

```
In [2]: import tensorflow as tf #Derin Ağ için tensorflow u import ettik
from tensorflow.keras.preprocessing.image import ImageDataGenerator #rescale edip oranlamak ve test değerlerini a

base_dir = 'C:/Users/s7522/Desktop/UrbanSound8K/spectrograms/' #spectrogramların olduğu path i aldık.

train_datagen = ImageDataGenerator(rescale = 1./255, validation_split = 0.1) #Her bir görseli 255 e bölerek numpy
test_datagen = ImageDataGenerator(rescale = 1./255, validation_split = 0.1)

train_datagen = train_datagen.flow_from_directory(base_dir, target_size = (100,100), subset = "training", batch_s
test_datagen = test_datagen.flow_from_directory(base_dir, target_size = (100,100), subset = "validation", batch_s

Found 7861 images belonging to 10 classes.
Found 871 images belonging to 10 classes.
```

```
In [3]: import matplotlib.pyplot as plt #5 tane spectrogram üzerinden test

for _ in range(5):
    img, label = test_datagen.next() #test_datagen de iki tane çıktı vardır image ve label
    print(img.shape) #Gelen resimlerin shape ini gördük.
    plt.imshow(img[0])
    print(img[0]) #img ve label gördük
    print(label[0])
    plt.show()
```

```
(2, 100, 100, 3)
[[[1.         1.         1.         ]
  [0.         0.         0.01176471]
  [0.         0.         0.01176471]
  ...
  [0.         0.         0.01176471]
  [0.         0.         0.01176471]
  [0.         0.         0.01176471]]

[[[1.         1.         1.         ]
  [0.         0.         0.01568628]
  [0.         0.         0.01568628]
  ...
  [0.         0.         0.01568628]
  [0.         0.         0.01568628]
  [0.         0.         0.01568628]]

[[[1.         1.         1.         ]
  [0.         0.         0.01568628]
  [0.         0.         0.01568628]
  ...
  [0.         0.         0.01568628]
  [0.         0.         0.01568628]
  [0.         0.         0.01568628]]

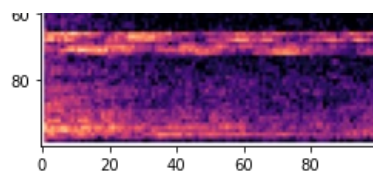
...

[[[1.         1.         1.         ]
  [0.80392164 0.2509804 0.4431373 ]
  [0.82745105 0.2627451 0.43137258]
  ...
  [0.19215688 0.06666667 0.39607847]
  [0.5137255  0.14901961 0.5058824 ]
  [0.5568628  0.16470589 0.5058824 ]]

[[[1.         1.         1.         ]
  [0.70980394 0.21176472 0.43529415]
  [0.20000002 0.05882353 0.39607847]
  ...
  [0.02745098 0.02352941 0.10980393]
  [0.06666667 0.04705883 0.18039216]
  [0.12156864 0.0627451  0.28627452]]

[[[1.         1.         1.         ]
  [1.         1.         1.         ]
  [1.         1.         1.         ]
  ...
  [1.         1.         1.         ]
  [1.         1.         1.         ]
  [1.         1.         1.         ]]]
[0. 0. 0. 0. 0. 0. 0. 0. 1. 0.]
```





```
(2, 100, 100, 3)
[[[1.          1.          1.          ]
  [0.          0.          0.01176471]
  [0.          0.          0.01176471]
  ...
  [0.          0.          0.01176471]
  [0.          0.          0.01176471]
  [0.          0.          0.01176471]]

  [[1.          1.          1.          ]
  [0.          0.          0.01568628]
  [0.          0.          0.01568628]
  ...
  [0.          0.          0.01568628]
  [0.03137255 0.02745098 0.11764707]
  [0.00784314 0.00784314 0.0509804  ]]

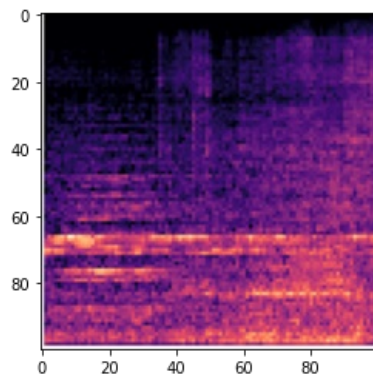
  [[1.          1.          1.          ]
  [0.          0.          0.01568628]
  [0.          0.          0.01568628]
  ...
  [0.02352941 0.01960784 0.09411766]
  [0.02745098 0.02352941 0.10980393]
  [0.03137255 0.02745098 0.11764707]]

  ...

  [[1.          1.          1.          ]
  [0.8588236   0.2784314   0.4156863   ]
  [0.9176471   0.3372549   0.3803922   ]
  ...
  [0.8980393   0.3137255   0.3921569   ]
  [0.8352942   0.26666668   0.427451    ]
  [0.6          0.1764706   0.5019608   ]]

  [[1.          1.          1.          ]
  [0.53333336   0.15294118   0.47450984]
  [0.3921569    0.10196079   0.47450984]
  ...
  [0.49411768   0.14117648   0.47450984]
  [0.47450984   0.13333334   0.47450984]
  [0.5137255    0.14901961   0.47450984]]

  [[1.          1.          1.          ]
  [1.          1.          1.          ]
  [1.          1.          1.          ]
  ...
  [1.          1.          1.          ]
  [1.          1.          1.          ]
  [1.          1.          1.          ]]]
[0. 0. 0. 0. 0. 0. 0. 0. 0. 1. 0.]
```



```
(2, 100, 100, 3)
[[[1.          1.          ]
  [0.          0.          0.01176471]
  [0.          0.          0.01176471]
  ...
  [0.          0.          0.01176471]
  [0.          0.          0.01176471]
  [0.          0.          0.01176471]]

  [[1.          1.          ]
  [0.          0.          0.01568628]
  [0.          0.          0.01568628]
  ...
  [0.          0.          0.01568628]
  [0.          0.          0.01568628]]
```

```

[0.      0.      0.01568628]]

[[1.      1.      1.      ]
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]
 ...
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]]

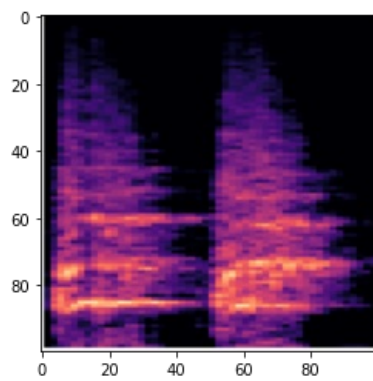
...

[[1.      1.      1.      ]
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]
 ...
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]]

[[1.      1.      1.      ]
 [0.      0.      0.01176471]
 [0.      0.      0.01176471]
 ...
 [0.03529412 0.02745098 0.1254902 ]
 [0.      0.      0.01176471]
 [0.      0.      0.01176471]]

[[1.      1.      1.      ]
 [1.      1.      1.      ]
 [1.      1.      1.      ]
 ...
 [1.      1.      1.      ]
 [1.      1.      1.      ]
 [1.      1.      1.      ]]]
[0. 0. 0. 1. 0. 0. 0. 0. 0. 0.]

```



```

(2, 100, 100, 3)
[[1.      1.      1.      ]
 [0.      0.      0.01176471]
 [0.      0.      0.01176471]
 ...
 [0.      0.      0.01176471]
 [0.      0.      0.01176471]
 [0.      0.      0.01176471]]

[[1.      1.      1.      ]
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]
 ...
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]]

[[1.      1.      1.      ]
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]
 ...
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]]

...

[[1.      1.      1.      ]
 [0.26666668 0.05882353 0.46274513]
 [0.      0.      0.01568628]
 ...
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]
 [0.      0.      0.01568628]]

[[1.      1.      1.      ]
 [0.22352943 0.05490196 0.4156863 ]

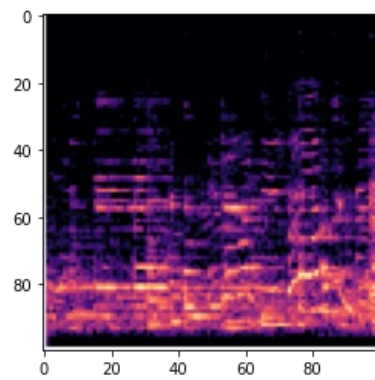
```

```

[0.      0.      0.01176471]
...
[0.      0.      0.01176471]
[0.      0.      0.01176471]
[0.      0.      0.01176471]]

[[1.      1.      1.      ]
 [1.      1.      1.      ]
 [1.      1.      1.      ]
 ...
 [1.      1.      1.      ]
 [1.      1.      1.      ]
 [1.      1.      1.      ]]]
[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]

```



```

(2, 100, 100, 3)
[[[1.      1.      1.      ]
  [0.      0.      0.01176471]
  [0.      0.      0.01176471]
  ...
