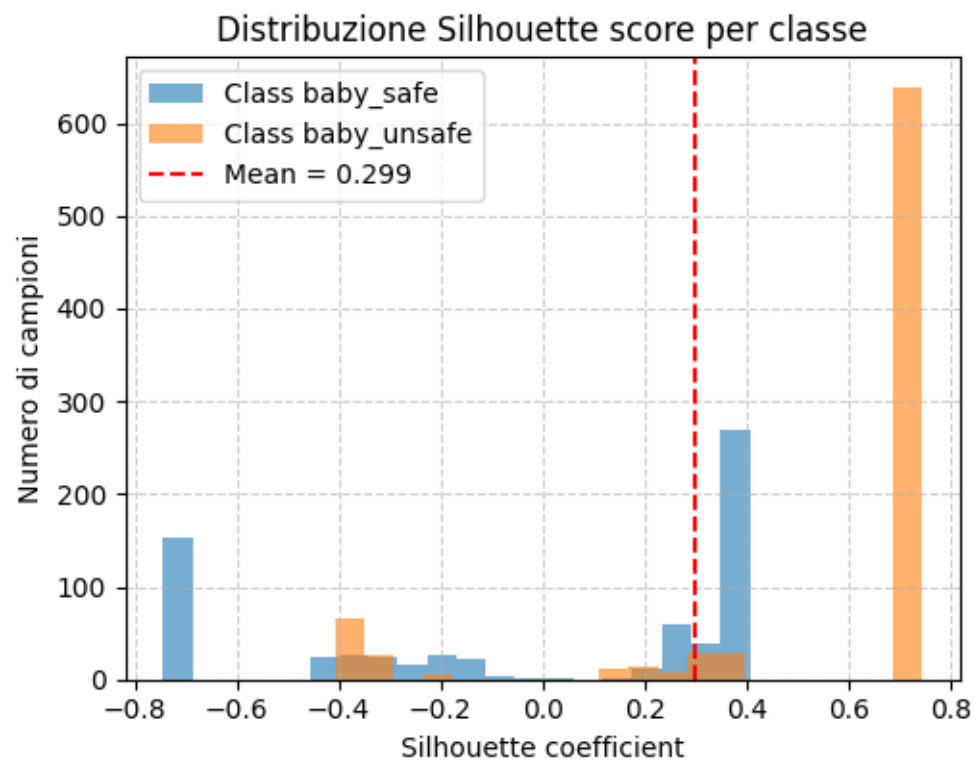


Recall at

R-----
0.6126016574337287

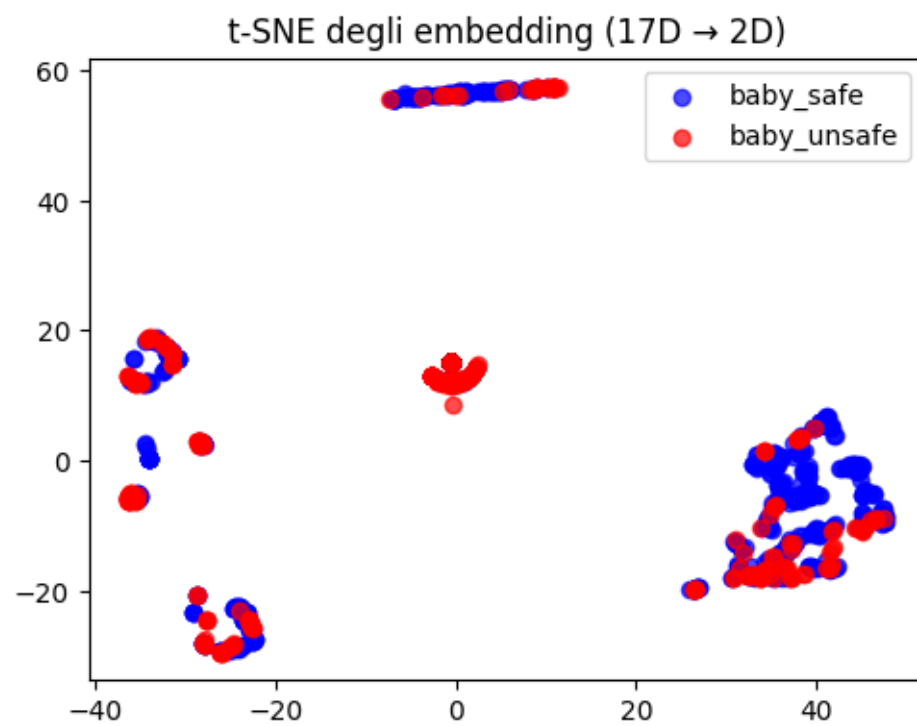
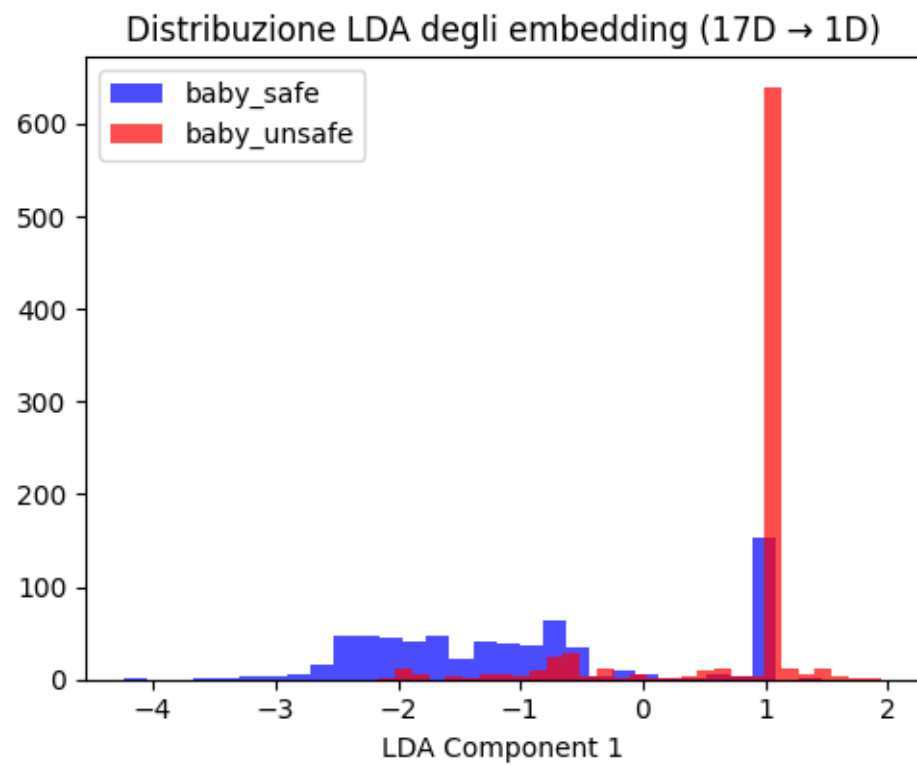
Silhouette

score-----



Embeddings

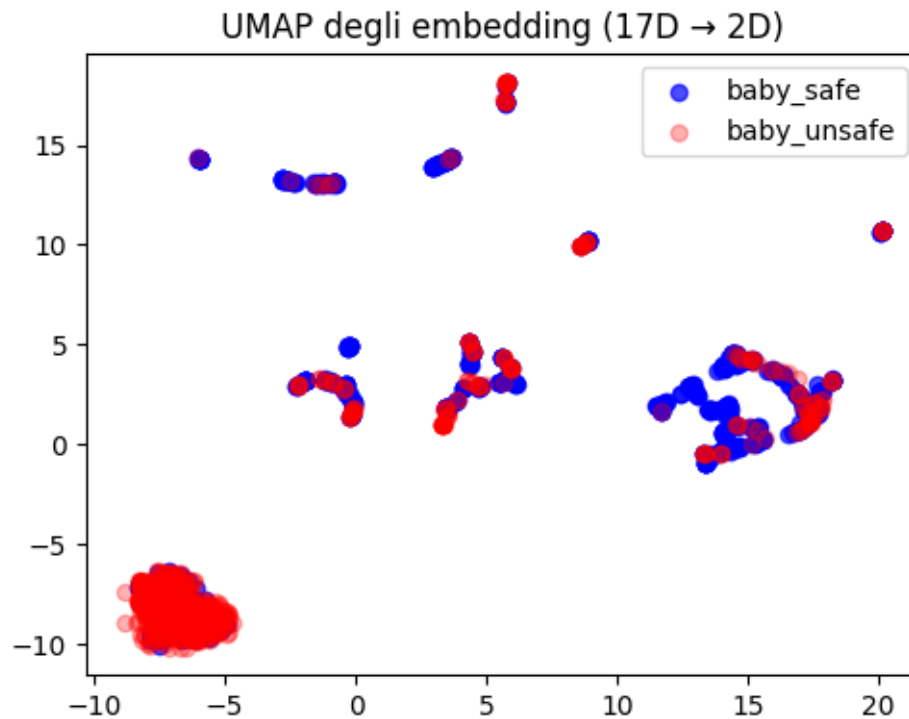
distributions-----



```

/home/terra/anaconda3/envs/SIDS_revelation_project/lib/python3.10/site-
packages/umap/umap_.py:1952: UserWarning: n_jobs value 1 overridden to 1 by
setting random_state. Use no seed for parallelism.
    warn(
/home/terra/anaconda3/envs/SIDS_revelation_project/lib/python3.10/site-
packages/sklearn/manifold/_spectral_embedding.py:328: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
    warnings.warn(

