

## best\_classifiers

September 10, 2025

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

      import ipynbname
      import torch
      import os
      import warnings
      import numpy as np
      from xgboost import XGBClassifier
      from libraries.classifier_utils import *
      from libraries.embeddings_utils import *
      from libraries.EmbeddingNet import *
      from libraries.retrieval_utils import *

      project_dir = f"{os.getcwd()}.
      ↪split('SIDS_revelation_project')[0]}SIDS_revelation_project/"
      image_dataset_path = f"{project_dir}datasets/onback_onstomach_v3"
      model_path_fd = f"{project_dir}/models/4.fd_weights/best.pt"
      model_path_pe = f"{project_dir}/models/2.pe_weights/best.pt"

      if torch.cuda.is_available():
          device = torch.device("cuda")
      elif torch.backends.mps.is_available():
          device = torch.device("mps")
      else:
          device = torch.device("cpu")
      print(f"Using device: {device}")
```

The autoreload extension is already loaded. To reload it, use:

```
%reload_ext autoreload
```

Using device: cpu

```
[74]: emb_builder = EmbeddingBuilder(model_path_fd, image_dataset_path,
      ↪"load", weights_path_pe=model_path_pe)
```

Extracting dataset info from .coco.json

```

file:-----
Dataset contains 4158 valid samples, and labels are {'baby_on_back': 1,
'baby_on_stomach': 2}
-----

Loading features from
.csv-----
Features loaded succesfully, in particular there are 4158 files in the dataset
-----

Embedding builder initialized
successfully-----
Face detection model: 4 (YOLOv8)
Dataset: /home/terra/Desktop/unimore/AI_engineering/SIDS_revelation_project/data
sets/onback_onstomach_v3
Dataset dimension: 4158
Dataset labels: {'baby_safe': 0, 'baby_unsafe': 1}
-----

```

## 0.1 Best model with approach Features selection: xgbc

Best parameters: {'colsample\_bytree': np.float64(0.9942601816442402), 'gamma': np.float64(0.1210276357557502), 'learning\_rate': np.float64(0.21164066422176356), 'max\_depth': 5, 'n\_estimators': 283, 'subsample': np.float64(0.6950550175969599)} - Best embeddings: original - {'face\_vertical\_length\_norm', 'flag\_mouth', 'flag\_eye1', 'face\_angle\_horizontal', 'flag\_nose', 'y\_eye2', 'x\_left\_elbow', 'x\_left\_hip', 'x\_eye2', 'nose\_hip\_right', 'flag\_eye2'} - Performance: 0.90 with all feature, 0.89 with 25\_top\_features, 0.84 with top\_10\_features

```

[75]: embeddings = emb_builder.create_embedding(flags=True,positions=True,
    ↪positions_normalized=True, geometric_info=True,k_positions_normalized=True,
    ↪,k_geometric_info=True)

```

```

Embedding
creation-----
Features: ['flag_eye1', 'flag_eye2', 'flag_nose', 'flag_mouth', 'x_eye1',
'y_eye1', 'x_eye2', 'y_eye2', 'x_nose', 'y_nose', 'x_mouth', 'y_mouth',
'x_eye1_norm', 'y_eye1_norm', 'x_eye2_norm', 'y_eye2_norm', 'x_nose_norm',
'y_nose_norm', 'x_mouth_norm', 'y_mouth_norm', 'eye_distance',
'eye_distance_norm', 'face_vertical_length', 'face_vertical_length_norm',
'face_angle_vertical', 'face_angle_horizontal', 'symmetry_diff', 'head_ration',
'x_nose_k', 'y_nose_k', 'x_left_eye_k', 'y_left_eye_k', 'x_right_eye_k',
'y_right_eye_k', 'x_left_ear', 'y_left_ear', 'x_right_ear', 'y_right_ear',
'x_left_shoulder', 'y_left_shoulder', 'x_right_shoulder', 'y_right_shoulder',
'x_left_elbow', 'y_left_elbow', 'x_right_elbow', 'y_right_elbow',

```

```
'x_left_wrist', 'y_left_wrist', 'x_right_wrist', 'y_right_wrist', 'x_left_hip',
'y_left_hip', 'x_right_hip', 'y_right_hip', 'x_left_knee', 'y_left_knee',
'x_right_knee', 'y_right_knee', 'x_left_ankle', 'y_left_ankle', 'x_right_ankle',
'y_right_ankle', 'shoulders_dist', 'shoulder_hip_right_dist',
'shoulder_hip_left_dist', 'nose_shoulder_right', 'nose_shoulder_left',
'shoulder_left_knee_right', 'shoulder_right_knee_left', 'knee_ankle_right',
'knee_ankle_left', 'nose_hip_right', 'nose_hip_left',
'elbow_shoulder_hip_right', 'elbow_shoulder_hip_left',
'shoulder_elbow_wrist_right', 'shoulder_elbow_wrist_left',
'shoulder_hip_knee_right', 'shoulder_hip_knee_left', 'hip_knee_ankle_right',
'hip_knee_ankle_left', 'shoulders_line_inclination', 'hips_line_inclination',
'torsion']
```

FINISHED: 4158 embedding created

```
[76]: features_to_drop = ['face_vertical_length_norm', 'flag_mouth', 'flag_eye1',
    ↳ 'face_angle_horizontal', 'flag_nose', 'y_eye2', 'x_left_elbow',
    ↳ 'x_left_hip', 'x_eye2', 'nose_hip_right', 'flag_eye2']
embeddings = embeddings.drop(columns=features_to_drop)
clf = Classifier(embeddings, emb_builder.y, emb_builder.classes_bs,
    ↳ image_paths=emb_builder.image_paths)

best_params = {
    'colsample_bytree': np.float64(0.9942601816442402),
    'gamma': np.float64(0.1210276357557502),
    'learning_rate': np.float64(0.21164066422176356),
    'max_depth': 5,
    'n_estimators': 283,
    'subsample': np.float64(0.6950550175969599),
    'use_label_encoder': False, # necessario con sklearn>=1.0
    'eval_metric': 'logloss'
}
model = XGBClassifier(**best_params)

with warnings.catch_warnings():
    warnings.simplefilter("ignore")
    results = clf.evaluation_pipeline_save_misclassified(model)
```

-----FIRST  
ANALYSIS-----

/home/terra/anaconda3/envs/SIDS\_project/lib/python3.10/site-  
packages/xgboost/core.py:377: FutureWarning: Your system has an old version of

glibc (< 2.28). We will stop supporting Linux distros with glibc older than 2.28 after **\*\*May 31, 2025\*\***. Please upgrade to a recent Linux distro (with glibc >= 2.28) to use future versions of XGBoost.

Note: You have installed the 'manylinux2014' variant of XGBoost. Certain features such as GPU algorithms or federated learning are not available. To use these features, please upgrade to a recent Linux distro with glibc 2.28+, and install the 'manylinux\_2\_28' variant.

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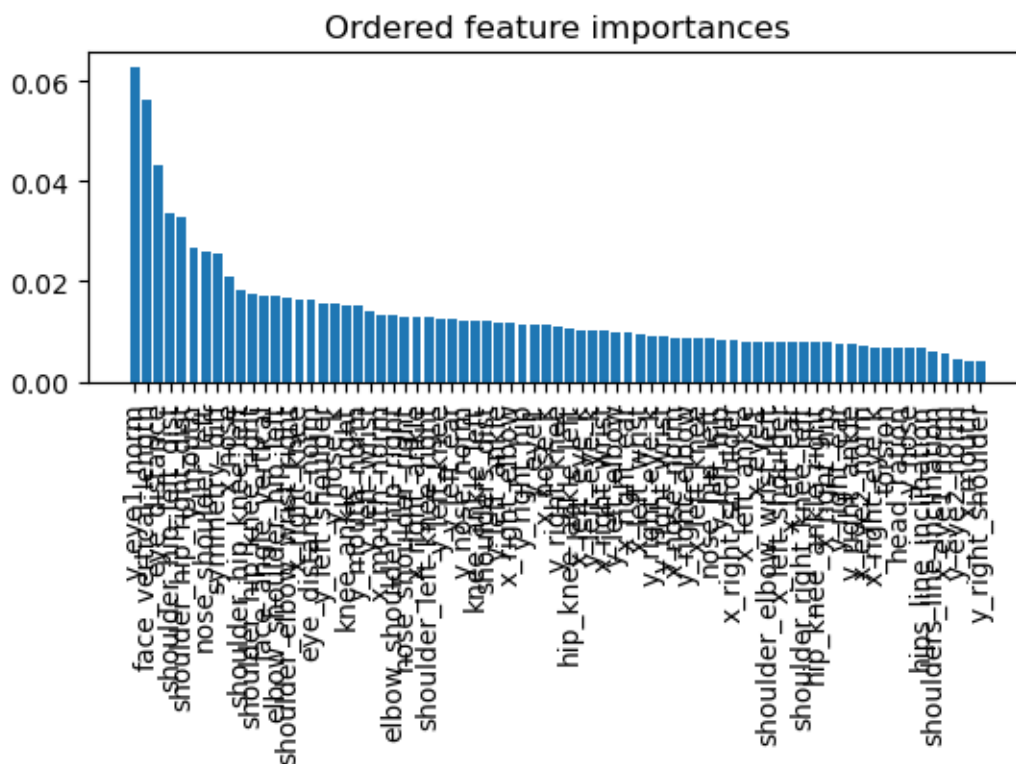
```

