

In [3]:

```
import numpy as np
import matplotlib.pyplot as plt
import os

from tensorflow import keras
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D,Dense,MaxPooling2D,Activation,Dropout,BatchNormaliza
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.utils import to_categorical
```

In [4]:

```
train = ImageDataGenerator(rescale=1/255)
test = ImageDataGenerator(rescale=1/255)
train_dataset = train.flow_from_directory(directory='C:/Users/HP/OneDrive/Desktop/seg_train')
test_dataset = test.flow_from_directory(directory='C:/Users/HP/OneDrive/Desktop/seg_test/se
```

Found 14034 images belonging to 6 classes.
Found 3000 images belonging to 6 classes.

In [5]:

```

indices = [np.random.randint(32) for i in range(10)]
print(indices)

plt.figure(figsize=(20,8))
for i in enumerate(indices):
    plt.subplot(2,5,i[0]+1)
    plt.imshow(train_dataset[0][0][i[1]])
    plt.title(train_dataset[0][1][i[1]])
plt.show()

```

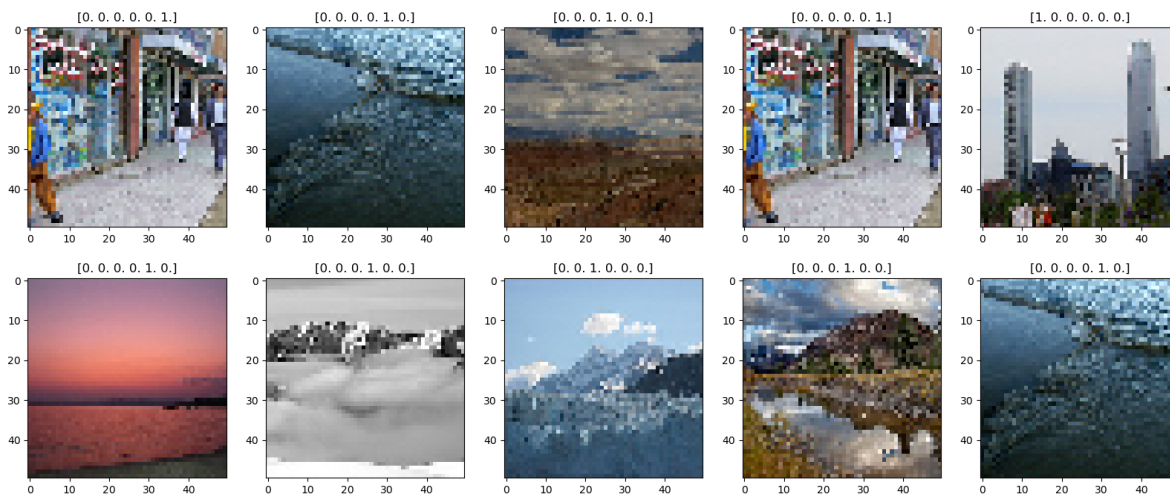
```
[1, 19, 18, 1, 27, 30, 10, 4, 17, 19]
```

C:\ProgramData\Anaconda3\lib\site-packages\matplotlib\text.py:1223: FutureWarning: elementwise comparison failed; returning scalar instead, but in the future will perform elementwise comparison

```

if s != self._text:

```



In [6]:

```

values = list(train_dataset.class_indices.values())
keys = list(train_dataset.class_indices.keys())

dics = list(map(lambda x,y:{x:y},values,keys))

from functools import reduce
mappings = reduce(lambda x,y:{**x,**y},dics)
mappings
#print(values)
#print(keys)
#print(dics)

```

Out[6]:

```

{0: 'buildings',
 1: 'forest',
 2: 'glacier',
 3: 'mountain',
 4: 'sea',
 5: 'street'}

```

In [7]:

```
model = Sequential()
model.add(Conv2D(filters=32,kernel_size=(3,3),padding='same',input_shape=(50,50,3)))
model.add(Activation('relu'))
model.add(Conv2D(filters=32,kernel_size=(3,3)))
model.add(Dropout(0.25))
model.add(Activation('relu'))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Conv2D(filters=50,kernel_size=(3,3),padding='same',input_shape=(50,50,3)))
model.add(Activation('relu'))
model.add(Conv2D(filters=50,kernel_size=(3,3)))
model.add(Dropout(0.25))
model.add(Activation('relu'))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Conv2D(filters=75,kernel_size=(3,3),padding='same',input_shape=(50,50,3)))
model.add(Activation('relu'))
model.add(Conv2D(filters=75,kernel_size=(3,3)))
model.add(Dropout(0.25))
model.add(Activation('relu'))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Flatten())
kernel_regularizer = keras.regularizers.l1_l2(l1=1e-5,l2=1e-4)
model.add(Dense(units=50,activation='relu',kernel_regularizer=kernel_regularizer))
model.add(Dense(50,activation='relu'))
model.add(Dropout(0.25))
model.add(Dense(6,activation='softmax'))

model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 50, 50, 32)	896
activation (Activation)	(None, 50, 50, 32)	0
conv2d_1 (Conv2D)	(None, 48, 48, 32)	9248
dropout (Dropout)	(None, 48, 48, 32)	0
activation_1 (Activation)	(None, 48, 48, 32)	0
max_pooling2d (MaxPooling2D)	(None, 24, 24, 32)	0
conv2d_2 (Conv2D)	(None, 24, 24, 50)	14450
activation_2 (Activation)	(None, 24, 24, 50)	0
conv2d_3 (Conv2D)	(None, 22, 22, 50)	22550
dropout_1 (Dropout)	(None, 22, 22, 50)	0

activation_3 (Activation)	(None, 22, 22, 50)	0
max_pooling2d_1 (MaxPooling 2D)	(None, 11, 11, 50)	0
conv2d_4 (Conv2D)	(None, 11, 11, 75)	33825
activation_4 (Activation)	(None, 11, 11, 75)	0
conv2d_5 (Conv2D)	(None, 9, 9, 75)	50700
dropout_2 (Dropout)	(None, 9, 9, 75)	0
activation_5 (Activation)	(None, 9, 9, 75)	0
max_pooling2d_2 (MaxPooling 2D)	(None, 4, 4, 75)	0
flatten (Flatten)	(None, 1200)	0
dense (Dense)	(None, 50)	60050
dense_1 (Dense)	(None, 50)	2550
dropout_3 (Dropout)	(None, 50)	0
dense_2 (Dense)	(None, 6)	306

```
=====
Total params: 194,575
Trainable params: 194,575
Non-trainable params: 0
```

In [8]:

```
model.compile(loss='CategoricalCrossentropy',optimizer = 'adam',metrics='accuracy')
history = model.fit(train_dataset,batch_size=80,epochs=2,validation_data=test_dataset)
```

Epoch 1/2

439/439 [=====] - 63s 141ms/step - loss: 1.1648 - accuracy: 0.5379 - val_loss: 1.0766 - val_accuracy: 0.6370

Epoch 2/2

439/439 [=====] - 70s 159ms/step - loss: 0.9480 - accuracy: 0.6298 - val_loss: 1.0103 - val_accuracy: 0.6287

In [9]:

```
plt.plot(history.history['accuracy'],label='accuracy')  
plt.plot(history.history['val_accuracy'],label='val_accuracy')  
plt.xlabel('Epoch')  
plt.ylabel('Accuracy')  
plt.legend(loc='lower right')
```

Out[9]:

<matplotlib.legend.Legend at 0x139a15a3ac0>