  [0.      0.      0.01176471]
  [0.      0.      0.01176471]
  [0.      0.      0.01176471]]

 [[1.      1.      1.      ]
  [0.      0.      0.01568628]
  [0.      0.      0.01568628]
  ...
  [0.      0.      0.01568628]
  [0.      0.      0.01568628]
  [0.      0.      0.01568628]]

 [[1.      1.      1.      ]
  [0.      0.      0.01568628]
  [0.      0.      0.01568628]
  ...
  [0.      0.      0.01568628]
  [0.      0.      0.01568628]
  [0.      0.      0.01568628]]

 ...

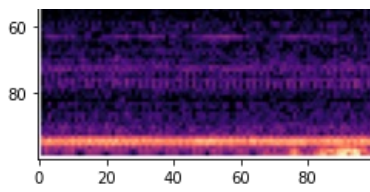
 [[1.      1.      1.      ]
  [0.39607847 0.10196079 0.5019608 ]
  [0.7490196  0.227451  0.4666667 ]
  ...
  [0.97647065 0.47450984 0.3647059 ]
  [0.79215693 0.24313727 0.44705886]
  [0.65882355 0.19607845 0.4901961 ]]

 [[1.      1.      1.      ]
  [0.31764707 0.07058824 0.46274513]
  [0.6039216  0.1764706  0.4666667 ]
  ...
  [0.86274517 0.30980393 0.36078432]
  [0.9058824  0.39607847 0.3372549 ]
  [0.8235295  0.27058825 0.38431376]]

 [[1.      1.      1.      ]
  [1.      1.      1.      ]
  [1.      1.      1.      ]
  ...
  [1.      1.      1.      ]
  [1.      1.      1.      ]
  [1.      1.      1.      ]]]
[0. 0. 0. 0. 0. 1. 0. 0. 0. 0.]

```





```
In [4]: import numpy as np
import pylab as pl
from keras import backend as K
import matplotlib.pyplot as plt
from keras.utils import np_utils
from keras.models import Sequential
from keras.layers.convolutional import Conv2D, MaxPooling2D
from keras.layers.core import Dense, Dropout, Activation, Flatten
from tensorflow.keras import layers, activations
```

```
In [5]: model = Sequential() #Sequential modeli oluşturduk.(Boş)

model.add(layers.Conv2D(filters = 4, activation = 'relu', kernel_size = (5,5), input_shape = (100,100,3))) #Oluşturulan
model.add(layers.MaxPooling2D(2,2)) #Üstteki işlemden sonra pooling yaptık.Filtreleri arttırarak bağlantıları güç

model.add(layers.Conv2D(filters = 8, activation = 'elu', kernel_size = (3,3)))
model.add(layers.MaxPooling2D(2,2))

