

# cs2\_data\_pipeline

July 12, 2021

## 1 Libraries

```
[1]: import warnings
warnings.filterwarnings('ignore')
import os
import numpy as np
import pandas as pd
from tqdm import tqdm
pd.set_option("display.max_colwidth", -1)
import tensorflow as tf
import cv2
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
```

```
[2]: #seeding
seed = 2021
np.random.seed = seed
tf.seed = seed
```

```
[3]: #Creating these two folder
!mkdir train test
```

## 2 Data

```
[4]: #Unzipping the training and testing folders into directories
print('Unzipping stage1_train.zip')
!unzip -q "../input/data-science-bowl-2018/stage1_train.zip" -d train/
print('Unzipped stage1_train.zip')

print('Unzipping stage1_test.zip')
!unzip -q "../input/data-science-bowl-2018/stage1_test.zip" -d test/
print('Unzipped stage1_test.zip')
```

```
Unzipping stage1_train.zip
Unzipped stage1_train.zip
Unzipping stage1_test.zip
Unzipped stage1_test.zip
```

```
[5]: # Root directories for training and testing
TRAIN_ROOT = './train'
TEST_ROOT = './test'
```

```
[6]: # Function to create a dataframe of files which will be used for further
      ↪processing
def files_df(root_dir):
    subdir = os.listdir(root_dir)
    files = []
    df = pd.DataFrame()
    for dir in subdir:
        files.append(os.path.join(root_dir,dir))
    df['files'] = files
    return df
```

```
[7]: train_df = files_df(TRAIN_ROOT)
test_df = files_df(TEST_ROOT)
```

```
[8]: # Hyperparameters
IMG_WIDTH = 256
IMG_HEIGHT = 256
IMG_CHANNELS = 3
CLASSES = 1
BATCH_SIZE = 8
```

```
[9]: # Function which will create a dataframe of image paths and mask paths along
      ↪with creating a single mask with multiple masks
def image_df(filename):
    image_paths = []
    mask_paths = []
    df = pd.DataFrame()
    for filename in tqdm(filename):
        file_path = os.path.join(filename, 'images')
        image_path = os.path.join(file_path, os.listdir(file_path)[0])
        image_paths.append(image_path)

        mask = np.zeros((IMG_WIDTH, IMG_HEIGHT, 1))
        mask_dir = file_path.replace("images", "masks")
        masks = os.listdir(mask_dir)
        for m in masks:
            mask_path = os.path.join(mask_dir, m)
            mask_ = cv2.imread(mask_path, cv2.IMREAD_UNCHANGED)
            mask_ = cv2.resize(mask_, (IMG_WIDTH, IMG_HEIGHT), interpolation=cv2.
            ↪INTER_NEAREST)
            mask_ = np.expand_dims(mask_, axis = -1)
            mask = np.maximum(mask, mask_)
        newmask_dir = mask_dir.replace("masks", "masks_")
```

```

    if not os.path.isdir(newmask_dir):
        os.mkdir(newmask_dir)
    newmask_path = image_path.replace("images", "masks_")
    mask_paths.append(newmask_path)
    cv2.imwrite(newmask_path, mask)
df['images'] = image_paths
df['masks'] = mask_paths
return df

```

```

[10]: # Training dataframe
train_filenames = train_df['files']
train = image_df(train_filenames)

```

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### 3 Train Test Split

```

[11]: X_train, X_val = train_test_split(train, test_size=0.1, random_state=42)

```

```

[12]: X_train.head()

```

```

[12]:
                                     images \
300  ./train/df9a4212ecb67bb4e58eba62f293b91f9d6f1dde73e38fa891c75661d419fc97/im
ages/df9a4212ecb67bb4e58eba62f293b91f9d6f1dde73e38fa891c75661d419fc97.png
60   ./train/be1916d0e5592c17f971315b5de720ef6894173087399daed94a52ef109c1572/im
ages/be1916d0e5592c17f971315b5de720ef6894173087399daed94a52ef109c1572.png
133  ./train/3477024fd843e46097840360f9cdee24b76bf5c593ed27a9aee7a5728a06aa51/im
ages/3477024fd843e46097840360f9cdee24b76bf5c593ed27a9aee7a5728a06aa51.png
30   ./train/1609b1b8480ee52652a644403b3f7d5511410a016750aa3b9a4c8ddb3e893e8e/im
ages/1609b1b8480ee52652a644403b3f7d5511410a016750aa3b9a4c8ddb3e893e8e.png
69   ./train/648636ee314d7bdba3ab2fc0fe49a863de35c3e2caf619039f678df67b526868/im
ages/648636ee314d7bdba3ab2fc0fe49a863de35c3e2caf619039f678df67b526868.png

                                     masks
300  ./train/df9a4212ecb67bb4e58eba62f293b91f9d6f1dde73e38fa891c75661d419fc97/ma
sks_/df9a4212ecb67bb4e58eba62f293b91f9d6f1dde73e38fa891c75661d419fc97.png
60   ./train/be1916d0e5592c17f971315b5de720ef6894173087399daed94a52ef109c1572/ma
sks_/be1916d0e5592c17f971315b5de720ef6894173087399daed94a52ef109c1572.png
133  ./train/3477024fd843e46097840360f9cdee24b76bf5c593ed27a9aee7a5728a06aa51/ma
sks_/3477024fd843e46097840360f9cdee24b76bf5c593ed27a9aee7a5728a06aa51.png
30   ./train/1609b1b8480ee52652a644403b3f7d5511410a016750aa3b9a4c8ddb3e893e8e/ma
sks_/1609b1b8480ee52652a644403b3f7d5511410a016750aa3b9a4c8ddb3e893e8e.png
69   ./train/648636ee314d7bdba3ab2fc0fe49a863de35c3e2caf619039f678df67b526868/ma
sks_/648636ee314d7bdba3ab2fc0fe49a863de35c3e2caf619039f678df67b526868.png

