



# Sound classification of the UrbanSound8K

**Subject:** Pattern Recognition and Machine Learning

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# Objectives

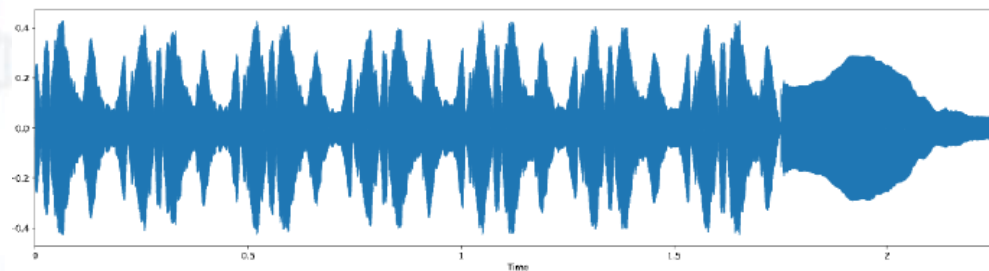
Develop an audio classifier



Capable of distinguishing between  
the several sound classes

- 0 = air conditioner
- 1 = car horn
- 2 = children playing
- 3 = dog bark
- 4 = drilling
- 5 = engine idling
- 6 = gun shot
- 7 = jackhammer
- 8 = siren
- 9 = street music

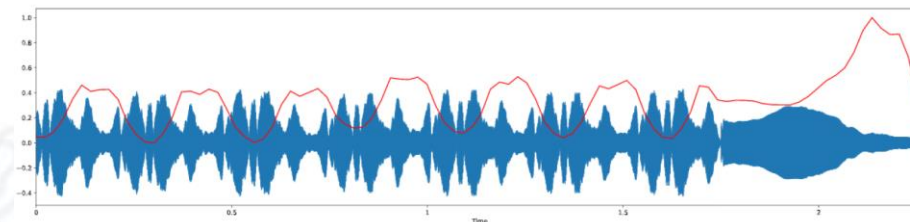
*Different labeled classes from the Urban Sound Dataset*



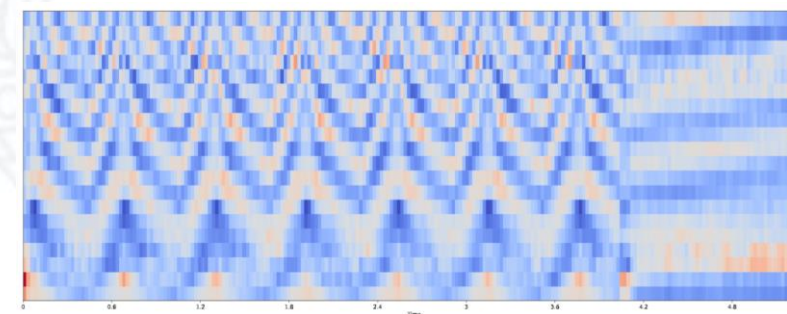
*Plot of the audio array in waveform manner*

# Candidate features

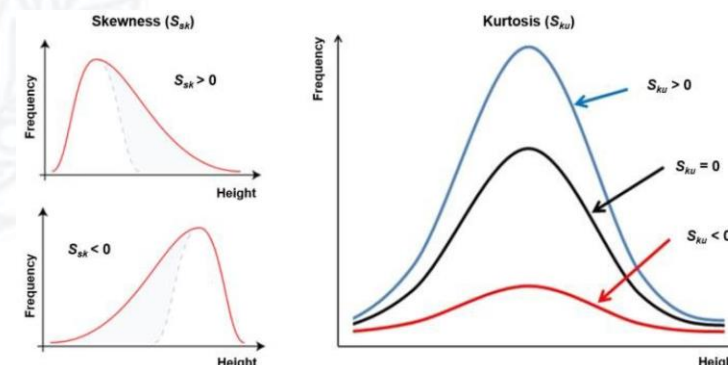
1. Zero Crossing Rate (1x1 dimension)
2. Spectral Centroid (1x1 dimension)
3. Spectral Rollof (1x1 dimension)
4. Mel Frequency Cepstrum Coefficient (MFCC) (13x1 dimension)
5. Spectral Flatness (1x1 dimension)
6. Skewness (1x1 dimension)
7. Kurtosis (1x1 dimension)
8. Spectral Entropy (1x1 dimension)
9. Tempo (1x1 dimension)
10. Tonal centroids (6x1 dimension)