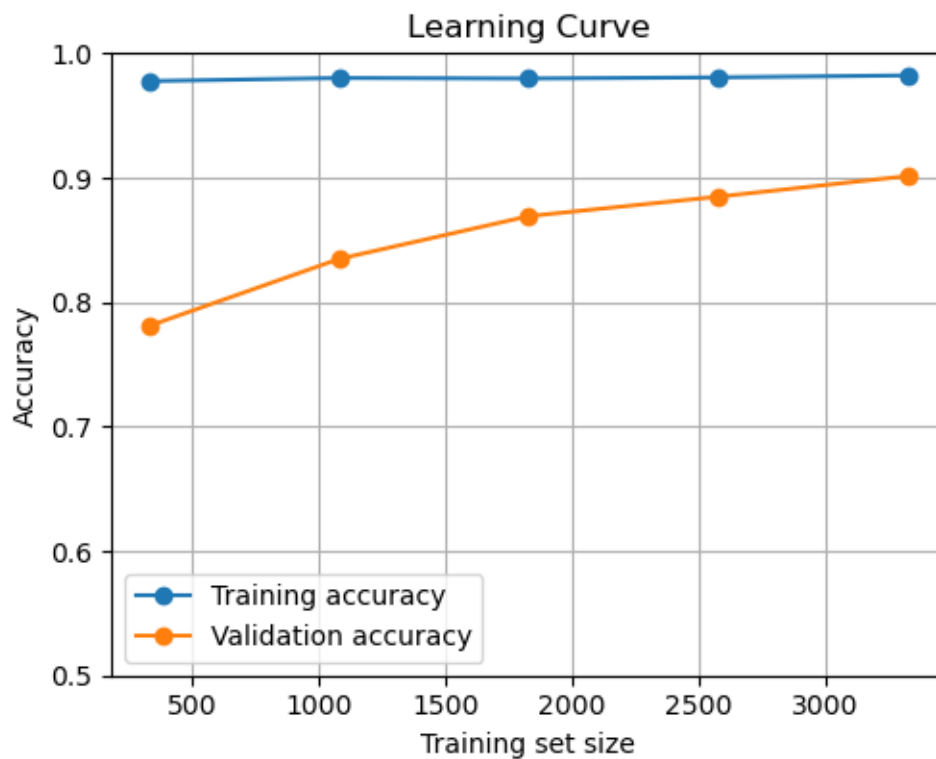
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warnings.warn(
```



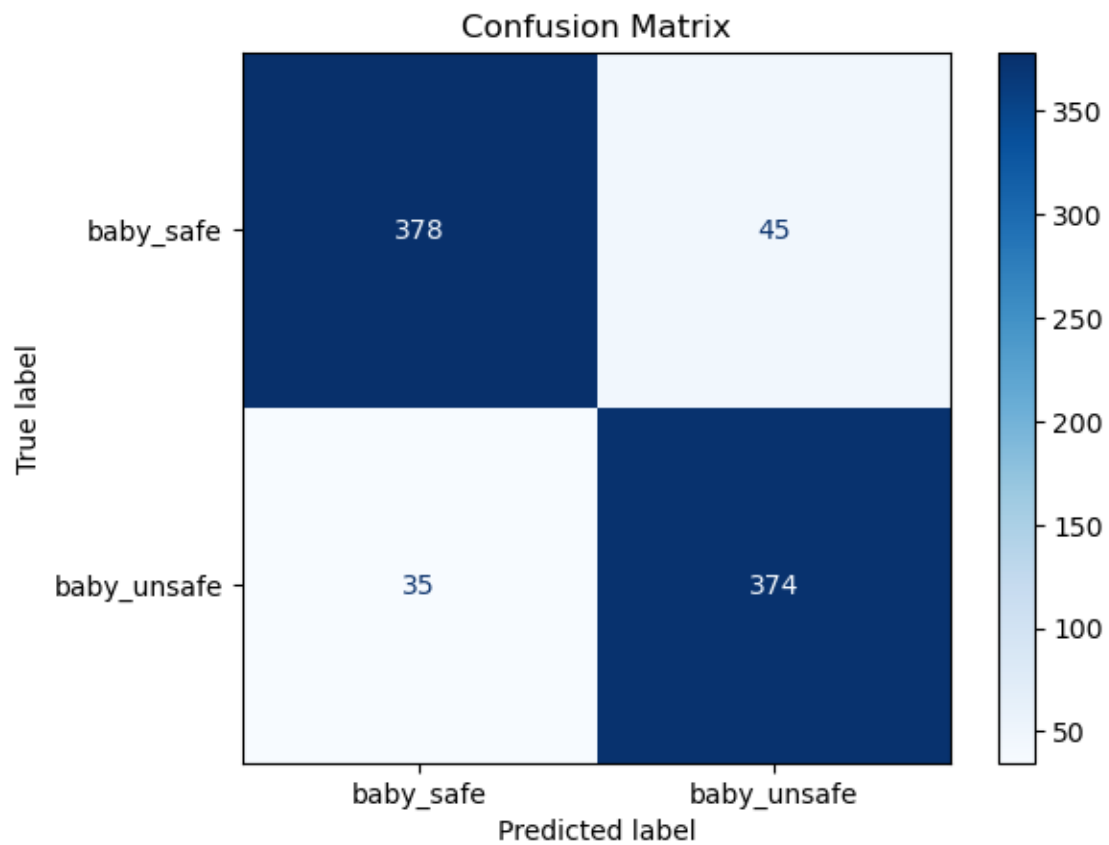


Dataset labels:-----  
{'baby\_safe': 0, 'baby\_unsafe': 1}

Report-----

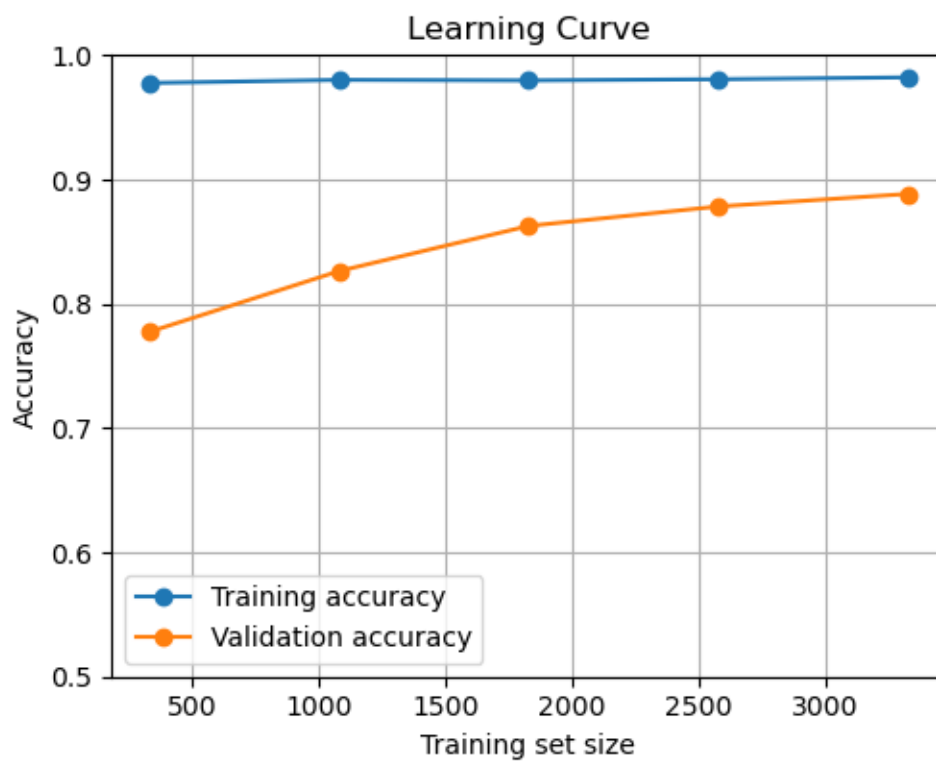
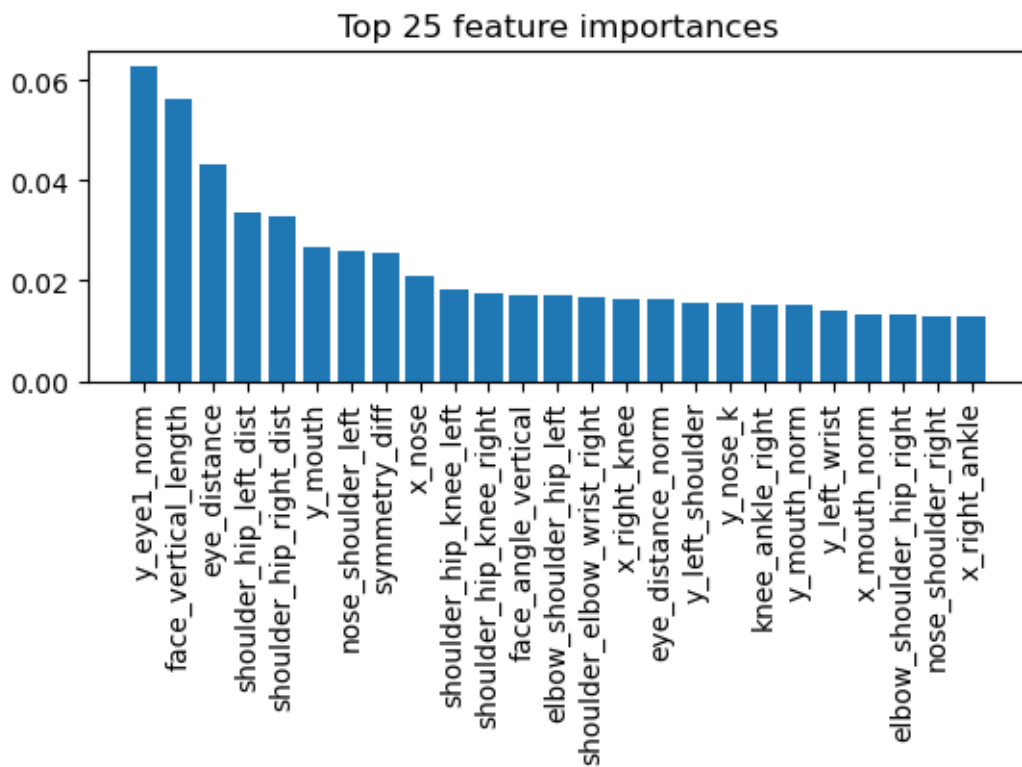
	precision	recall	f1-score	support
baby_safe	0.92	0.89	0.90	423
baby_unsafe	0.89	0.91	0.90	409
accuracy			0.90	832
macro avg	0.90	0.90	0.90	832
weighted avg	0.90	0.90	0.90	832

Confusion matrix-----



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-----  
-----TOP 25 FEATURES  
ANALYSIS-----  
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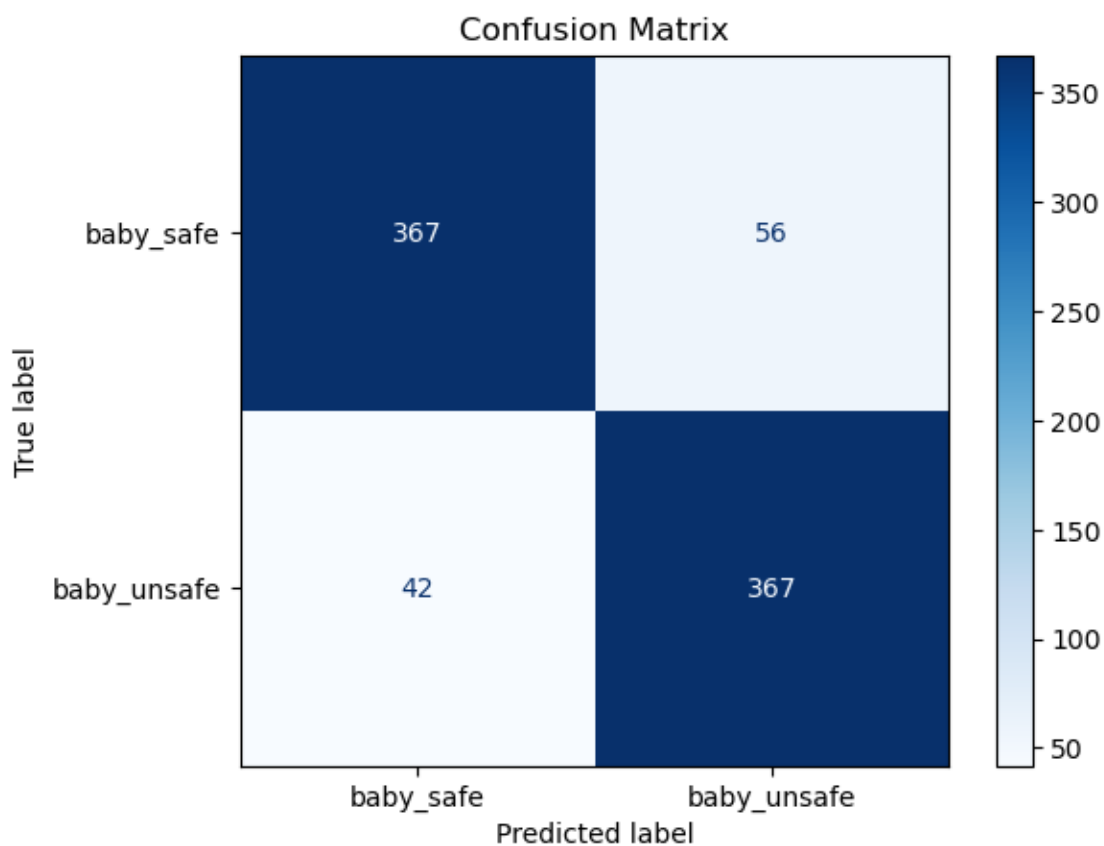