model.add(layers.Conv2D(filters = 16, activation = 'elu', kernel_size = (2,2)))
model.add(layers.MaxPooling2D(2,2))

model.add(layers.Conv2D(filters = 32, activation = 'elu', kernel_size = (2,2)))

model.add(layers.Flatten()) #CNN in son aşaması olarak düzleştirme yaptık.

model.add(layers.Dense(50, activation="relu")) #Oluşturulan ağırları dense yaptık ve son outputta 10 adet sınıf istediği
model.add(layers.Dense(100, activation="relu"))
model.add(layers.Dense(100, activation="relu"))
model.add(layers.Dense(50, activation="relu"))
model.add(layers.Dense(10, activation="softmax"))
```

```
In [6]: model.summary() #Oluşturulan modelin özetini çıkardık.
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 96, 96, 4)	304
max_pooling2d (MaxPooling2D)	(None, 48, 48, 4)	0
conv2d_1 (Conv2D)	(None, 46, 46, 8)	296
max_pooling2d_1 (MaxPooling2D)	(None, 23, 23, 8)	0
conv2d_2 (Conv2D)	(None, 22, 22, 16)	528
max_pooling2d_2 (MaxPooling2D)	(None, 11, 11, 16)	0
conv2d_3 (Conv2D)	(None, 10, 10, 32)	2080
flatten (Flatten)	(None, 3200)	0
dense (Dense)	(None, 50)	160050
dense_1 (Dense)	(None, 100)	5100
dense_2 (Dense)	(None, 100)	10100

dense_3 (Dense)	(None, 50)	5050
dense_4 (Dense)	(None, 10)	510

=====
Total params: 184,018
Trainable params: 184,018
Non-trainable params: 0
=====

In [7]: `#Model Eğitimi`

In [8]: `import tensorflow as tf
optimizer = tf.keras.optimizers.Adamax(learning_rate = 0.001) #Optimizer seçtik ve learning_rate değeri belirledik
loss = tf.keras.losses.CategoricalCrossentropy()`

In [9]: `model.compile(optimizer = optimizer, loss = loss, metrics = ["mse","accuracy"], run_eagerly=True) #compiler ettik`

In [10]: `result = model.fit(train_datagen, epochs= 5, verbose = 1, validation_data = test_datagen) #Modeli test ettik`

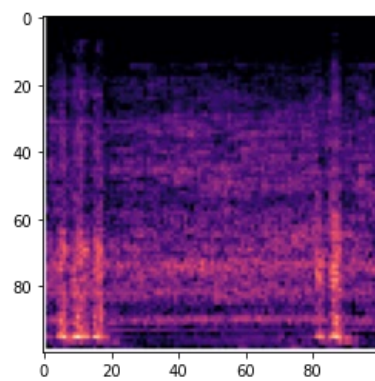
Epoch 1/5
3931/3931 [=====] - 142s 36ms/step - loss: 1.4909 - mse: 0.0658 - accuracy: 0.4633 - val_loss: 1.2512 - val_mse: 0.0587 - val_accuracy: 0.5476
Epoch 2/5
3931/3931 [=====] - 133s 34ms/step - loss: 0.9482 - mse: 0.0439 - accuracy: 0.6776 - val_loss: 1.2788 - val_mse: 0.0596 - val_accuracy: 0.5522