```

```

[13]: X_val.head()

```

```
[13]:
images \
361 ./train/bf566e75d5cb0196de4139573f8bbbda0fa38d5048edf7267fe8793dcc094a66/im
ages/bf566e75d5cb0196de4139573f8bbbda0fa38d5048edf7267fe8793dcc094a66.png
158 ./train/cdab367b30db47061df837c1ae9fa875d6057614f797332d37d3513517d6c694/im
ages/cdab367b30db47061df837c1ae9fa875d6057614f797332d37d3513517d6c694.png
480 ./train/1023509cf8d4c155467800f89508690be9513431992f470594281cd37dbd020d/im
ages/1023509cf8d4c155467800f89508690be9513431992f470594281cd37dbd020d.png
640 ./train/b2c23ddb04531158da6a0abcaca78fec0ae5c6f64f60166e4f36f4a161efd76f/im
ages/b2c23ddb04531158da6a0abcaca78fec0ae5c6f64f60166e4f36f4a161efd76f.png
275 ./train/a22b7882fa85b9f0fcef659a7b82bfcddf01710f9a7617a9e036e84ac6901841/im
ages/a22b7882fa85b9f0fcef659a7b82bfcddf01710f9a7617a9e036e84ac6901841.png

masks
361 ./train/bf566e75d5cb0196de4139573f8bbbda0fa38d5048edf7267fe8793dcc094a66/ma
sks_/bf566e75d5cb0196de4139573f8bbbda0fa38d5048edf7267fe8793dcc094a66.png
158 ./train/cdab367b30db47061df837c1ae9fa875d6057614f797332d37d3513517d6c694/ma
sks_/cdab367b30db47061df837c1ae9fa875d6057614f797332d37d3513517d6c694.png
480 ./train/1023509cf8d4c155467800f89508690be9513431992f470594281cd37dbd020d/ma
sks_/1023509cf8d4c155467800f89508690be9513431992f470594281cd37dbd020d.png
640 ./train/b2c23ddb04531158da6a0abcaca78fec0ae5c6f64f60166e4f36f4a161efd76f/ma
sks_/b2c23ddb04531158da6a0abcaca78fec0ae5c6f64f60166e4f36f4a161efd76f.png
275 ./train/a22b7882fa85b9f0fcef659a7b82bfcddf01710f9a7617a9e036e84ac6901841/ma
sks_/a22b7882fa85b9f0fcef659a7b82bfcddf01710f9a7617a9e036e84ac6901841.png
```

## 4 Data Preprocessing

```
[14]: # Function to parse image and mask file path and convert them into image and
↪mask
def parse_function(image_path, mask_path):
    image_string = tf.io.read_file(image_path)
    image = tf.image.decode_png(image_string, channels=IMG_CHANNELS)#
    image = tf.image.convert_image_dtype(image, tf.float32)
    image = tf.image.resize(image, [IMG_HEIGHT, IMG_WIDTH]) # height x width

    mask_string = tf.io.read_file(mask_path)
    mask = tf.image.decode_png(mask_string, channels=IMG_CHANNELS)#
    mask = tf.image.convert_image_dtype(mask, tf.float32)
    mask = tf.image.resize(mask, [IMG_HEIGHT, IMG_WIDTH])
    return image, mask
```

```
[15]: # Function to perform data augmentation
def train_preprocess(image, mask):
    a = tf.random.uniform(())
    if a<0.2:
        image = tf.image.random_flip_left_right(image)
        mask = tf.image.random_flip_left_right(mask)
```

```

elif a<0.4:
    image = tf.image.random_brightness(image, max_delta=32.0 / 255.0)
    mask = tf.image.random_brightness(mask, max_delta=32.0 / 255.0)
elif a<0.6:
    image = tf.image.random_saturation(image, lower=0.5, upper=1.5)
    mask = tf.image.random_saturation(mask, lower=0.5, upper=1.5)
elif a<0.8:
    image = tf.image.random_flip_up_down(image)
    mask = tf.image.random_flip_up_down(mask)
else:
    image = tf.image.random_hue(image, 0.2)
    mask = tf.image.random_hue(mask, 0.2)

image = tf.clip_by_value(image, 0.0, 1.0)
mask = tf.clip_by_value(mask, 0.0, 1.0)
return image, mask