Dataset labels:-----  
{'baby\_safe': 0, 'baby\_unsafe': 1}

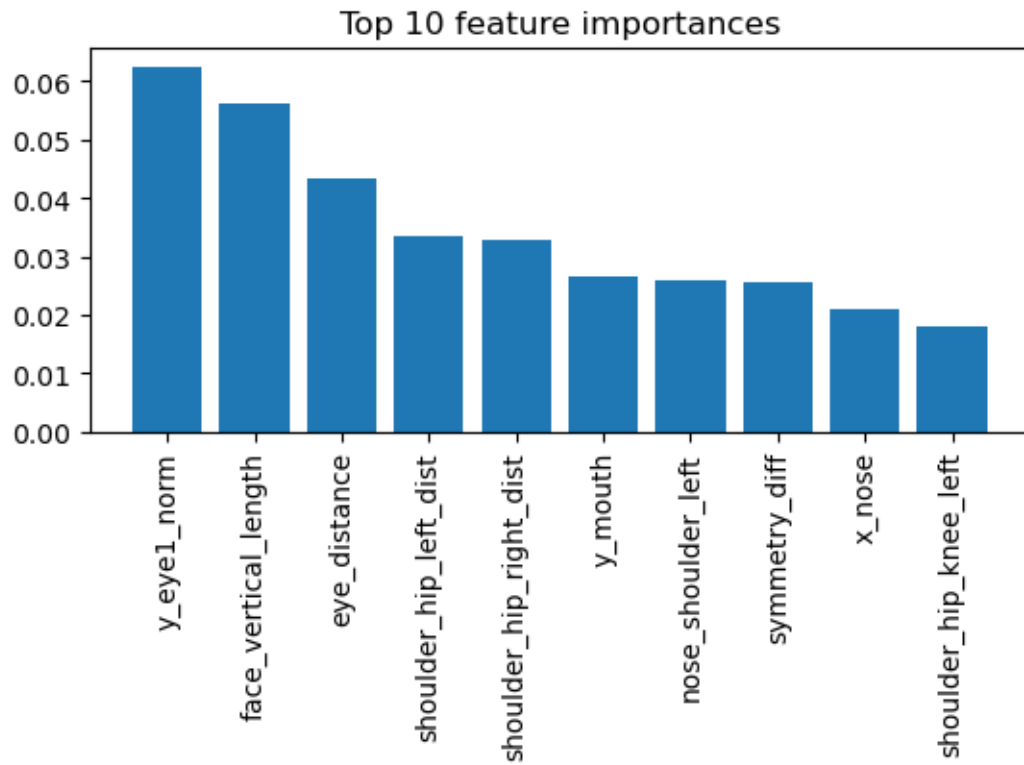
Report-----

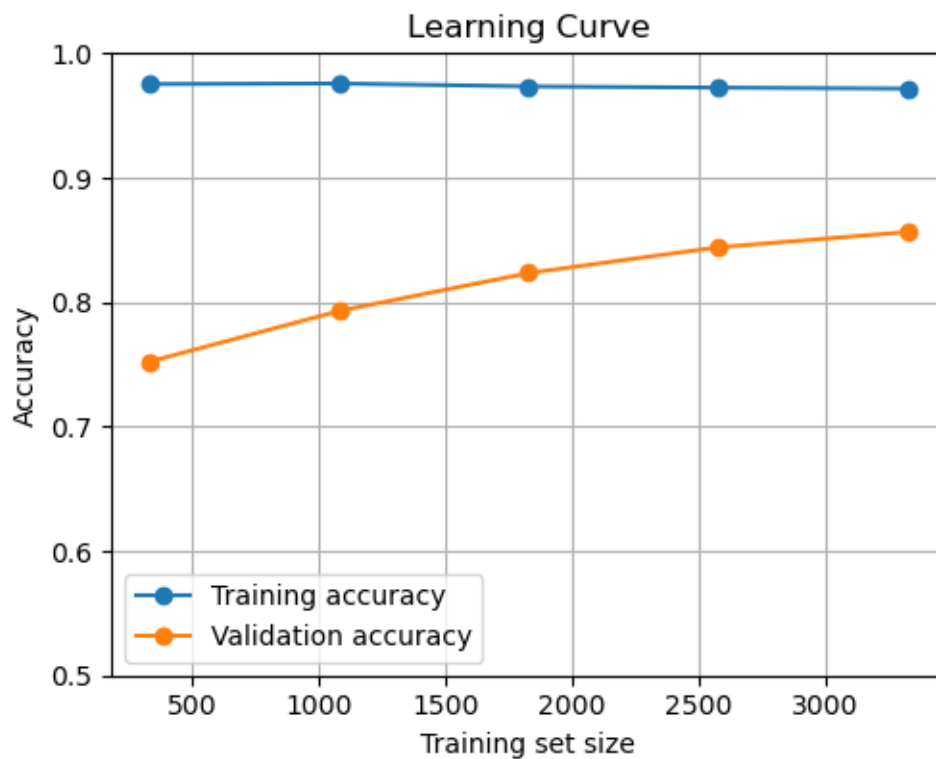
	precision	recall	f1-score	support
baby_safe	0.90	0.87	0.88	423
baby_unsafe	0.87	0.90	0.88	409
accuracy			0.88	832
macro avg	0.88	0.88	0.88	832
weighted avg	0.88	0.88	0.88	832

Confusion matrix-----



-----TOP 10 FEATURES  
ANALYSIS-----





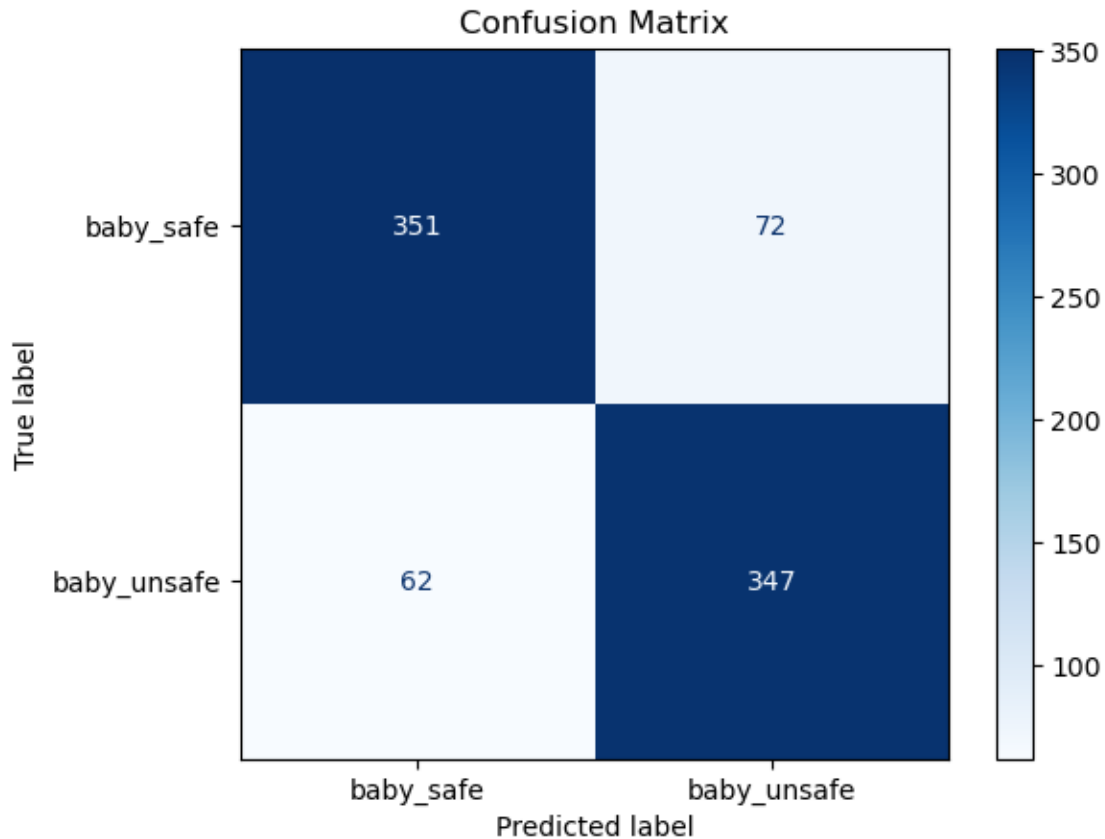
Dataset labels:-----

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

Report-----

	precision	recall	f1-score	support
baby_safe	0.85	0.83	0.84	423
baby_unsafe	0.83	0.85	0.84	409
accuracy			0.84	832
macro avg	0.84	0.84	0.84	832
weighted avg	0.84	0.84	0.84	832

Confusion matrix-----