Epoch 3/5
3931/3931 [=====] - 134s 34ms/step - loss: 0.7010 - mse: 0.0325 - accuracy: 0.7701 - val_loss: 1.1812 - val_mse: 0.0515 - val_accuracy: 0.6039
Epoch 4/5
3931/3931 [=====] - 134s 34ms/step - loss: 0.5422 - mse: 0.0253 - accuracy: 0.8223 - val_loss: 1.4011 - val_mse: 0.0568 - val_accuracy: 0.5913
Epoch 5/5
3931/3931 [=====] - 134s 34ms/step - loss: 0.4293 - mse: 0.0204 - accuracy: 0.8601 - val_loss: 1.4353 - val_mse: 0.0576 - val_accuracy: 0.5924

In [11]: `model.evaluate(test_datagen) # Loss mse ve accuracy değerlerini kısa bir testte gördük`

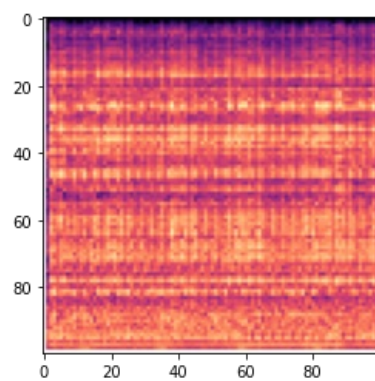
Out[11]: `436/436 [=====] - 8s 19ms/step - loss: 1.4353 - mse: 0.0576 - accuracy: 0.5924
[1.435295820236206, 0.0576358400285244, 0.5924224853515625]`

In [12]: `print(test_datagen.class_indices) #Baştaki pathtten 100 tane resim seçip sınıf değerlerine göre sonuç vermesini istedik
for i in range(100):
 img,label = test_datagen.next()
 a = model.predict(img)
 np.argmax(a[0])
 plt.imshow(img[0])
 if np.argmax(a[0]) == 0:
 print("air_conditioner")
 if np.argmax(a[0]) == 1:
 print("car_horn")
 if np.argmax(a[0]) == 2:
 print("children_playing")
 if np.argmax(a[0]) == 3:
 print("dog_bark")
 if np.argmax(a[0]) == 4:
 print('drilling')
 if np.argmax(a[0]) == 5:
 print("engine_idling")
 if np.argmax(a[0]) == 6:
 print("gun_shot")
 if np.argmax(a[0]) == 7:
 print("jackhammer")
 if np.argmax(a[0]) == 8:
 print("siren")
 if np.argmax(a[0]) == 9:
 print("street_music")
 plt.show()`

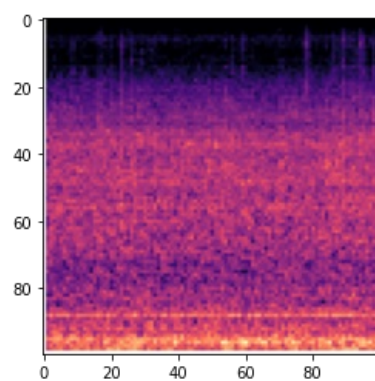
{'0': 0, '1': 1, '2': 2, '3': 3, '4': 4, '5': 5, '6': 6, '7': 7, '8': 8, '9': 9}
1/1 [=====] - 0s 48ms/step
drilling



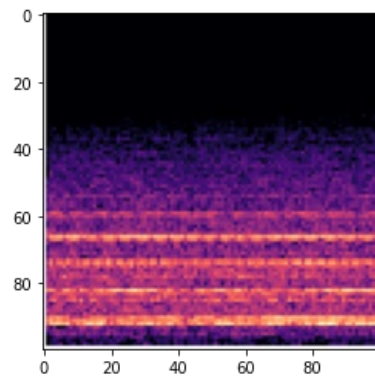
1/1 [=====] - 0s 31ms/step
jackhammer



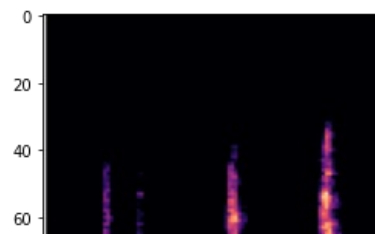
1/1 [=====] - 0s 31ms/step
engine_idling

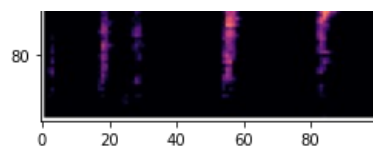


1/1 [=====] - 0s 31ms/step
car_horn

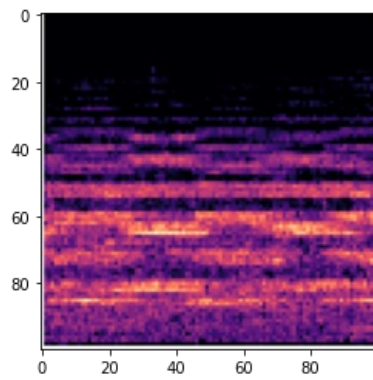


1/1 [=====] - 0s 29ms/step