```

```

[16]: # Training dataset
train_ds = tf.data.Dataset.from_tensor_slices((X_train['images'],
→X_train['masks']))
train_ds = train_ds.shuffle(X_train.shape[0])
train_ds = train_ds.map(parse_function, num_parallel_calls=tf.data.AUTOTUNE)
train_ds = train_ds.map(train_preprocess, num_parallel_calls=tf.data.AUTOTUNE)
train_ds = train_ds.batch(BATCH_SIZE)
train_ds = train_ds.prefetch(1)

```

```

[17]: # Validation dataset
val_ds = tf.data.Dataset.from_tensor_slices((X_val['images'], X_val['masks']))
val_ds = val_ds.shuffle(X_val.shape[0])
val_ds = val_ds.map(parse_function, num_parallel_calls=tf.data.AUTOTUNE)
val_ds = val_ds.batch(BATCH_SIZE)
val_ds = val_ds.prefetch(1)

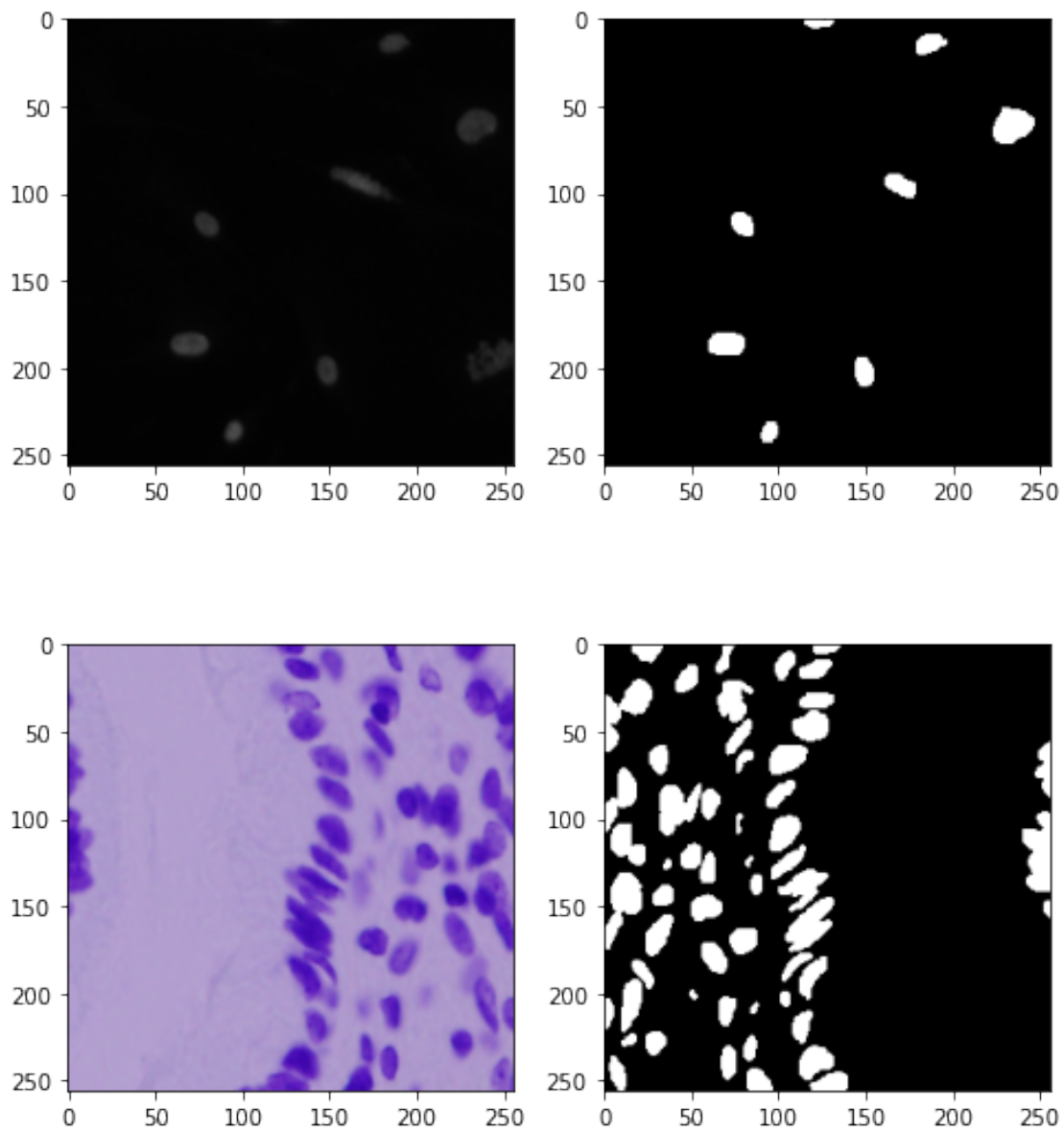
```