```



1.8 4. Train model classification

```

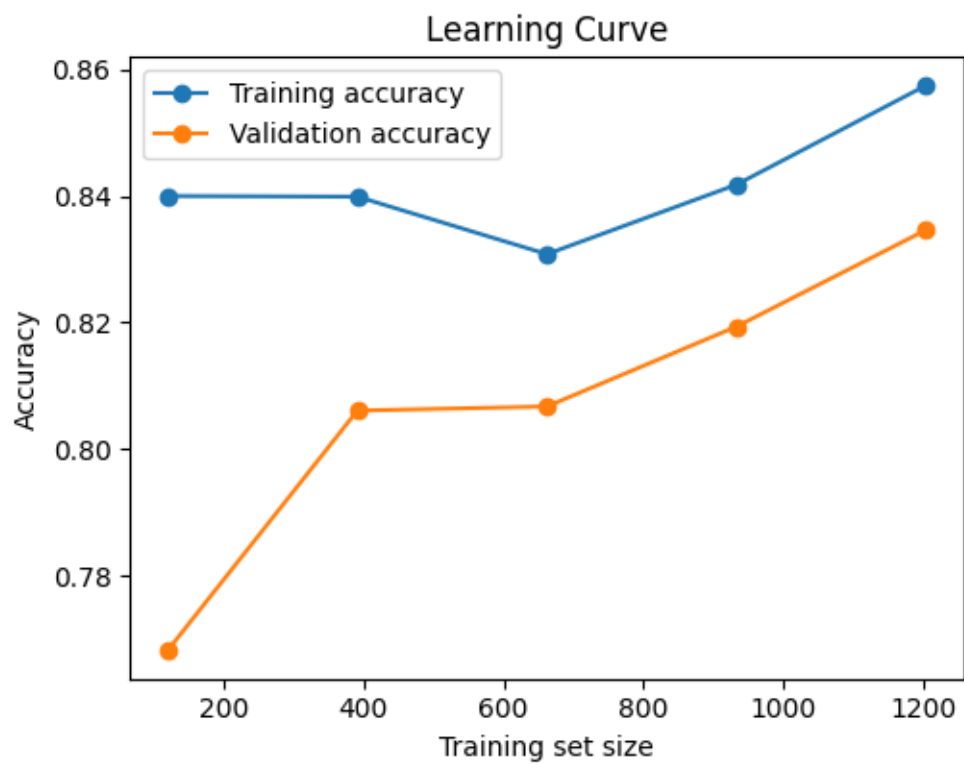
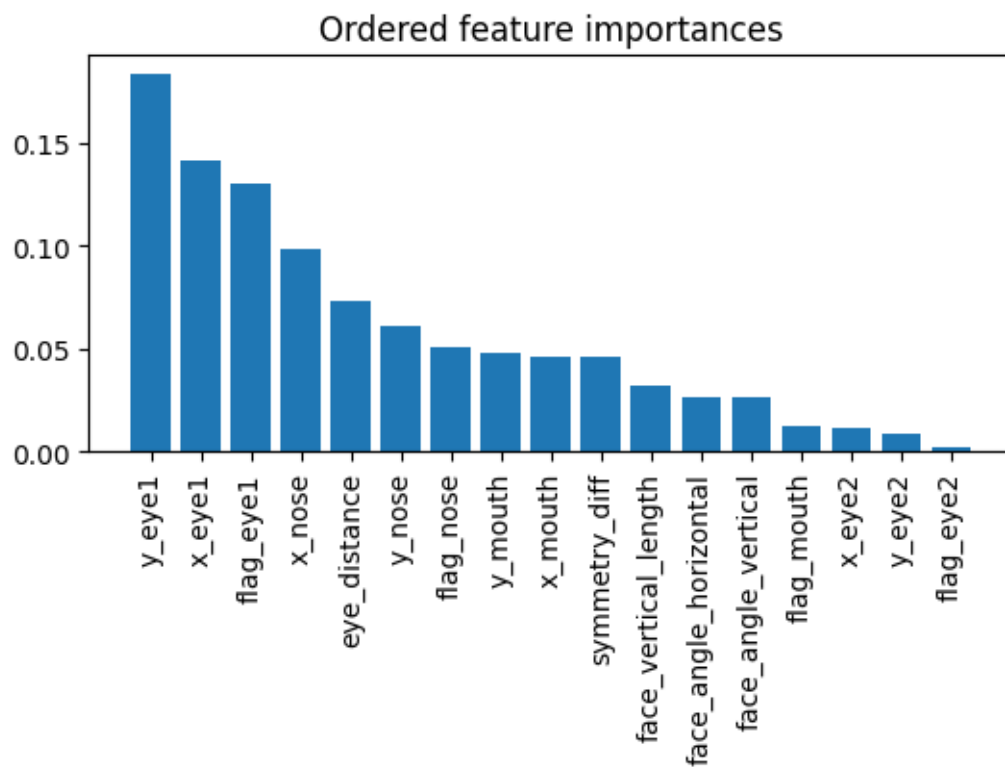
[14]: clf = Classifier(embeddings, emb_builder.y, emb_builder.classes_bs)
      clf.random_forest()