```
[77]: prediction = results["all_features"]["y_predicted"]
true_y= clf.y_test
misclassified = np.where(true_y != prediction)[0]

misclassified_images = [clf.images_paths_test[i] for i in misclassified]
image_dataset_path=emb_builder.dataset

folder_path = f"{project_dir}image_prediction/approach_features_selection/"
if not os.path.exists(folder_path):
    os.makedirs(folder_path)

from PIL import Image, ImageDraw, ImageFont

for img_path, prediction in zip(misclassified_images, prediction):
    img = Image.open(f"{image_dataset_path}/{img_path}")
    draw = ImageDraw.Draw(img)

    try:
        font = ImageFont.truetype("DejaVuSans-Bold.ttf", size=34) #_
        ↪Imposta la dimensione del font
```

```

except IOError:
    font = ImageFont.load_default() # Usa il font di default se il
    ↳file ttf non è trovato

    predicted_class = [key for key, value in emb_builder.classes_bs.items()]
    ↳if value == prediction][0]
    text = f"{predicted_class}"
    text_position = (50, 50)
    text_color = (255, 0, 0)
    draw.text(text_position, text, fill=text_color, font=font)

    new_image_path = f"{folder_path}{img_path}"
    img.save(new_image_path)

print(f"Misclassified images successfully saved in {folder_path}")

```

Misclassified images successfully saved in /home/terra/Desktop/unimore/AI\_engineering/SIDS\_revelation\_project/image\_prediction/approach\_features\_selection/

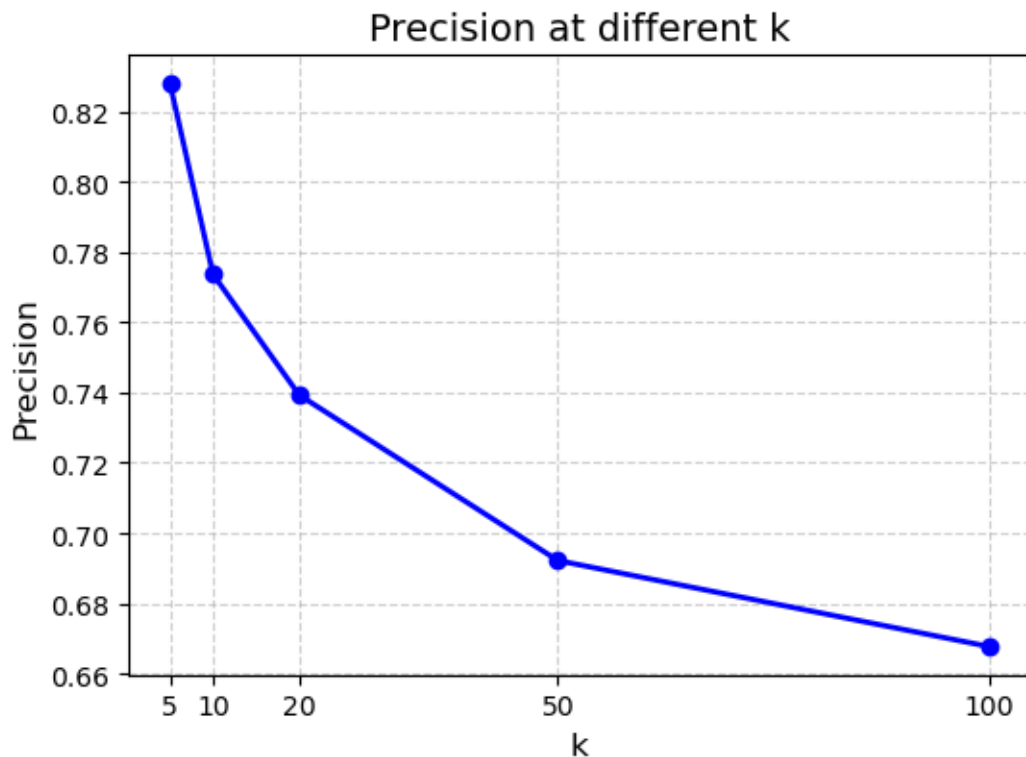
```