*Plot Spectral Centroid (red) with the waveform (blue)*

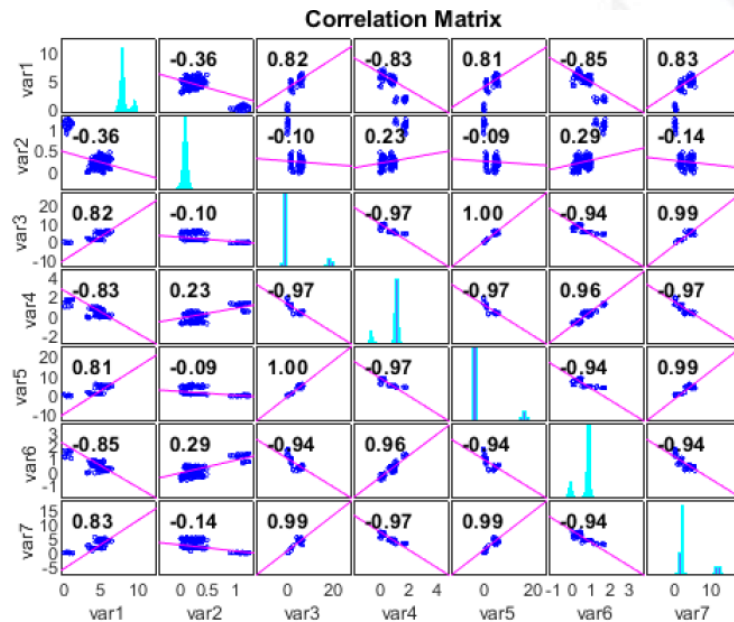


*MFCCs scaled such that each coefficient dimension has zero mean and unit variance*