dog_bark

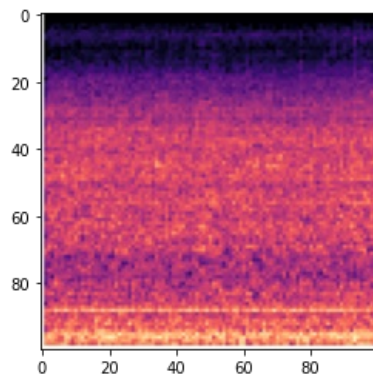




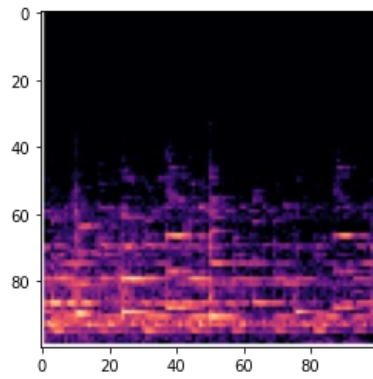
1/1 [=====] - 0s 29ms/step
street_music



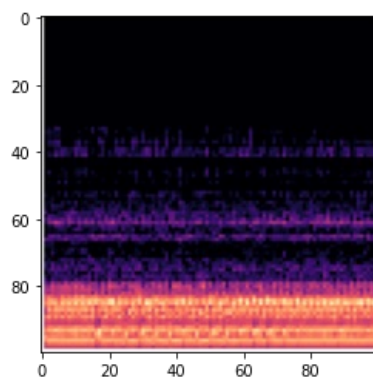
1/1 [=====] - 0s 31ms/step
engine_idling



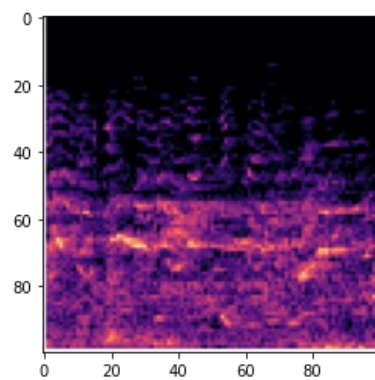
1/1 [=====] - 0s 30ms/step
street_music



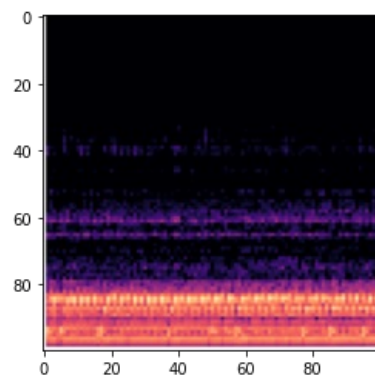
1/1 [=====] - 0s 30ms/step
street_music



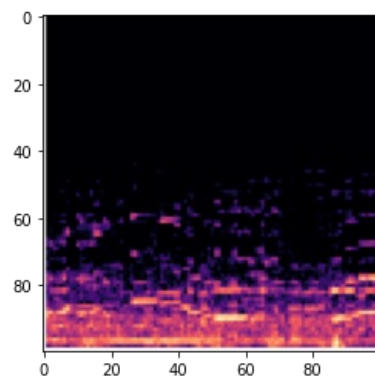
1/1 [=====] - 0s 30ms/step
siren



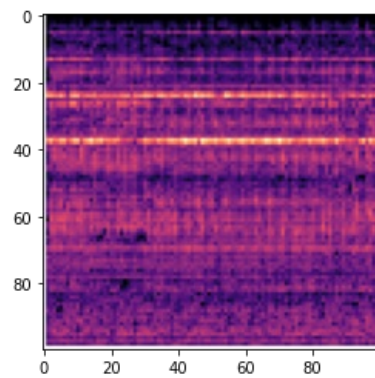
1/1 [=====] - 0s 30ms/step
street_music



1/1 [=====] - 0s 33ms/step
street_music

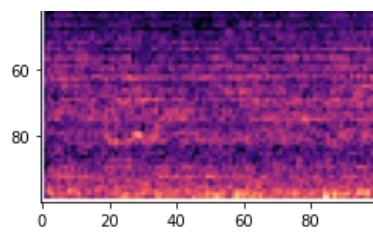


1/1 [=====] - 0s 29ms/step
jackhammer

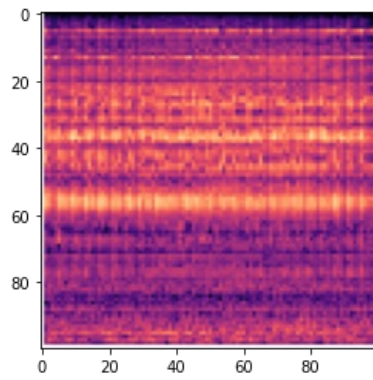


1/1 [=====] - 0s 29ms/step
siren

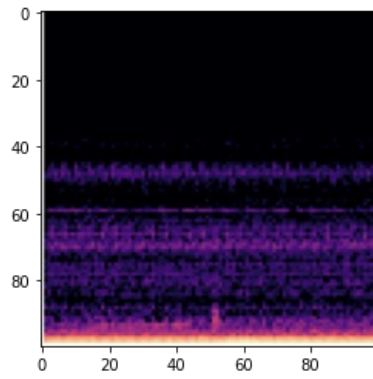




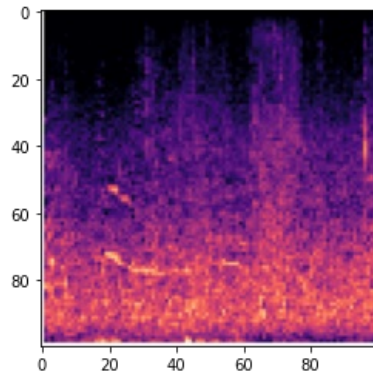
1/1 [=====] - 0s 31ms/step
jackhammer



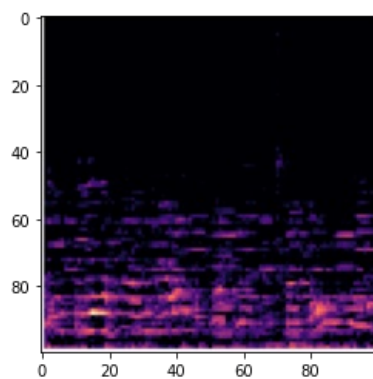
1/1 [=====] - 0s 30ms/step
engine_idling



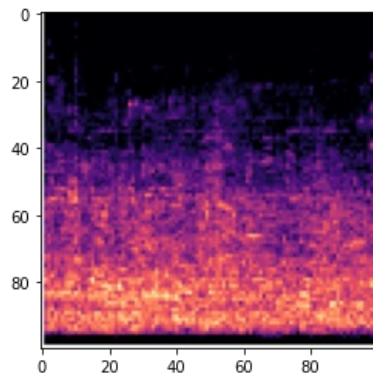
1/1 [=====] - 0s 29ms/step
drilling



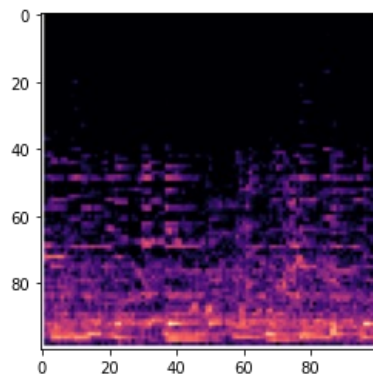
1/1 [=====] - 0s 29ms/step
street_music



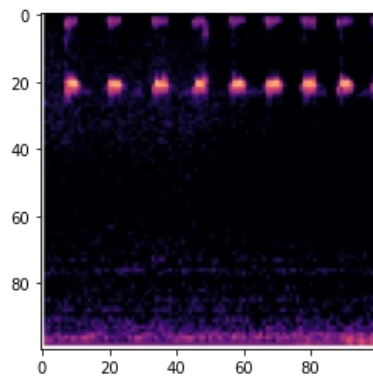
1/1 [=====] - 0s 32ms/step
street_music



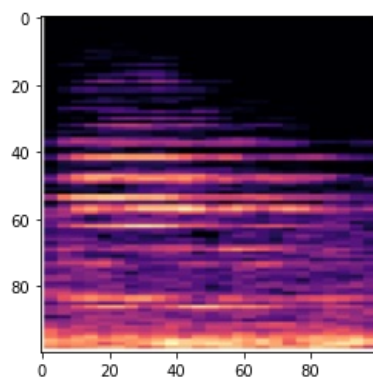
1/1 [=====] - 0s 29ms/step
street_music



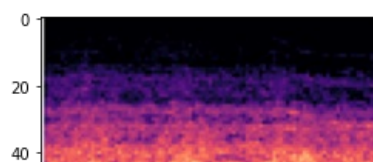
1/1 [=====] - 0s 32ms/step
siren

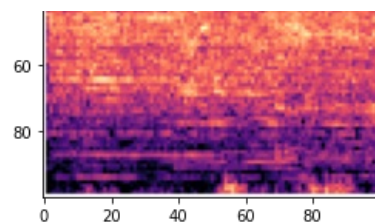


1/1 [=====] - 0s 30ms/step
car_horn

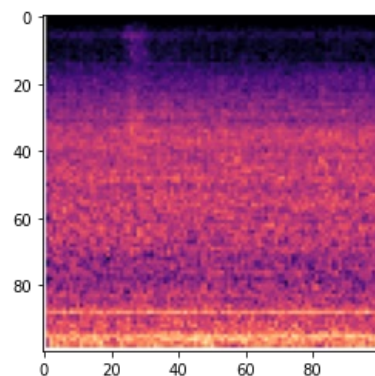


1/1 [=====] - 0s 30ms/step
drilling

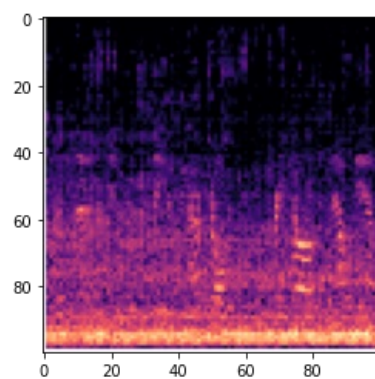




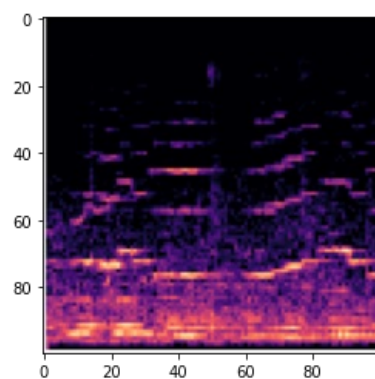
1/1 [=====] - 0s 31ms/step
engine_idling



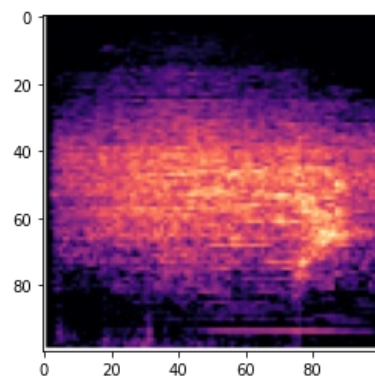
1/1 [=====] - 0s 33ms/step
siren



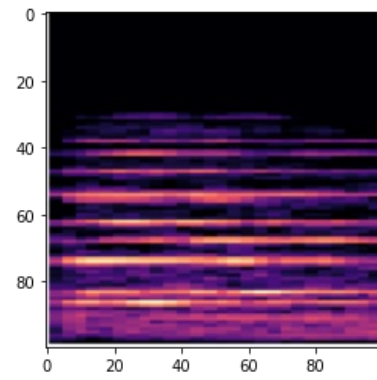
1/1 [=====] - 0s 34ms/step
street_music



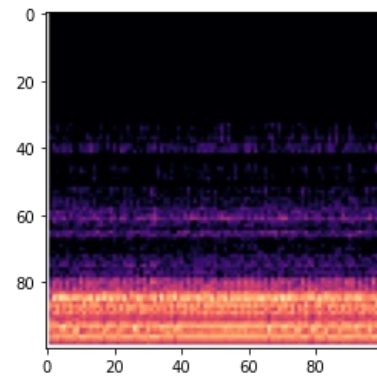
1/1 [=====] - 0s 32ms/step
drilling



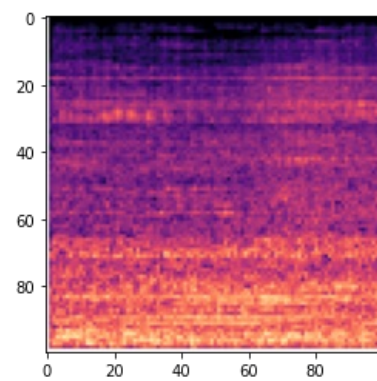
1/1 [=====] - 0s 29ms/step
car_horn



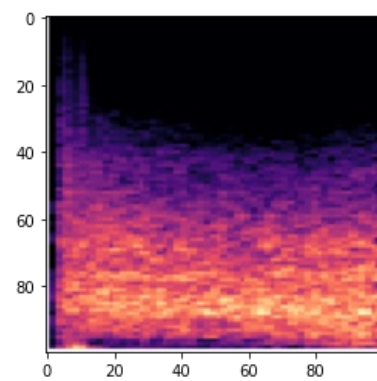