[78]: ret = ImageRetrieval(embeddings, emb_builder.y, emb_builder.image_paths,
    ↳emb_builder.dataset, emb_builder.classes_bs)
    ret.report('euclidean')

```

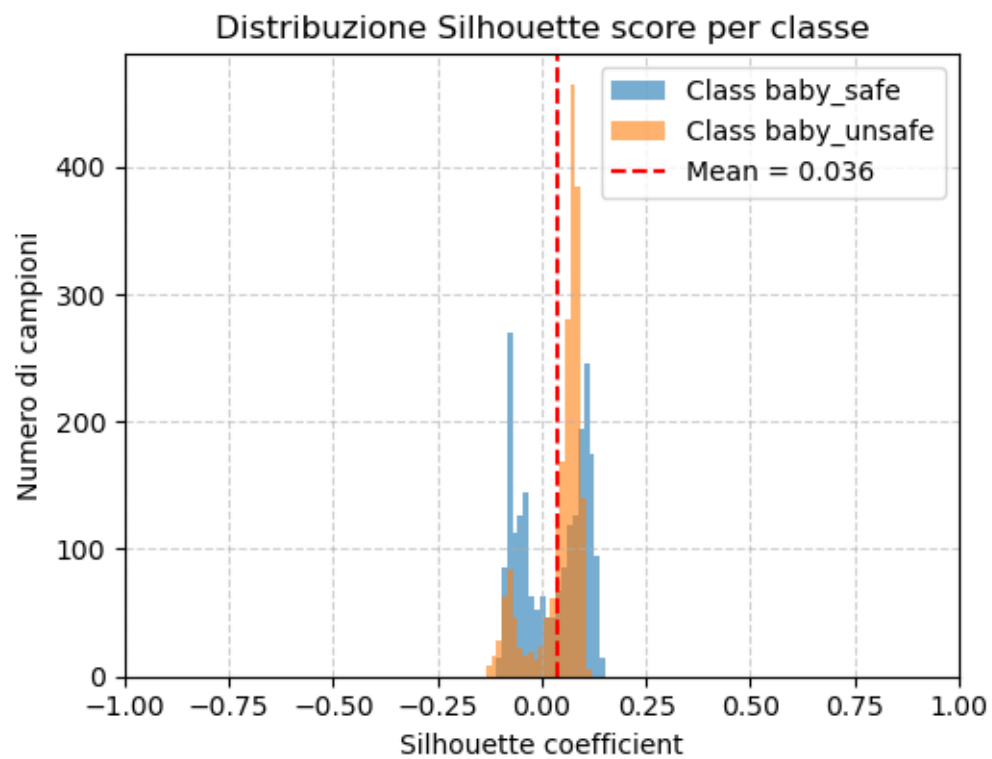
Precision at different

k:-----



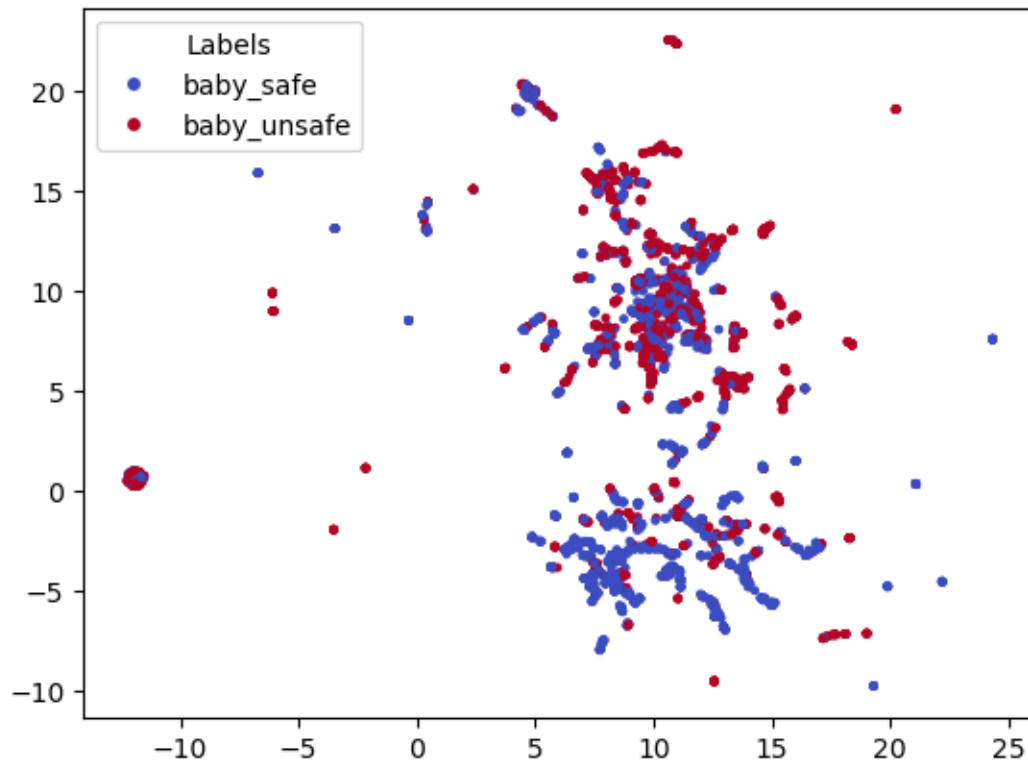
Recall at  
R-----  
0.5437276847656125

Silhouette  
score-----



Silhouette score (euclidean): 0.036

Embeddings  
distributions-----



```
[79]: image_paths = emb_builder.image_paths
      idx_query = 99
      image_to_retrieve = f"{emb_builder.dataset}/{image_paths[idx_query]}"

      print("Image to retrieve")
      img = mpimg.imread(image_to_retrieve)
      plt.figure(figsize=(3, 3))
      plt.imshow(img)
      plt.axis('off')
      plt.show()

      distances_all, image_paths_similar_all = ret.
      ↪ retrieve_similar(idx_query=idx_query,k=5,verbose=False)
      ret.show_images(image_paths_similar_all)
```

Image to retrieve





## 0.2 Best model with approach Supervised Learning Metric

XGBC with no optimization

```
[80]: embeddings = emb_builder.create_embedding(flags=True,positions=True,
↪positions_normalized=True,geometric_info=True,k_positions_normalized=True,
↪k_geometric_info=True)
```

Embedding

creation-----

Features: ['flag\_eye1', 'flag\_eye2', 'flag\_nose', 'flag\_mouth', 'x\_eye1', 'y\_eye1', 'x\_eye2', 'y\_eye2', 'x\_nose', 'y\_nose', 'x\_mouth', 'y\_mouth', 'x\_eye1\_norm', 'y\_eye1\_norm', 'x\_eye2\_norm', 'y\_eye2\_norm', 'x\_nose\_norm', 'y\_nose\_norm', 'x\_mouth\_norm', 'y\_mouth\_norm', 'eye\_distance', 'eye\_distance\_norm', 'face\_vertical\_length', 'face\_vertical\_length\_norm', 'face\_angle\_vertical', 'face\_angle\_horizontal', 'symmetry\_diff', 'head\_ration', 'x\_nose\_k', 'y\_nose\_k', 'x\_left\_eye\_k', 'y\_left\_eye\_k', 'x\_right\_eye\_k', 'y\_right\_eye\_k', 'x\_left\_ear', 'y\_left\_ear', 'x\_right\_ear', 'y\_right\_ear', 'x\_left\_shoulder', 'y\_left\_shoulder', 'x\_right\_shoulder', 'y\_right\_shoulder', 'x\_left\_elbow', 'y\_left\_elbow', 'x\_right\_elbow', 'y\_right\_elbow',