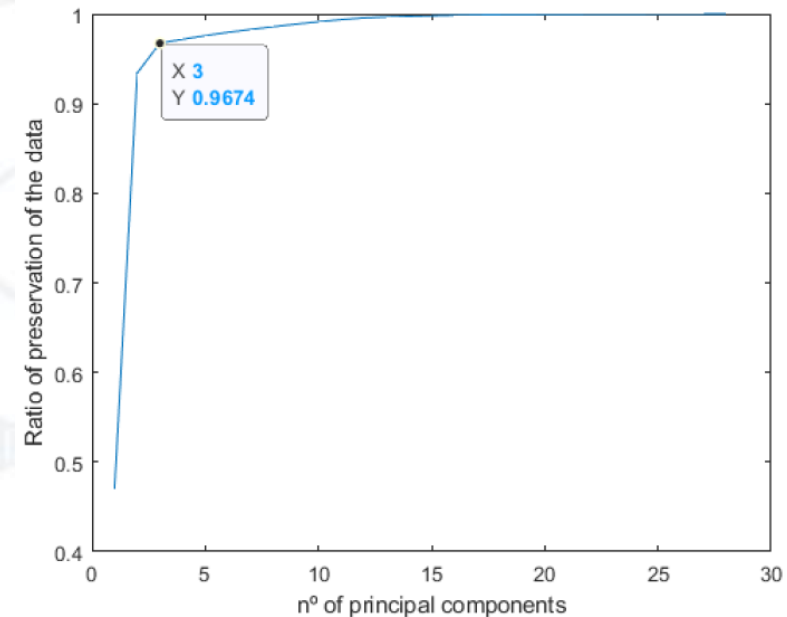


*Skewness and Kurtosis graphically explained*

# Reduction of the features



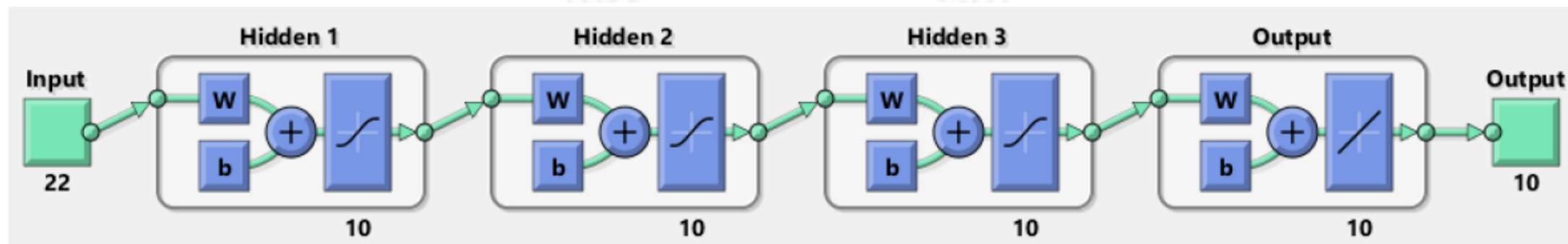
Correlation matrix with 7 features of the original dataset



Number of components that explain the data

As it can be perceived, it is needed 3 components to explain the 97% of the data.