```

```

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```

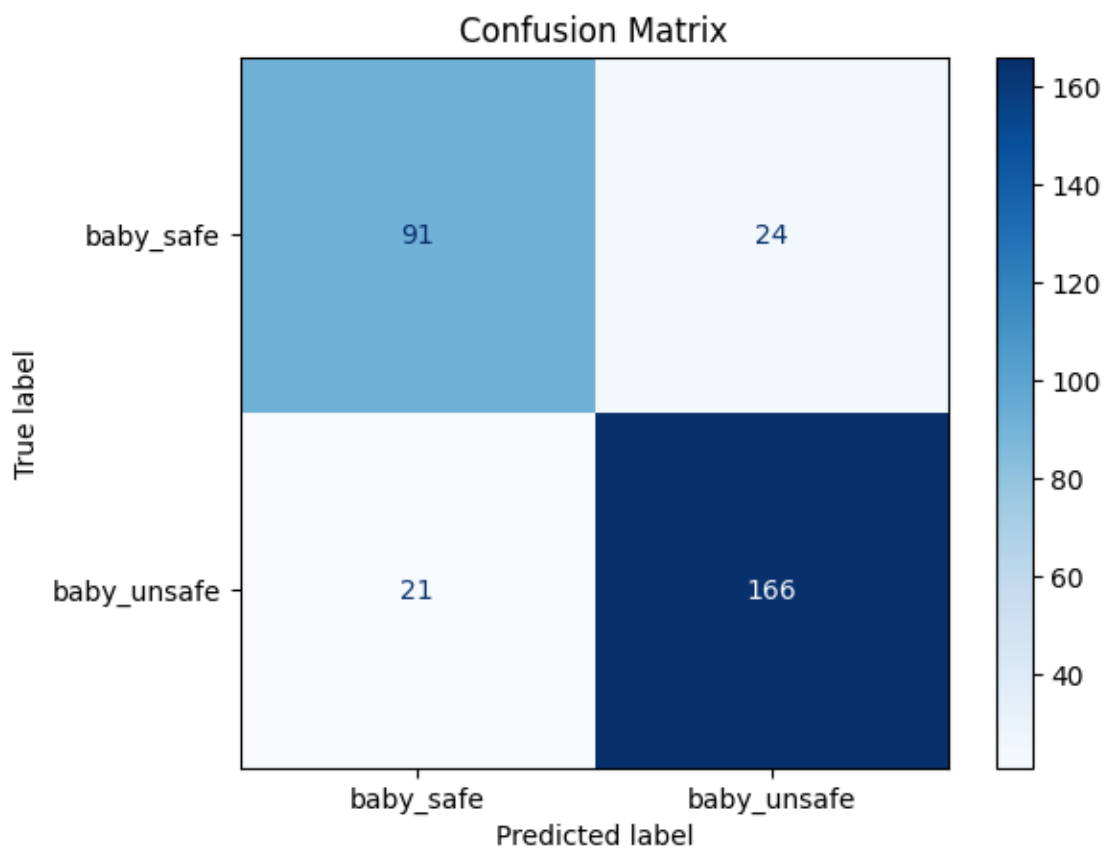


Dataset labels:-----
{'baby_safe': 0, 'baby_unsafe': 1}

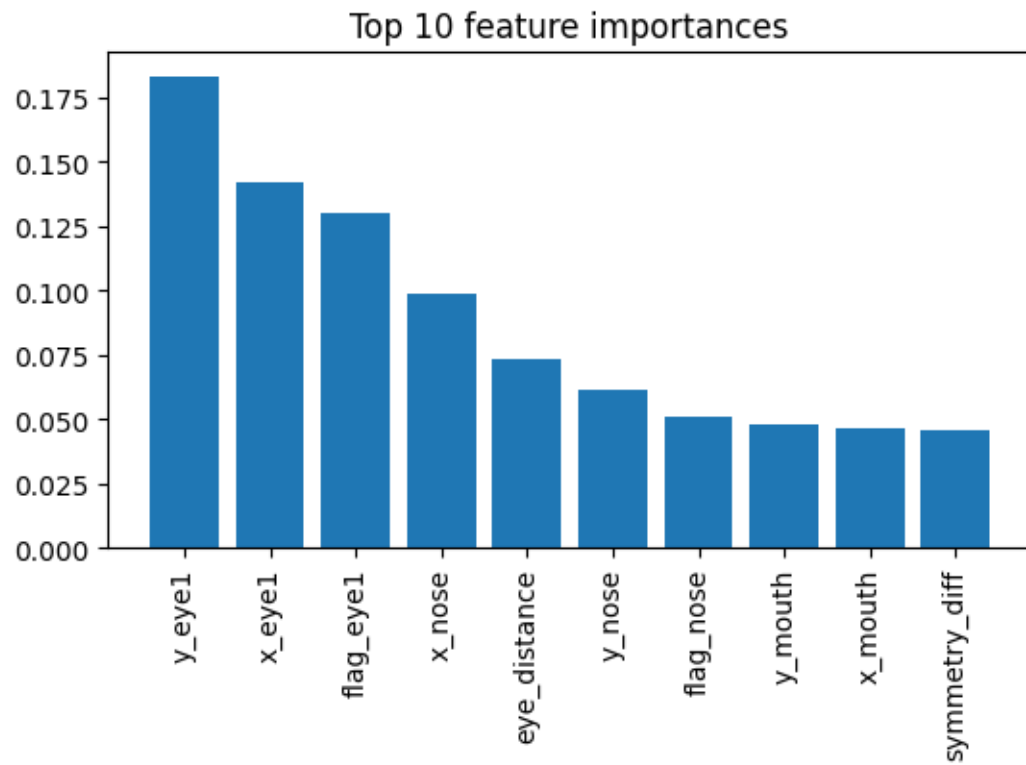
