

SPECTRAL ROLLOFF

It defines the shape of the signal, It represents the frequency below which a specified percentage of the total spectral energy, e.g. 85%, lies (0.85 is by default when value is not defined using the spectral rolloff module of librosa). This can also be used to, approximate the maximum (or minimum) frequency by setting roll_percent to a value close to 1 (or 0).

```
In [1]: import librosa as lr
import librosa.display
import matplotlib.pyplot as plt
import sklearn.preprocessing
import matplotlib.style as ms
%matplotlib inline
import numpy as np
import os
```

Visualizing the roll off frequency and power spectrum,

```
In [2]: audio='/home/vivek/Genre_Detection/genres/metal/metal.00001.au'
x, sr = lr.load(audio)
```

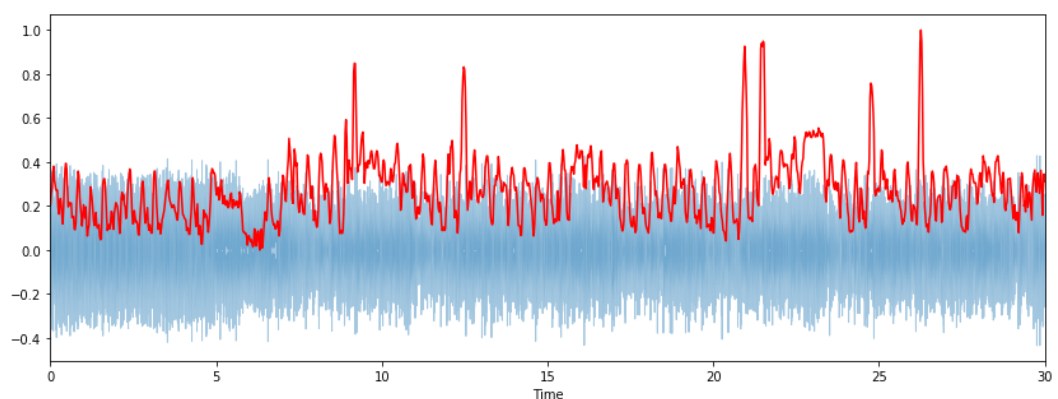
```
In [3]: spectral_rolloff = librosa.feature.spectral_rolloff(x, sr=sr, roll_percent=0.85)[0]

frames = range(len(spectral_rolloff))
t = librosa.frames_to_time(frames)

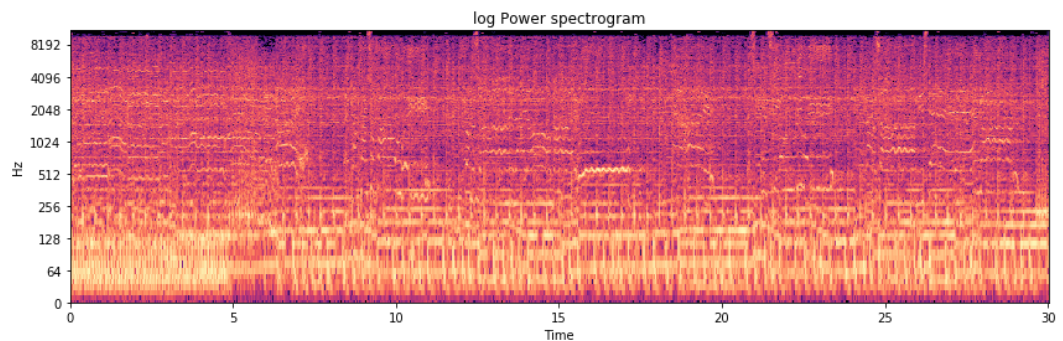
def normalize(x, axis=0):
    return sklearn.preprocessing.minmax_scale(x, axis=axis)

plt.figure(figsize=(14, 5))
librosa.display.waveplot(x, sr=sr, alpha=0.4)
plt.plot(t, normalize(spectral_rolloff), color='r')
```

```
Out[3]: [<matplotlib.lines.Line2D at 0x7febdd78de48>]
```



```
In [5]: S, phase = librosa.magphase(librosa.stft(x))
plt.figure(figsize=(12,4))
librosa.display.specshow(librosa.amplitude_to_db(S, ref=np.max),y_axis='log'
, x_axis='time')
plt.title('log Power spectrogram')
plt.tight_layout()
```



```
In [6]: #mean of the spectral rolloff acts as a good feature for training the model
and predicting it
np.mean(spectral_rolloff)
```

```
Out[6]: 6037.074141712587
```

```
In [ ]:
```