# First classifier: NN



*Configuration of the network.*

3 methods to explain  
the different types of  
training functions

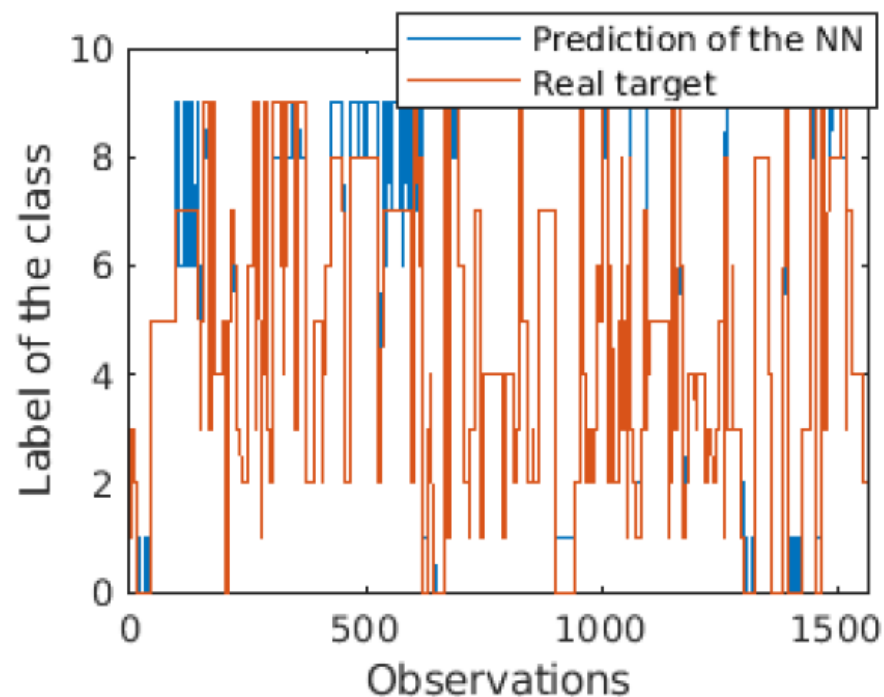


1. Levenberg-Marquardt
2. Bayesian Regularization
3. BFGS Quasi-Newton

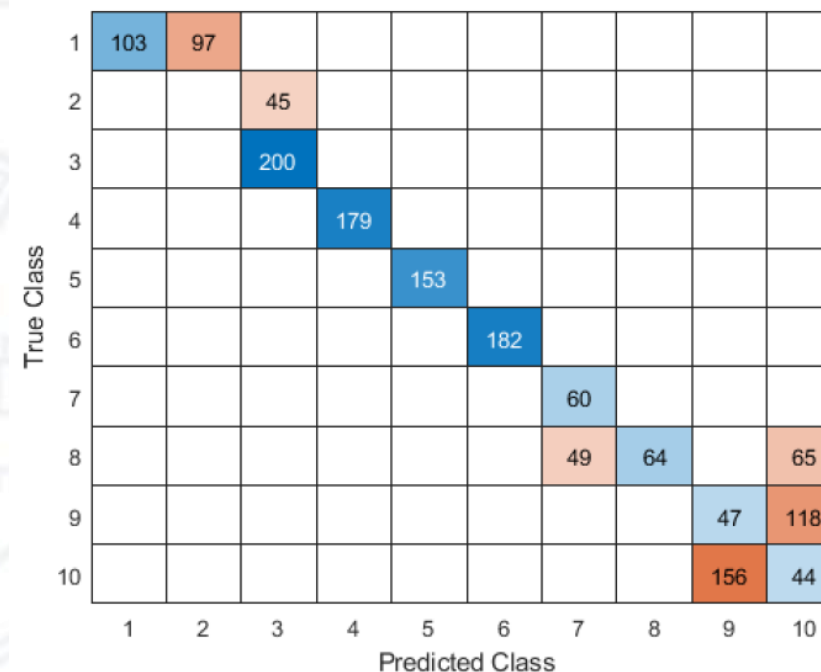


Training functions	1	2	3
Accuracy(%) all features	100	100	24.9
Accuracy(%) applying PCA	66.1	100	77.8

# First classifier: NN



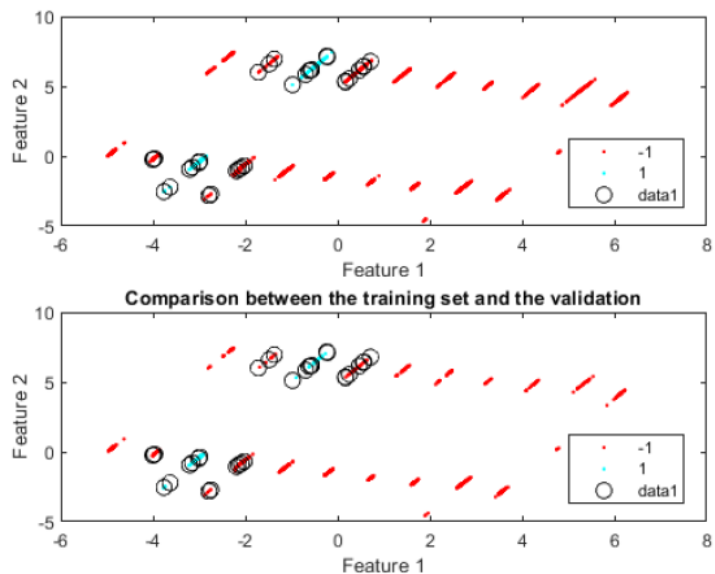
*Results between the prediction and real target using Levenberg-Marquardt*



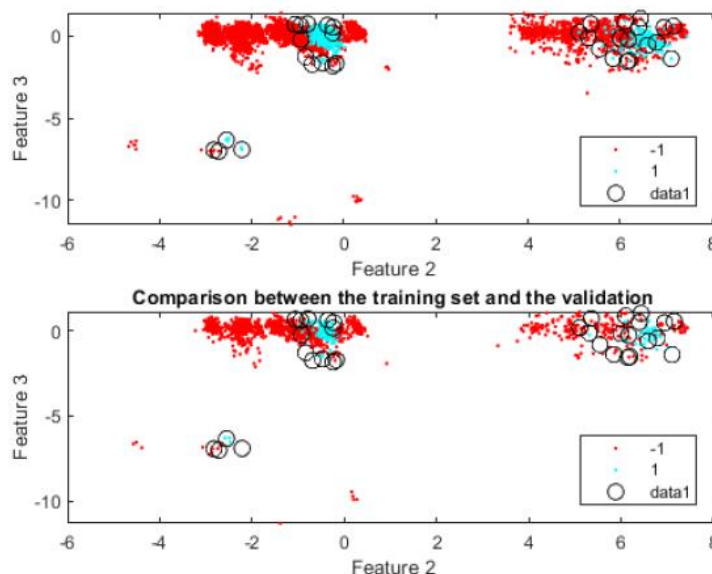
*Plot of the confusion matrix of NN applied to the validation data.*