Report-----

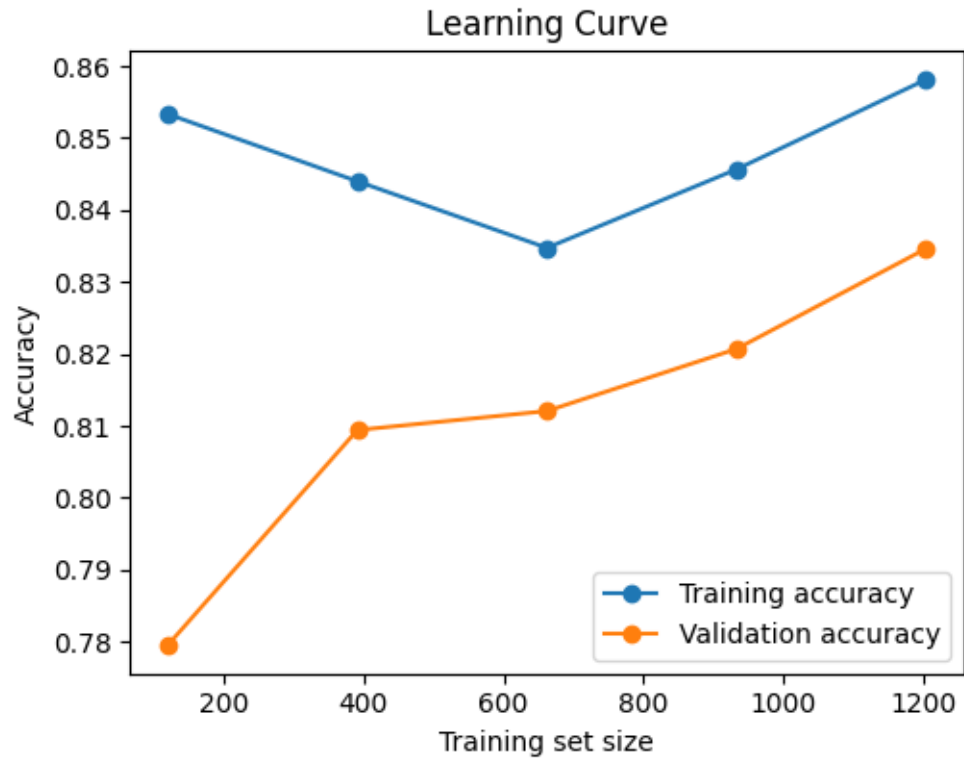
	precision	recall	f1-score	support
baby_safe	0.81	0.79	0.80	115
baby_unsafe	0.87	0.89	0.88	187
accuracy			0.85	302
macro avg	0.84	0.84	0.84	302