1/1 [=====] - 0s 31ms/step
engine_idling



1/1 [=====] - 0s 30ms/step
jackhammer

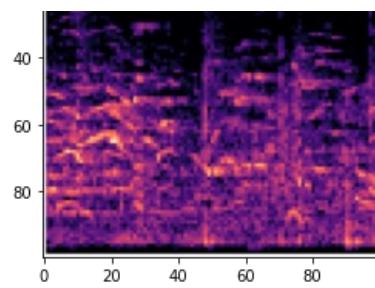


1/1 [=====] - 0s 33ms/step
gun_shot

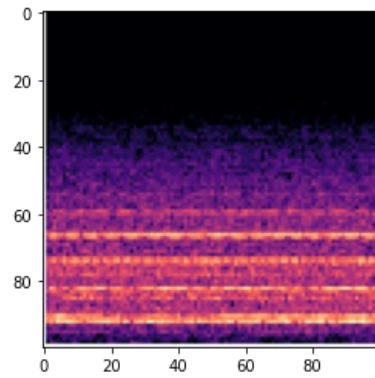


1/1 [=====] - 0s 32ms/step
children_playing

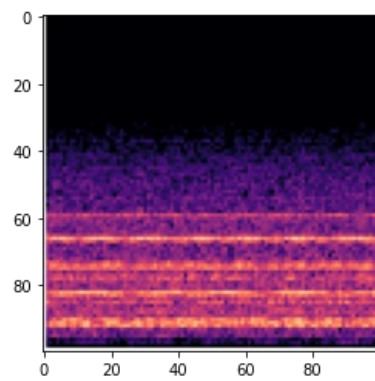




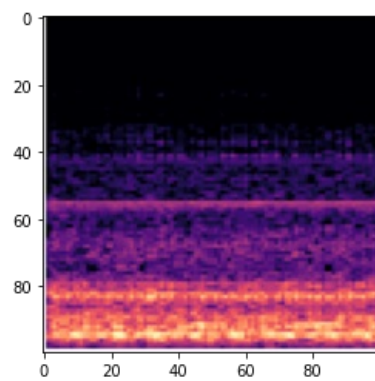
1/1 [=====] - 0s 31ms/step
car_horn



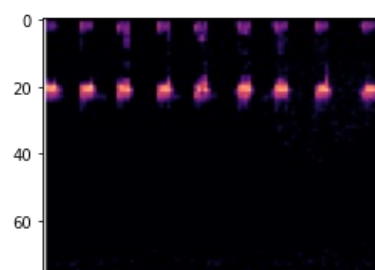
1/1 [=====] - 0s 30ms/step
car_horn

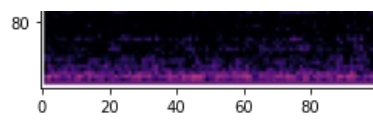


1/1 [=====] - 0s 32ms/step
street_music

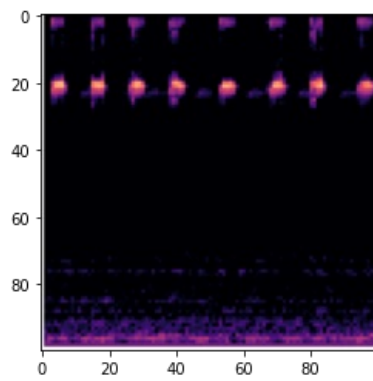


1/1 [=====] - 0s 29ms/step
siren

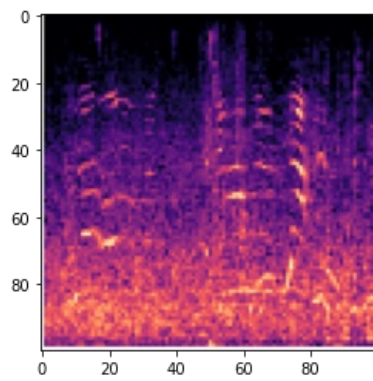




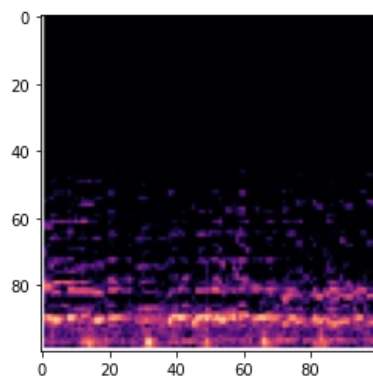
1/1 [=====] - 0s 31ms/step
siren



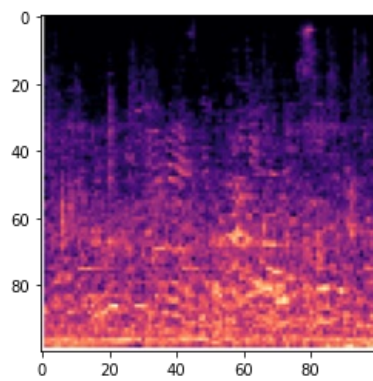
1/1 [=====] - 0s 30ms/step
children_playing



1/1 [=====] - 0s 32ms/step
street_music

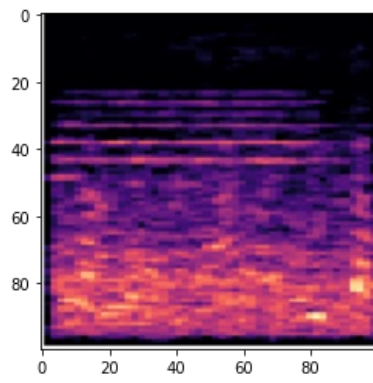


1/1 [=====] - 0s 29ms/step
street_music

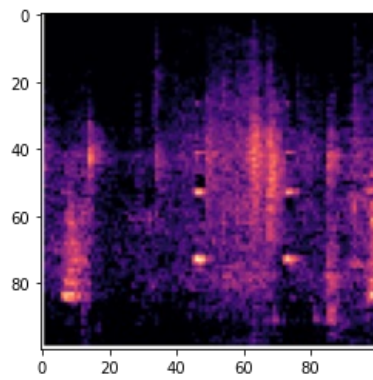


1/1 [=====] - 0s 26ms/step

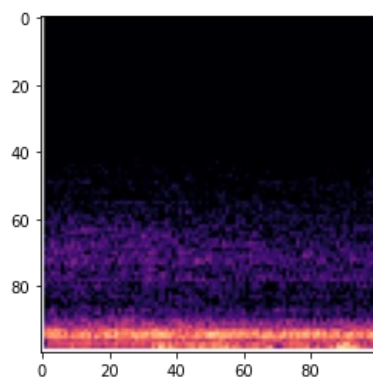
street_music



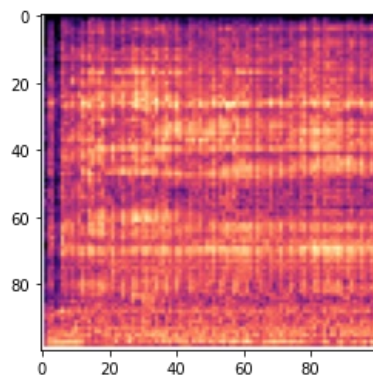
1/1 [=====] - 0s 25ms/step
dog_bark



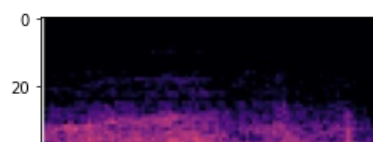
1/1 [=====] - 0s 22ms/step
siren

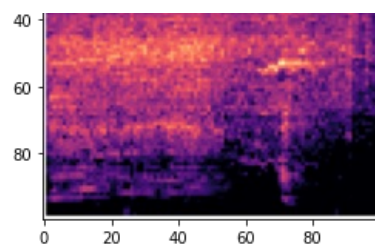


1/1 [=====] - 0s 23ms/step
drilling

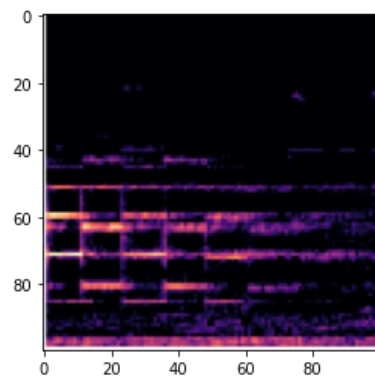


1/1 [=====] - 0s 24ms/step
drilling

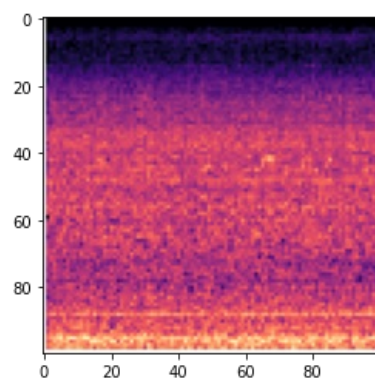




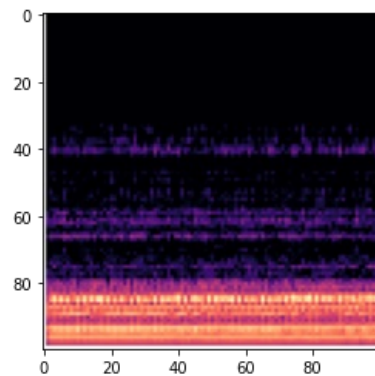
1/1 [=====] - 0s 22ms/step
siren



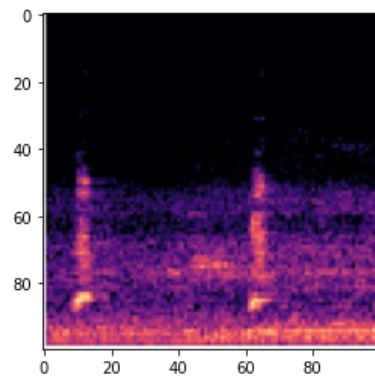
1/1 [=====] - 0s 22ms/step
engine_idling



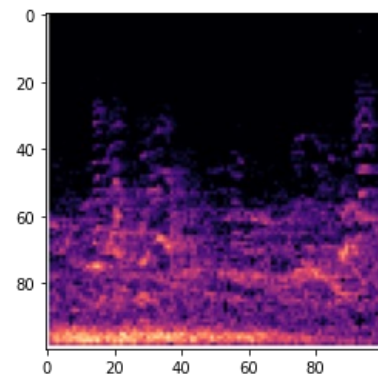
1/1 [=====] - 0s 25ms/step
car_horn



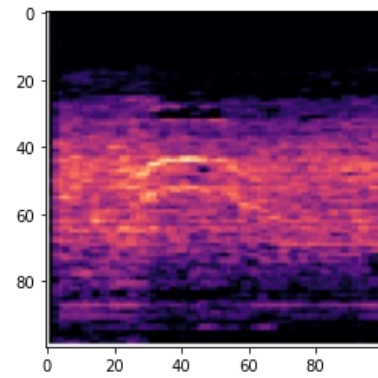
1/1 [=====] - 0s 23ms/step
dog_bark



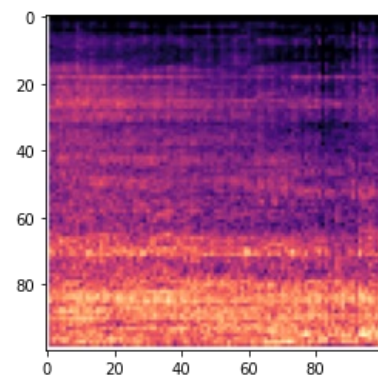
1/1 [=====] - 0s 25ms/step
children_playing



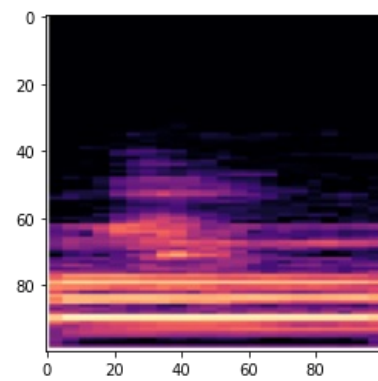
1/1 [=====] - 0s 24ms/step
children_playing



1/1 [=====] - 0s 24ms/step
air_conditioner