```
'x_left_wrist', 'y_left_wrist', 'x_right_wrist', 'y_right_wrist', 'x_left_hip',
'y_left_hip', 'x_right_hip', 'y_right_hip', 'x_left_knee', 'y_left_knee',
'x_right_knee', 'y_right_knee', 'x_left_ankle', 'y_left_ankle', 'x_right_ankle',
'y_right_ankle', 'shoulders_dist', 'shoulder_hip_right_dist',
'shoulder_hip_left_dist', 'nose_shoulder_right', 'nose_shoulder_left',
'shoulder_left_knee_right', 'shoulder_right_knee_left', 'knee_ankle_right',
'knee_ankle_left', 'nose_hip_right', 'nose_hip_left',
'elbow_shoulder_hip_right', 'elbow_shoulder_hip_left',
'shoulder_elbow_wrist_right', 'shoulder_elbow_wrist_left',
'shoulder_hip_knee_right', 'shoulder_hip_knee_left', 'hip_knee_ankle_right',
'hip_knee_ankle_left', 'shoulders_line_inclination', 'hips_line_inclination',
'torsion']
```

FINISHED: 4158 embedding created

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```
[87]: dataset = EmbeddingDataset(embeddings.to_numpy(), emb_builder.y, device=device)
model = dataset.train_embeddings(embed_dim=32, epochs=50, batch_size=128,
    ↪lr=1e-3, verbose=False, weight_decay=1e-7, dropout_rate=0.05)

embeddings_new = dataset.extract_embeddings(model)
embeddings_new= pd.DataFrame(embeddings_new.to_numpy(), columns=[f"f_{i}" for i
    ↪in range(embeddings_new.shape[1])])
clf = Classifier(embeddings_new, emb_builder.y, emb_builder.classes_bs,
    ↪image_paths=emb_builder.image_paths)
```

```
[88]: params = {
    'n_estimators': 300,
    'max_depth': 5,
    'learning_rate': 0.05,
    'subsample': 0.8,
    'colsample_bytree': 0.8,
    'reg_lambda': 1,
    'reg_alpha': 0.5,