weighted avg	0.85	0.85	0.85	302

Confusion matrix-----



-----TOP 10 FEATURES
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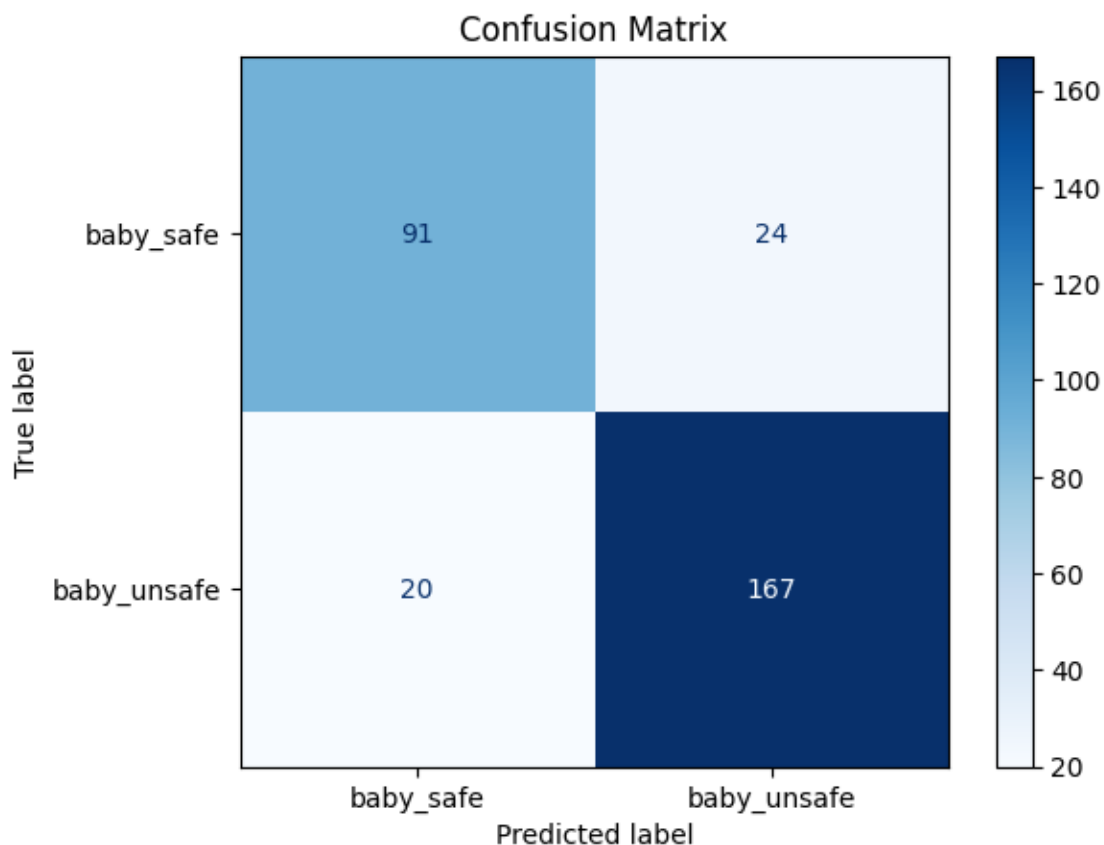
Dataset labels:-----