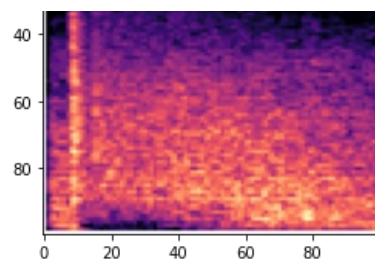


1/1 [=====] - 0s 26ms/step
jackhammer

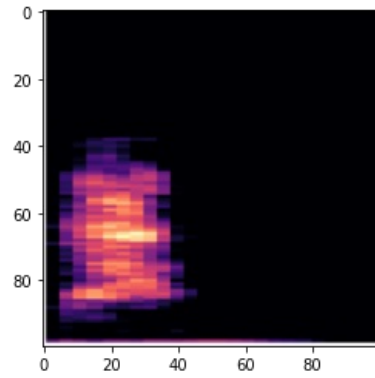


1/1 [=====] - 0s 25ms/step
gun_shot

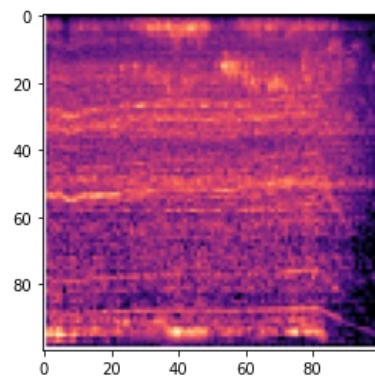




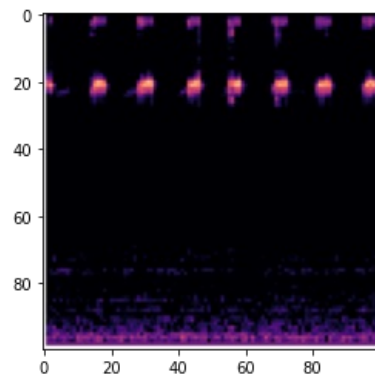
1/1 [=====] - 0s 22ms/step
dog_bark



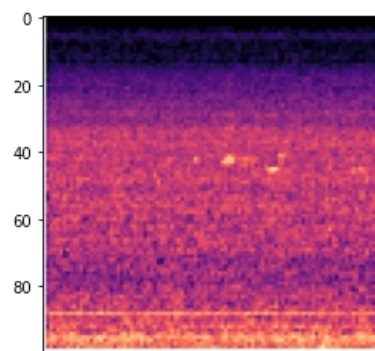
1/1 [=====] - 0s 24ms/step
drilling



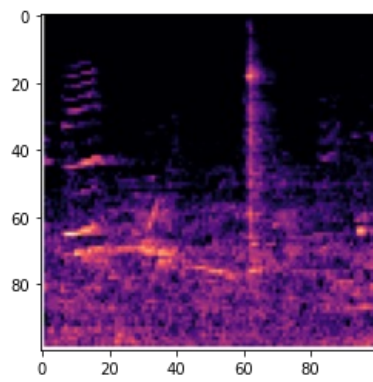
1/1 [=====] - 0s 23ms/step
dog_bark



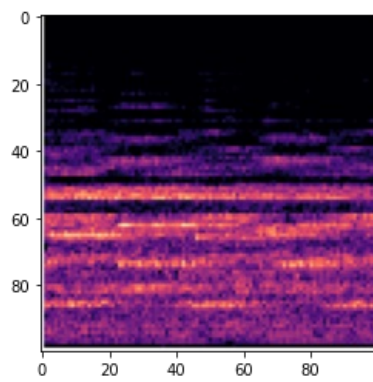
1/1 [=====] - 0s 23ms/step
engine_idling



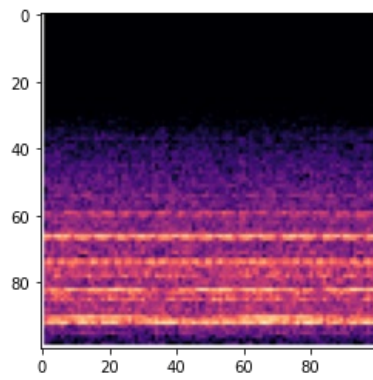
0 20 40 60 80
1/1 [=====] - 0s 24ms/step
dog_bark



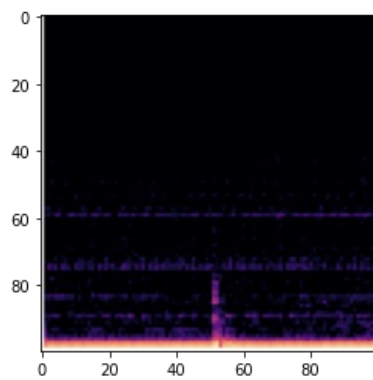
1/1 [=====] - 0s 22ms/step
siren



1/1 [=====] - 0s 24ms/step
car_horn

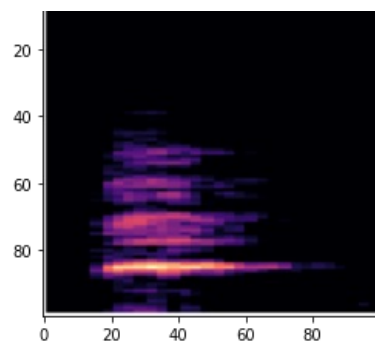


1/1 [=====] - 0s 26ms/step
children_playing

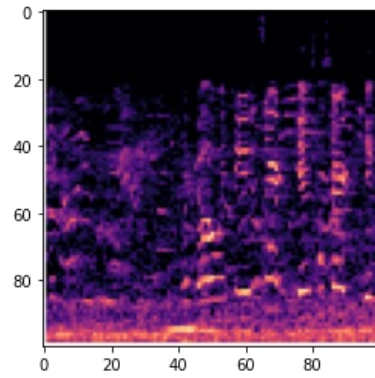


1/1 [=====] - 0s 22ms/step
dog_bark

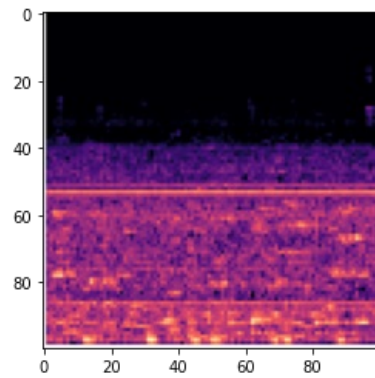




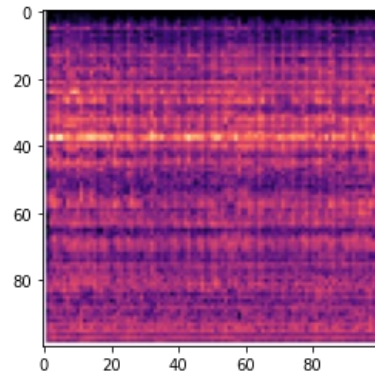
1/1 [=====] - 0s 22ms/step
children_playing



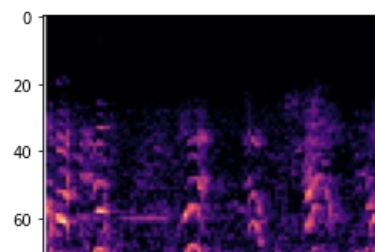
1/1 [=====] - 0s 23ms/step
air_conditioner

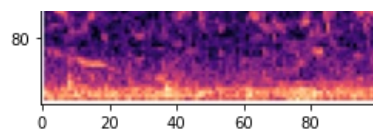


1/1 [=====] - 0s 22ms/step
drilling

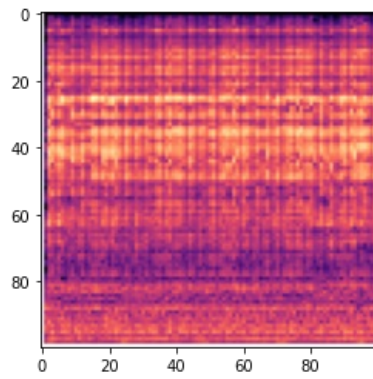


1/1 [=====] - 0s 24ms/step
children_playing

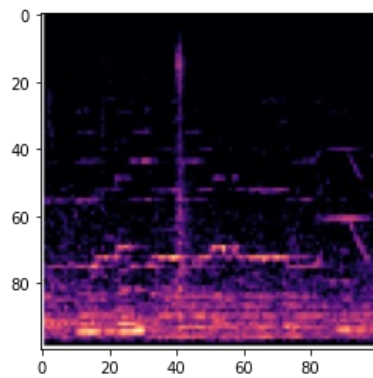




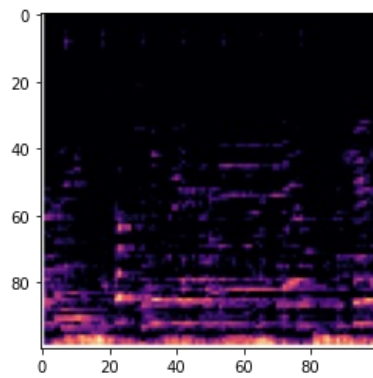
1/1 [=====] - 0s 23ms/step
jackhammer



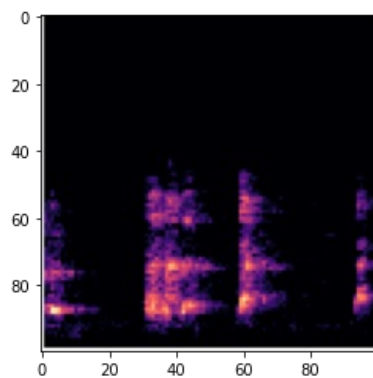
1/1 [=====] - 0s 22ms/step
children_playing



1/1 [=====] - 0s 25ms/step
street_music

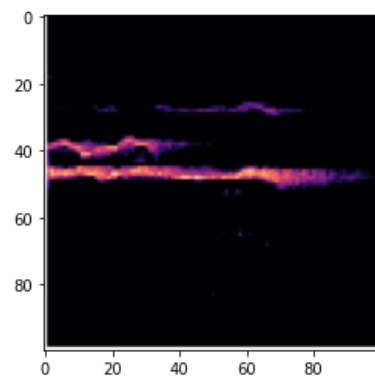


1/1 [=====] - 0s 23ms/step
dog_bark

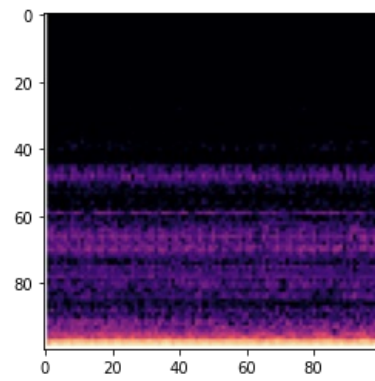


1/1 [=====] - 0s 24ms/step

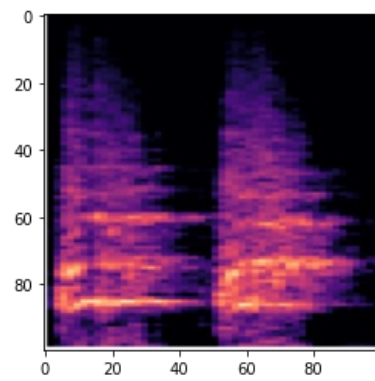
children_playing



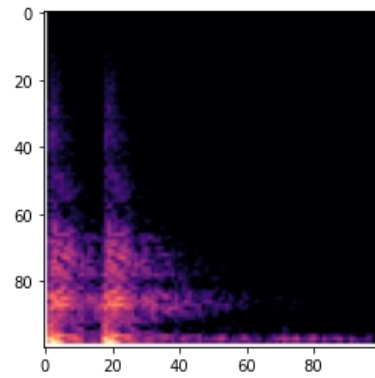
1/1 [=====] - 0s 25ms/step
children_playing



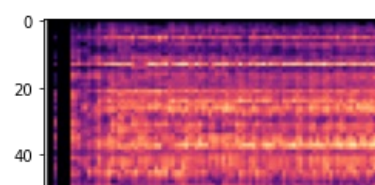
1/1 [=====] - 0s 22ms/step
dog_bark

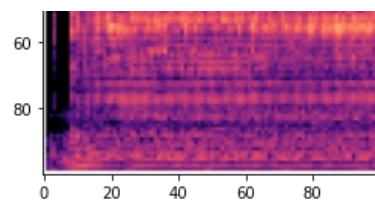


1/1 [=====] - 0s 22ms/step
gun_shot