    'random_state': None
}
model = XGBClassifier(**params)

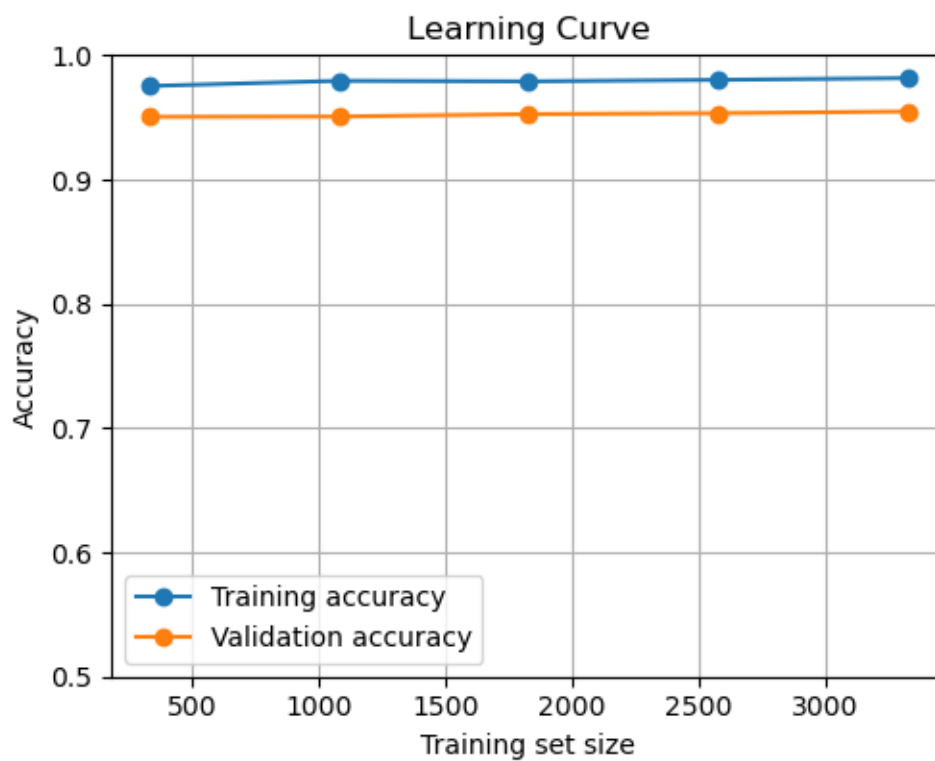
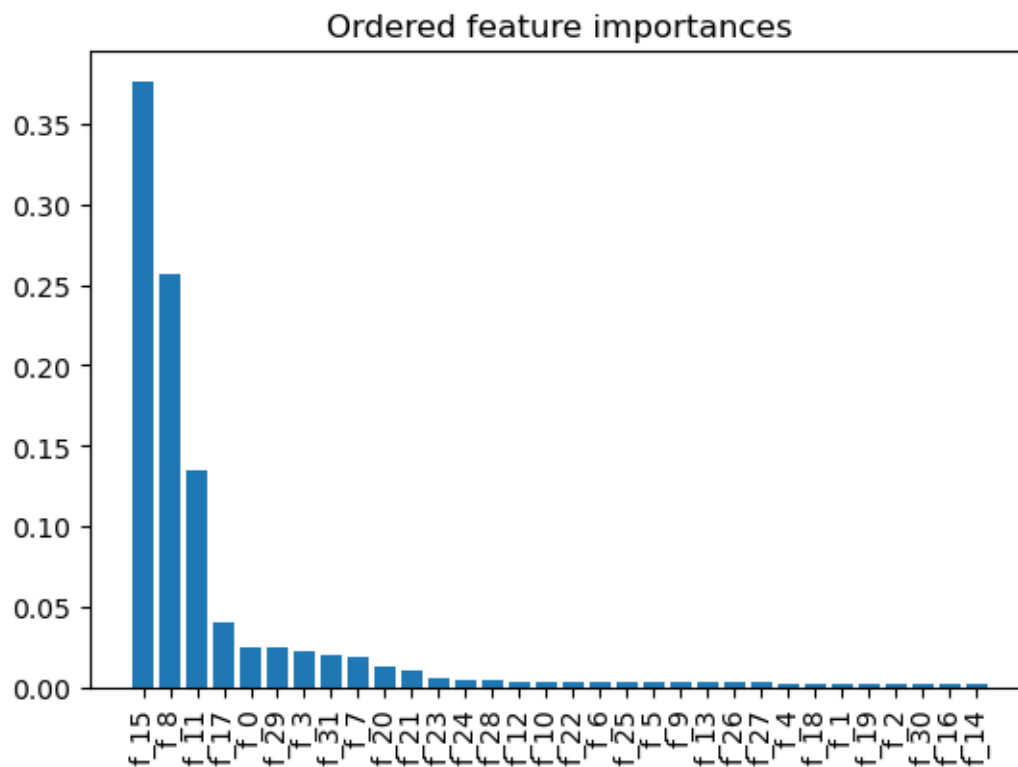
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-----FIRST

ANALYSIS-----

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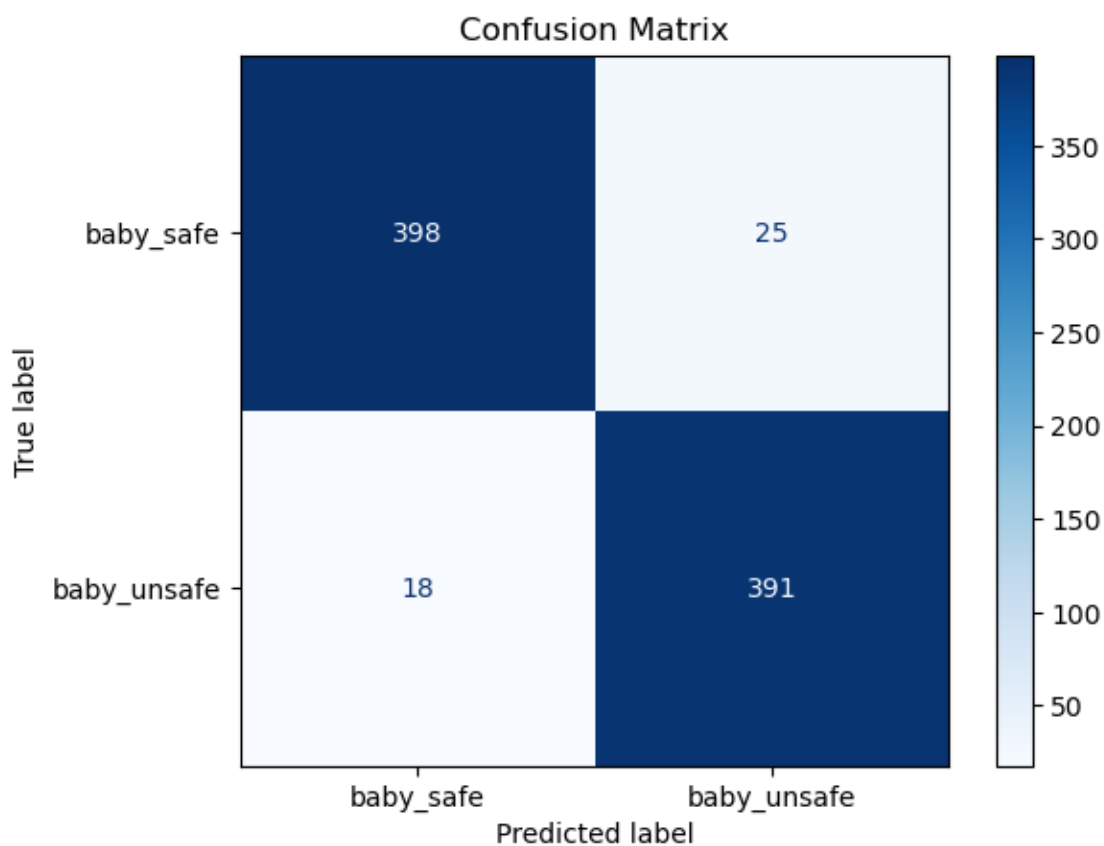


Dataset labels:-----  
{'baby\_safe': 0, 'baby\_unsafe': 1}

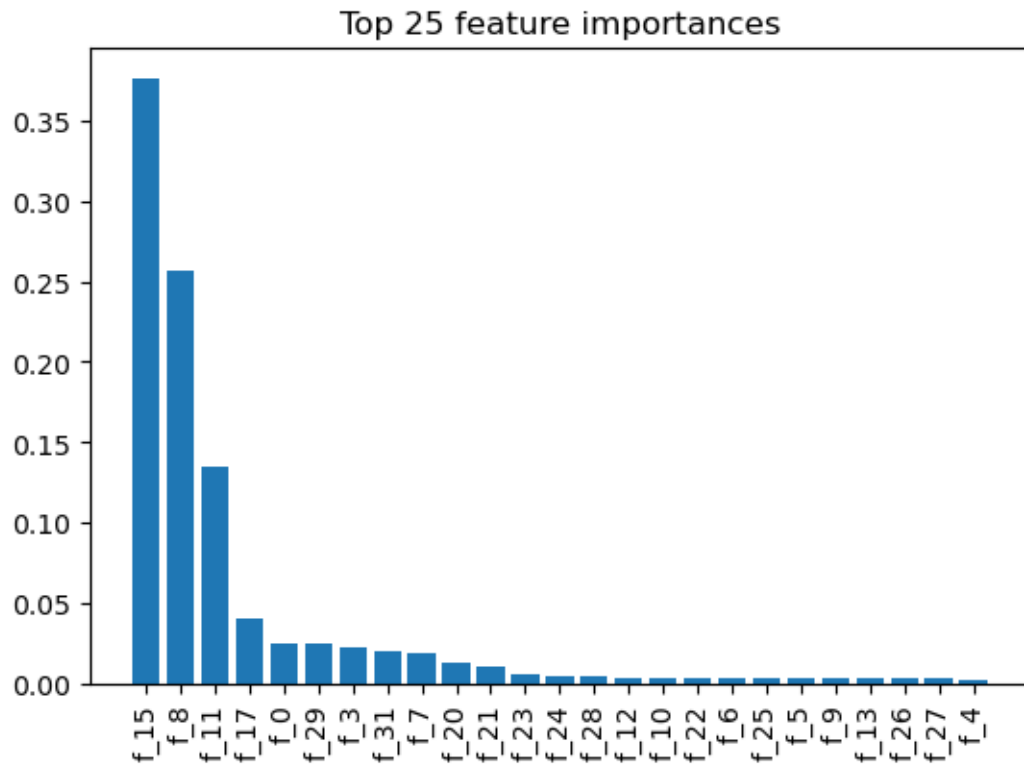
Report-----

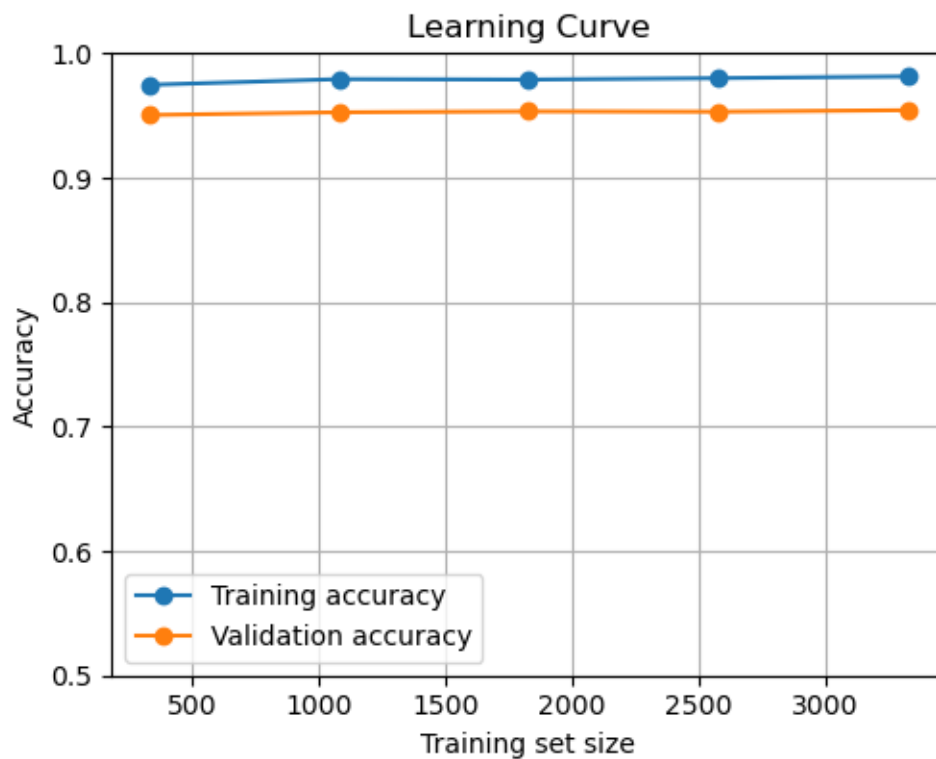
	precision	recall	f1-score	support
baby_safe	0.96	0.94	0.95	423
baby_unsafe	0.94	0.96	0.95	409
accuracy			0.95	832
macro avg	0.95	0.95	0.95	832
weighted avg	0.95	0.95	0.95	832

Confusion matrix-----



-----TOP 25 FEATURES  
ANALYSIS-----





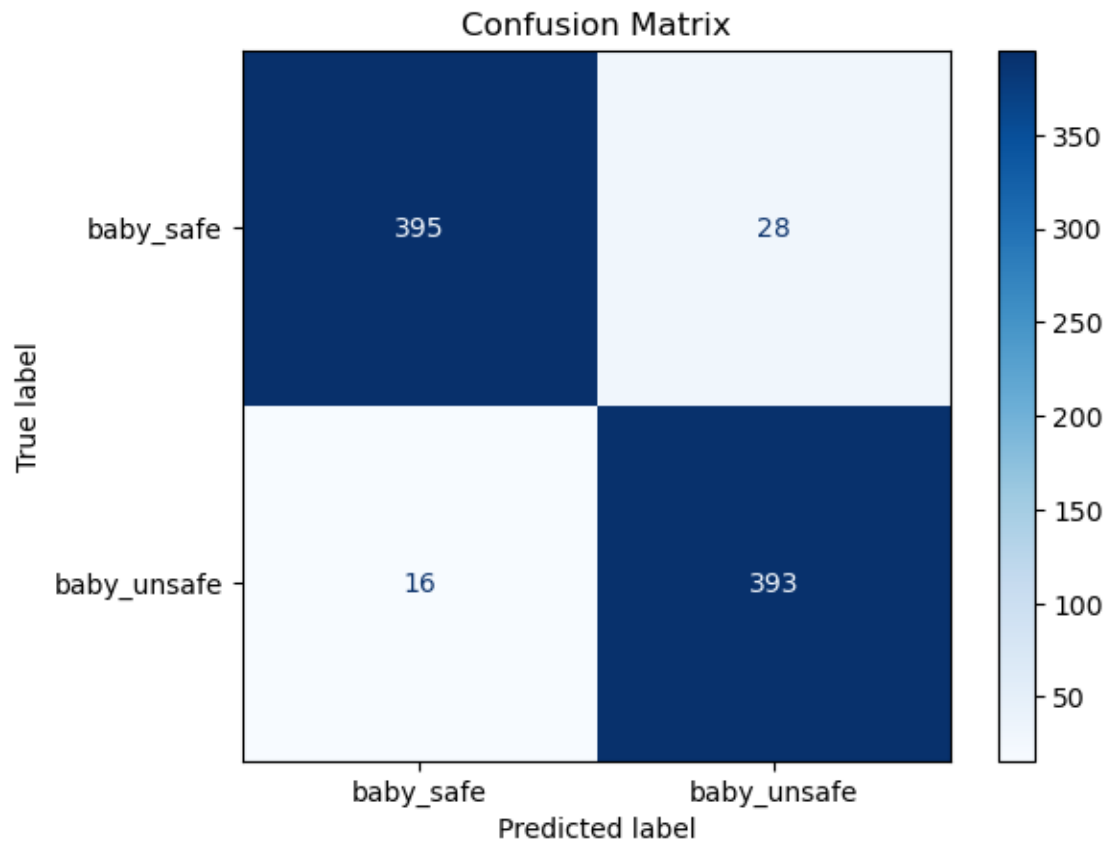
Dataset labels:-----

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

Report-----

	precision	recall	f1-score	support
baby_safe	0.96	0.93	0.95	423
baby_unsafe	0.93	0.96	0.95	409
accuracy			0.95	832
macro avg	0.95	0.95	0.95	832
weighted avg	0.95	0.95	0.95	832

Confusion matrix-----



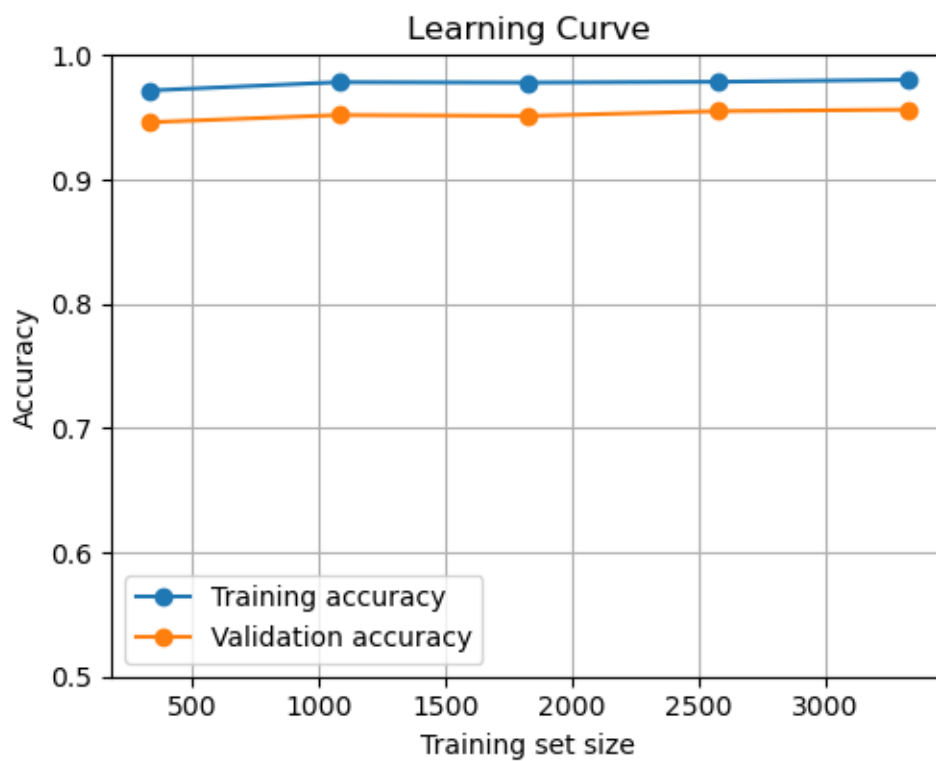
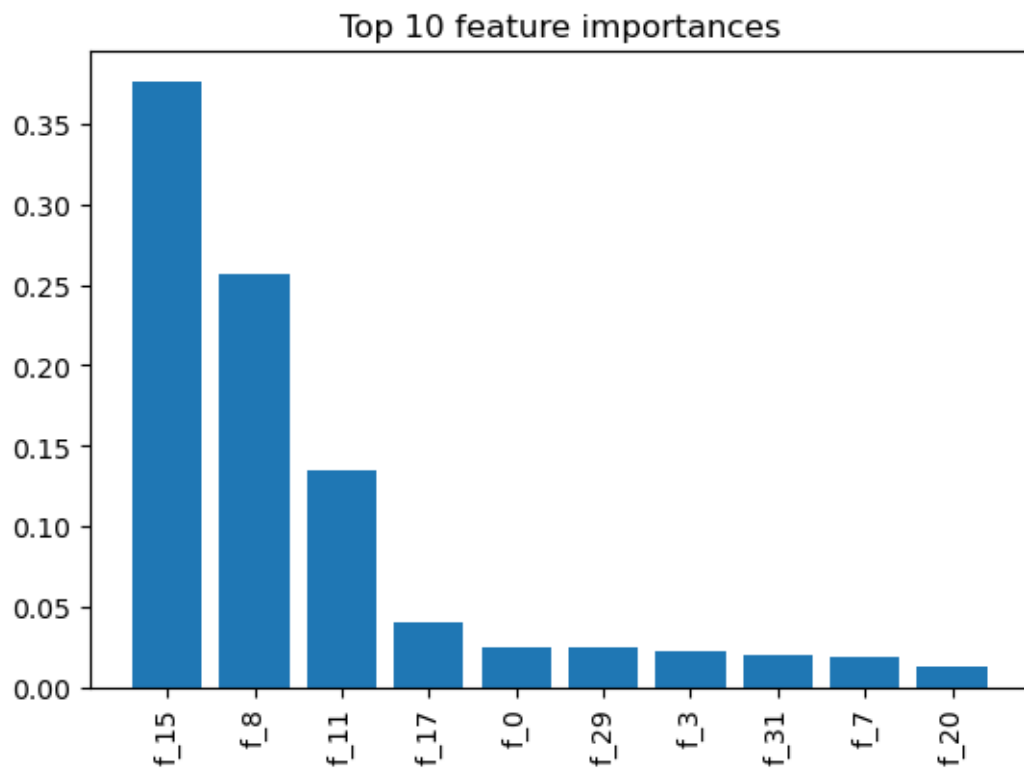
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-----TOP 10 FEATURES

ANALYSIS-----

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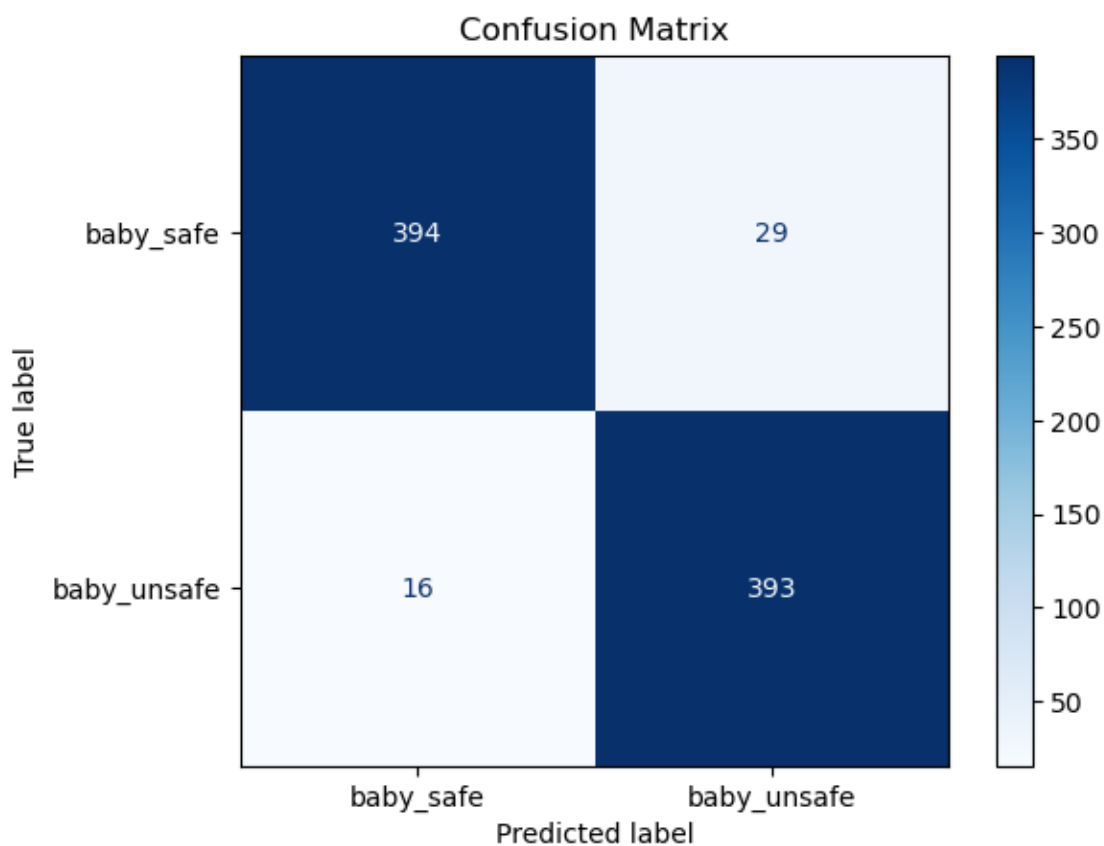


Dataset labels:-----  
{'baby\_safe': 0, 'baby\_unsafe': 1}

Report-----

	precision	recall	f1-score	support
baby_safe	0.96	0.93	0.95	423
baby_unsafe	0.93	0.96	0.95	409
accuracy			0.95	832
macro avg	0.95	0.95	0.95	832
weighted avg	0.95	0.95	0.95	832

Confusion matrix-----