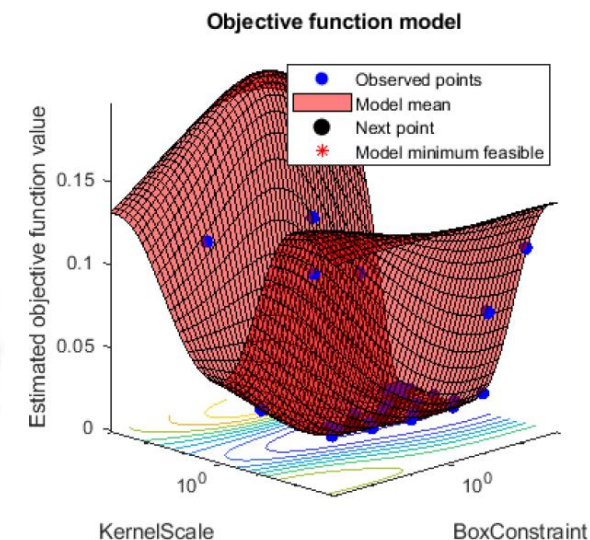
# Second classifier: SVM



First and second features and support vectors for the training (above) and testing (below)



Second and third features and support vectors for the training (above) and testing (below)



Plot of the objective function when the class is 2

10 different models  
with its optimized  
parameters



Observing results in  
training and predicting

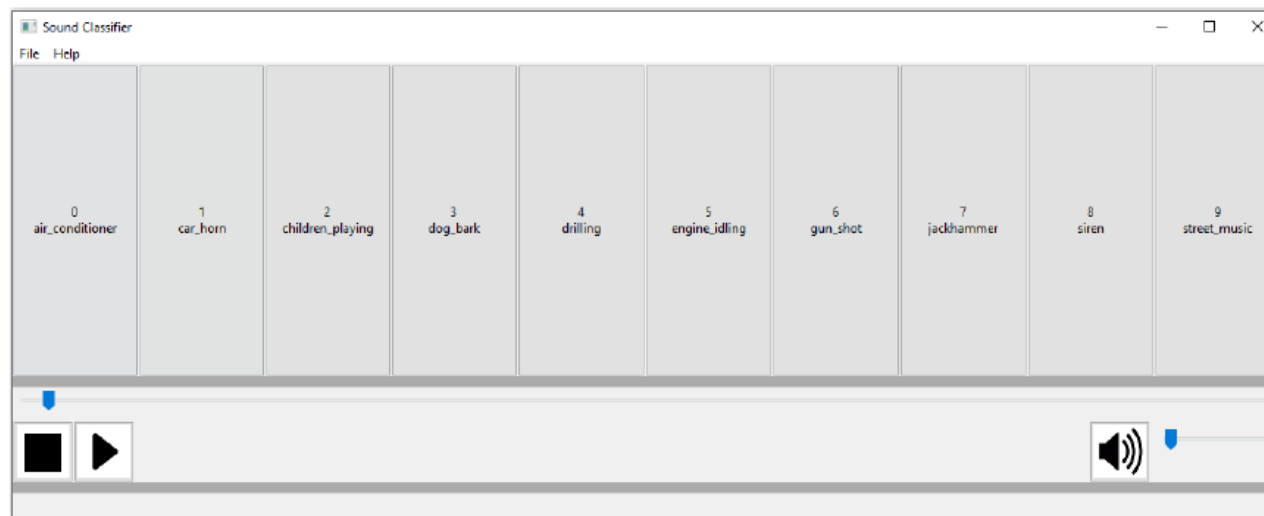


Accuracy obtained with  
SVM is 99,88%,



# Interface to classify new sounds

- To annotate new incoming data that were to make its way to the existing dataset.
- Make it easier and less time consuming for the user to store a label.
- Created with the framework wxPython, a cross-platform GUI toolkit for the Python language.



*Snapshot of the interface created to ease the data labelling process*



# Conclusions

- Analyzing a dataset and extracting the features.
- Applying methods studied during the course and prove how it changes the performance tuning the different parameters of these methods.
- Best results using the Levenberg-Marquardt and Bayesian Regularization functions with 1000 epochs.
- Respect the SVM classification, a very good performance has been proven.
- The computational cost is lower using SVM respect the NN, however the accuracy using NN is 100 respect the 99,8% of the SVM.
- Finally, an interface has been created to obtain new sounds.



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Thank you