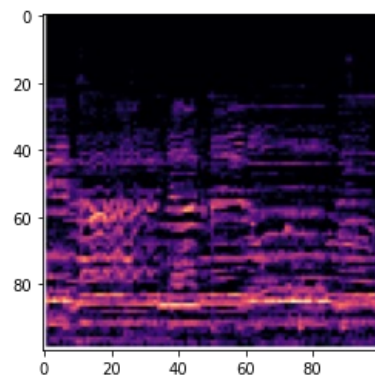


1/1 [=====] - 0s 24ms/step
jackhammer

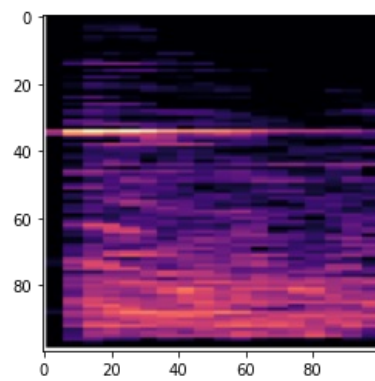




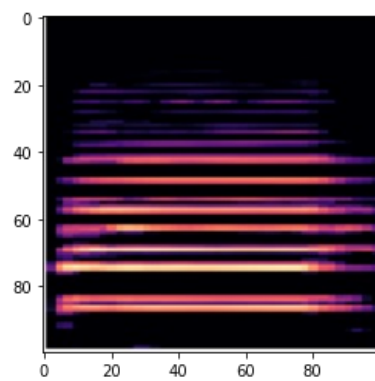
1/1 [=====] - 0s 23ms/step
street_music



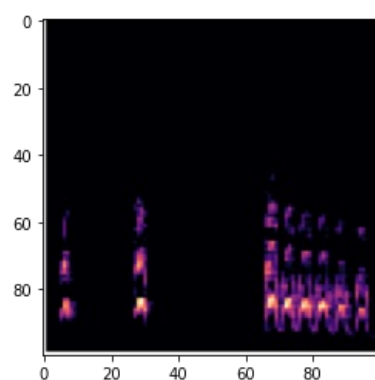
1/1 [=====] - 0s 24ms/step
car_horn



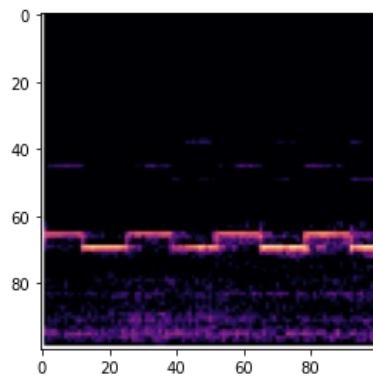
1/1 [=====] - 0s 22ms/step
car_horn



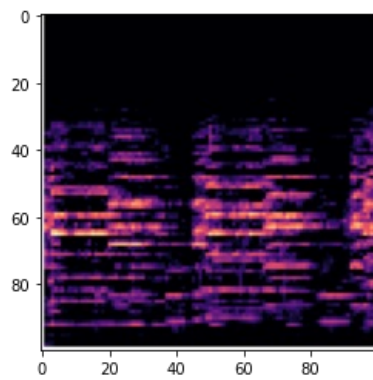
1/1 [=====] - 0s 23ms/step
dog_bark



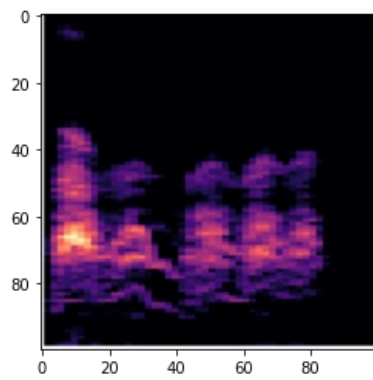
1/1 [=====] - 0s 23ms/step
siren



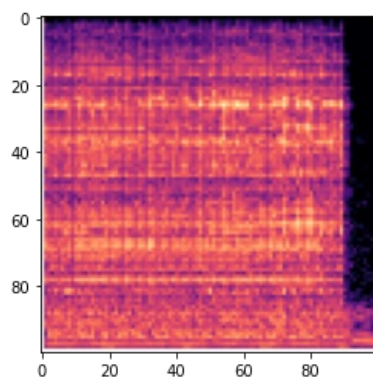
1/1 [=====] - 0s 22ms/step
siren



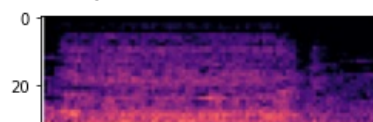
1/1 [=====] - 0s 23ms/step
dog_bark

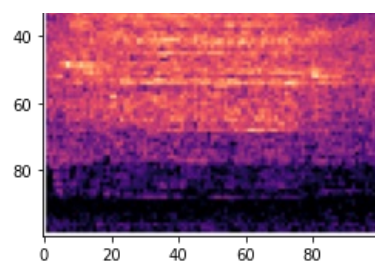


1/1 [=====] - 0s 22ms/step
jackhammer

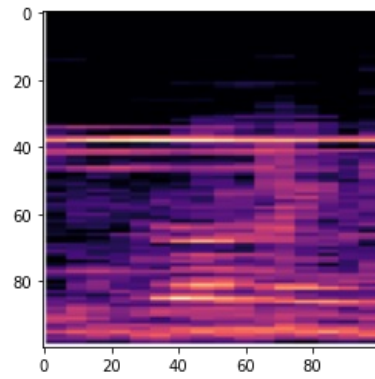


1/1 [=====] - 0s 23ms/step
drilling

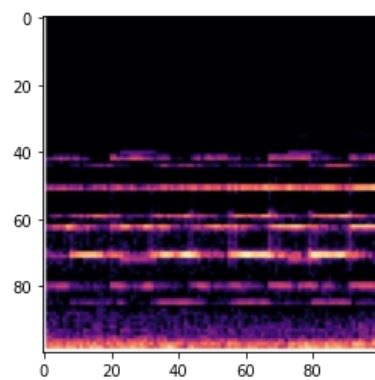




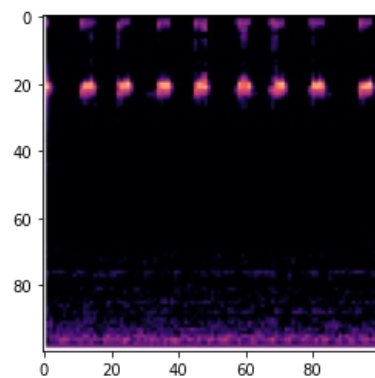
1/1 [=====] - 0s 23ms/step
car_horn



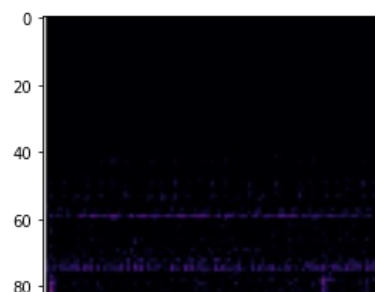
1/1 [=====] - 0s 22ms/step
siren

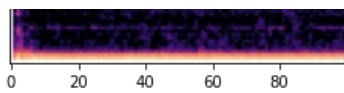


1/1 [=====] - 0s 29ms/step
siren

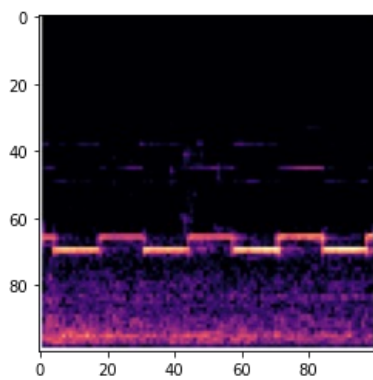


1/1 [=====] - 0s 28ms/step
air_conditioner

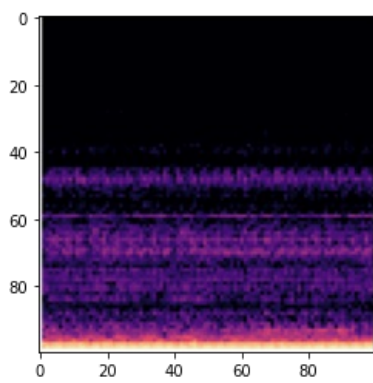




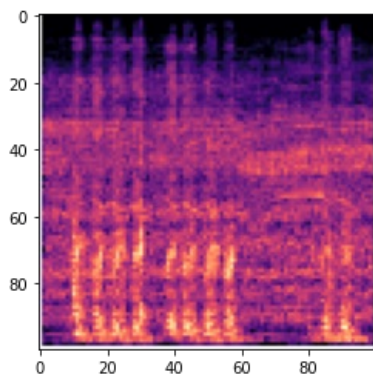
1/1 [=====] - 0s 26ms/step
siren



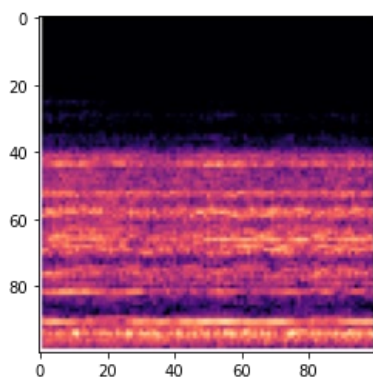
1/1 [=====] - 0s 28ms/step
children_playing



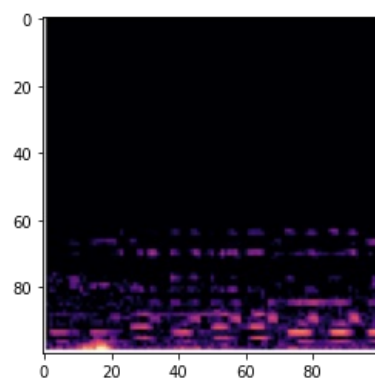
1/1 [=====] - 0s 29ms/step
drilling



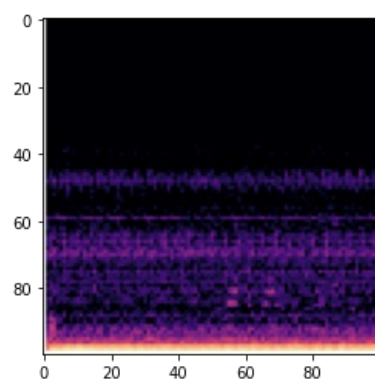
1/1 [=====] - 0s 22ms/step
siren



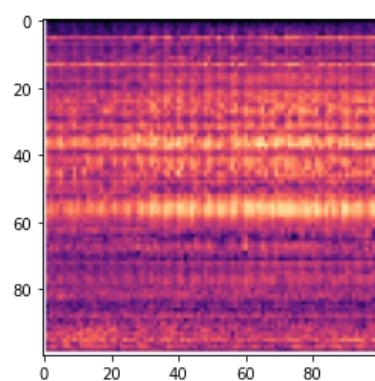
1/1 [=====] - 0s 23ms/step
street_music



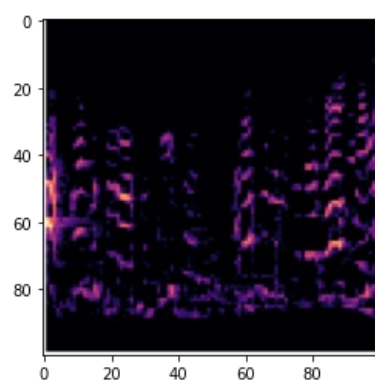
1/1 [=====] - 0s 25ms/step
engine_idling



1/1 [=====] - 0s 21ms/step
jackhammer

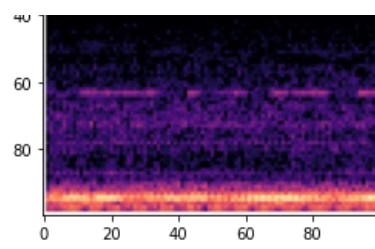


1/1 [=====] - 0s 24ms/step
children_playing

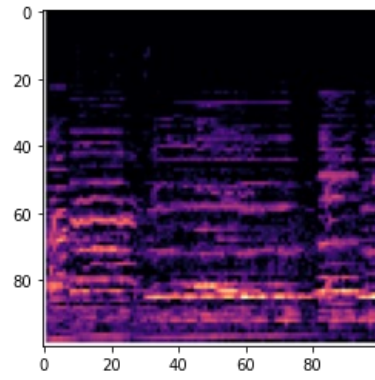


1/1 [=====] - 0s 24ms/step
siren





1/1 [=====] - 0s 22ms/step
street_music



1/1 [=====] - 0s 26ms/step
gun_shot