```
{'baby_safe': 0, 'baby_unsafe': 1}
```

Report-----

	precision	recall	f1-score	support
baby_safe	0.82	0.79	0.81	115
baby_unsafe	0.87	0.89	0.88	187
accuracy			0.85	302
macro avg	0.85	0.84	0.84	302
weighted avg	0.85	0.85	0.85	302

Confusion matrix-----



1.9 2. Extract embeddings from dataset

Create embeddings

```
[15]: embeddings = emb_builder.embedding_all_features_norm()
```

Creation of all features

embedding-----

FINISHED: 1506 embedding created

```
[16]: embeddings.head()
```

```
[16]:   flag_eye1  flag_eye2  flag_nose  flag_mouth   x_eye1   y_eye1   x_eye2  \
0         0         0         0         0 -1.000000 -1.000000 -1.000000
1         1         1         1         1  0.747866  0.955937  0.746319
2         0         0         0         0 -1.000000 -1.000000 -1.000000
3         1         1         1         1  0.533123  0.143157  0.374687
4         1         1         1         1  0.859706  0.598094  0.846316
```

	y_eye2	x_nose	y_nose	...	x_mouth_norm	y_mouth_norm	\
0	-1.000000	-1.000000	-1.000000	...	-7.346064	-3.560087	
1	0.680452	0.709361	0.853581	...	0.814792	1.258686	
2	-1.000000	-1.000000	-1.000000	...	-1.126908	-1.251334	
3	0.148591	0.457184	0.163462	...	1.014402	1.536646	
4	0.728694	0.827010	0.658031	...	0.911147	0.985550	

	eye_distance	eye_distance_norm	face_vertical_length	\
0	-1.000000	-1.000000	-1.000000	
1	0.275490	0.685025	0.060125	
2	-1.000000	-1.000000	-1.000000	
3	0.158529	0.397289	0.040059	
4	0.131284	0.496643	0.039298	

	face_vertical_length_norm	face_angle_vertical	face_angle_horizontal	\
0	-1.000000	-1.000000	-1.000000	
1	0.090708	122.830696	147.334481	
2	-1.000000	-1.000000	-1.000000	
3	0.172300	87.692268	154.811363	
4	0.121073	104.018161	136.106213	

	symmetry_diff	head_ration
0	0.000000	1.292278
1	0.085138	1.648186
2	0.000000	1.728926
3	0.004185	0.582652
4	0.002724	1.227875

[5 rows x 28 columns]

1.10 3. Retrieval to evaluate embedding goodness

```
[ ]: ret = ImageRetrieval( embeddings, emb_builder.y, emb_builder.image_paths,
    ↪ image_dataset_path, emb_builder.classes_bs)
ret.report("euclidean")
```

```
[ ]: ret.report("cosine")
```

```
[ ]: ret.report("minkowski")
```

1.11 4. Train model classification

```
[ ]: clf = Classifier(embeddings, emb_builder.y, emb_builder.classes_bs)
clf.random_forest()
```

```
[ ]: clf.XGBC()
```

```
[ ]: clf.logistic_regression()
```

1.12 Save notebook

```
[ ]: file_manager.save_as_pdf(ipynbname.path())
```