```
[89]: prediction = results["all_features"]["y_predicted"]
true_y= clf.y_test
misclassified = np.where(true_y != prediction)[0]

misclassified_images = [clf.images_paths_test[i] for i in misclassified]
image_dataset_path=emb_builder.dataset

folder_path = f"{project_dir}image_prediction/
↳approach_supervised_learning_metric/"
if not os.path.exists(folder_path):
    os.makedirs(folder_path)

from PIL import Image, ImageDraw, ImageFont

for img_path, prediction in zip(misclassified_images, prediction):
    img = Image.open(f"{image_dataset_path}/{img_path}")
    draw = ImageDraw.Draw(img)

    try:
        font = ImageFont.truetype("DejaVuSans-Bold.ttf", size=34) #↳
        ↳Imposta la dimensione del font
    except IOError:
        font = ImageFont.load_default() # Usa il font di default se il
        ↳file ttf non è trovato

    predicted_class = [key for key, value in emb_builder.classes_bs.items()
↳if value == prediction][0]
    text = f"{predicted_class}"
    text_position = (50, 50)
    text_color = (255, 0, 0)
    draw.text(text_position, text, fill=text_color, font=font)

    new_image_path = f"{folder_path}/{img_path}"
    img.save(new_image_path)

print(f"Misclassified images successfully saved in {folder_path}")
```

Misclassified images successfully saved in /home/terra/Desktop/unimore/AI\_engineering/SIDS\_revelation\_project/image\_prediction/approach\_supervised\_learning\_metric/

```
[ ]: ret = ImageRetrieval(embeddings_new, emb_builder.y, emb_builder.image_paths,
↳emb_builder.dataset, emb_builder.classes_bs)
ret.report('euclidean')
```

Precision at different

k:-----

```
[ ]: image_paths = emb_builder.image_paths
idx_query = 99
image_to_retrieve = f"{emb_builder.dataset}/{image_paths[idx_query]}"

print("Image to retrieve")
img = mpimg.imread(image_to_retrieve)
plt.figure(figsize=(3, 3))
plt.imshow(img)
plt.axis('off')
plt.show()

distances_all, image_paths_similar_all = ret.
    ↳ retrieve_similar(idx_query=idx_query,k=5,verbose=False)
ret.show_images(image_paths_similar_all)
```

### 0.3 Best model with approach Supervised Learning Metric

XGBC with optimization

Best parameters : {'colsample\_bytree': np.float64(0.6547542520275229), 'gamma': np.float64(0.45000932092405255), 'learning\_rate': np.float64(0.2721670232687546), 'max\_depth': 4, 'n\_estimators': 367, 'subsample': np.float64(0.853239225342979)} Best mean cross-validation accuracy: 0.9735415114362482

```
[ ]: embeddings = emb_builder.create_embedding(flags=True,positions=True,
    ↳ positions_normalized=True, geometric_info=True,k_positions_normalized=True,
    ↳ k_geometric_info=True)
```

```
[ ]: dataset = EmbeddingDataset(embeddings.to_numpy(),emb_builder.y,device=device)
model = dataset.train_embeddings(embed_dim=32, epochs=50, batch_size=128,
    ↳ lr=1e-3,verbose=False,weight_decay=1e-7,dropout_rate=0.05)

embeddings_new = dataset.extract_embeddings(model)
embeddings_new= pd.DataFrame(embeddings_new.to_numpy(), columns=[f"f_{i}" for i_
    ↳ in range(embeddings_new.shape[1])])
clf = Classifier(embeddings_new, emb_builder.y, emb_builder.classes_bs)
```

```
[ ]: best_params = {
    'colsample_bytree': np.float64(0.6547542520275229),
    'gamma': np.float64(0.45000932092405255),
    'learning_rate': np.float64(0.2721670232687546),
    'max_depth': 4,
    'n_estimators': 367,
    'subsample': np.float64(0.853239225342979)
}
model = XGBClassifier(**best_params)
```

```

with warnings.catch_warnings():
    warnings.simplefilter("ignore")
    results = clf.evaluation_pipeline_save_misclassified(model)

```

```

[ ]: prediction = results["all_features"]["y_predicted"]
true_y= clf.y_test
misclassified = np.where(true_y != prediction)[0]

misclassified_images = [clf.images_paths_test[i] for i in misclassified]
image_dataset_path=emb_builder.dataset

folder_path = f"{project_dir}image_prediction/
↳approach_supervised_learning_metric_optimized/"
if not os.path.exists(folder_path):
    os.makedirs(folder_path)

from PIL import Image, ImageDraw, ImageFont

for img_path, prediction in zip(misclassified_images, prediction):
    img = Image.open(f"{image_dataset_path}/{img_path}")
    draw = ImageDraw.Draw(img)

    try:
        font = ImageFont.truetype("DejaVuSans-Bold.ttf", size=34) #↳
        ↳Imposta la dimensione del font
    except IOError:
        font = ImageFont.load_default() # Usa il font di default se il↳
        ↳file ttf non è trovato

    predicted_class = [key for key, value in emb_builder.classes_bs.items()↳
    ↳if value == prediction][0]
    text = f"{predicted_class}"
    text_position = (50, 50)
    text_color = (255, 0, 0)
    draw.text(text_position, text, fill=text_color, font=font)

    new_image_path = f"{folder_path}/{img_path}"
    img.save(new_image_path)

print(f"Misclassified images successfully saved in {folder_path}")

```

Retrieval is invariant, same embeddings as previous point

```

[ ]: ret = ImageRetrieval(embeddings_new, emb_builder.y, emb_builder.image_paths,↳
    ↳emb_builder.dataset, emb_builder.classes_bs)
ret.report('euclidean')

```

```
[ ]: image_paths = emb_builder.image_paths
idx_query = 99
image_to_retrieve = f"{emb_builder.dataset}/{image_paths[idx_query]}"

print("Image to retrieve")
img = mpimg.imread(image_to_retrieve)
plt.figure(figsize=(3, 3))
plt.imshow(img)
plt.axis('off')
plt.show()

distances_all, image_paths_similar_all = ret.
    ↳ retrieve_similar(idx_query=idx_query,k=5,verbose=False)
ret.show_images(image_paths_similar_all)
```

```
[ ]: import ipynbname
from libraries.file_manager_utils import *

save_as_pdf(ipynbname.path())
```