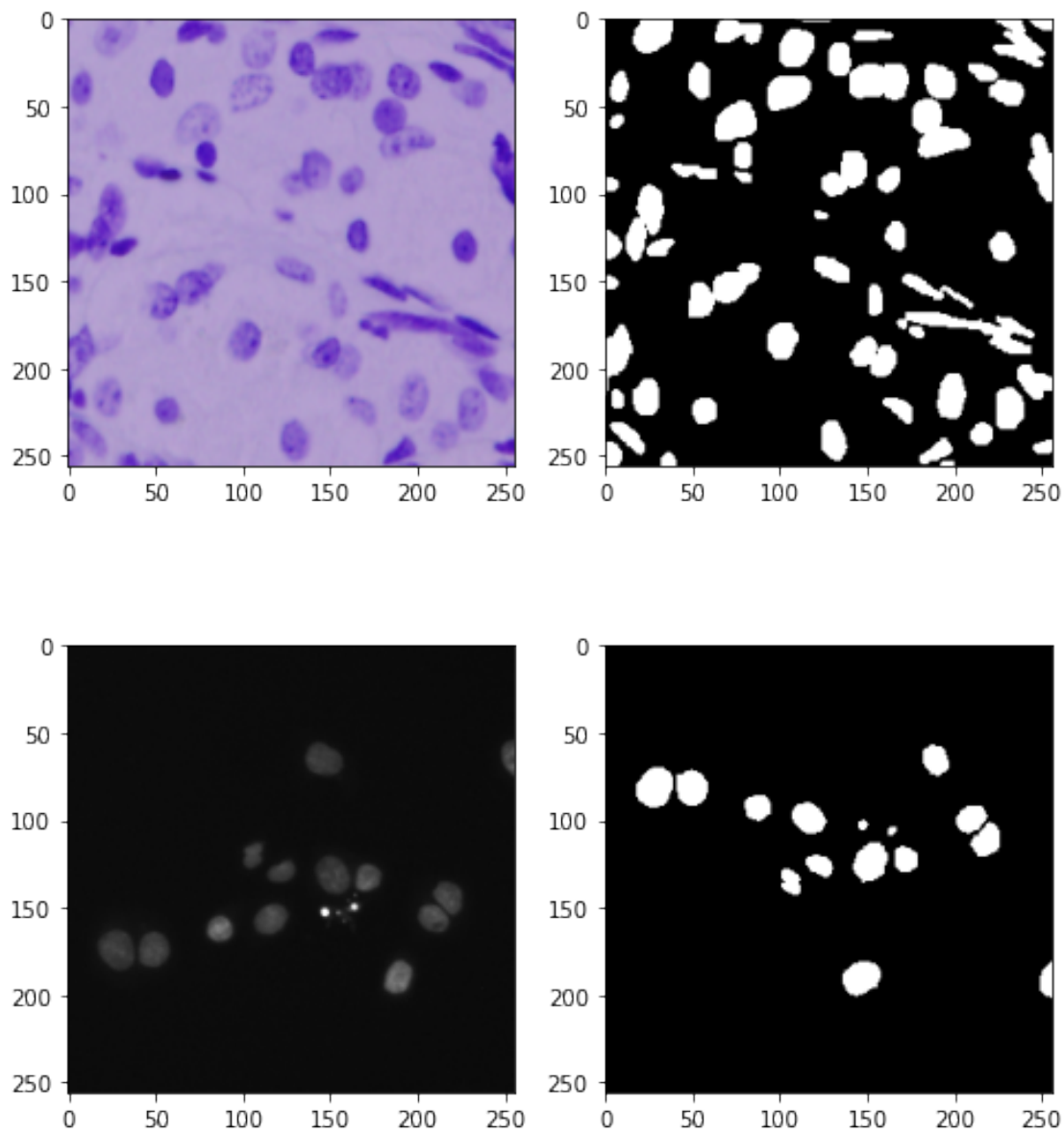
## 5 Sample of Train and Validation datasets

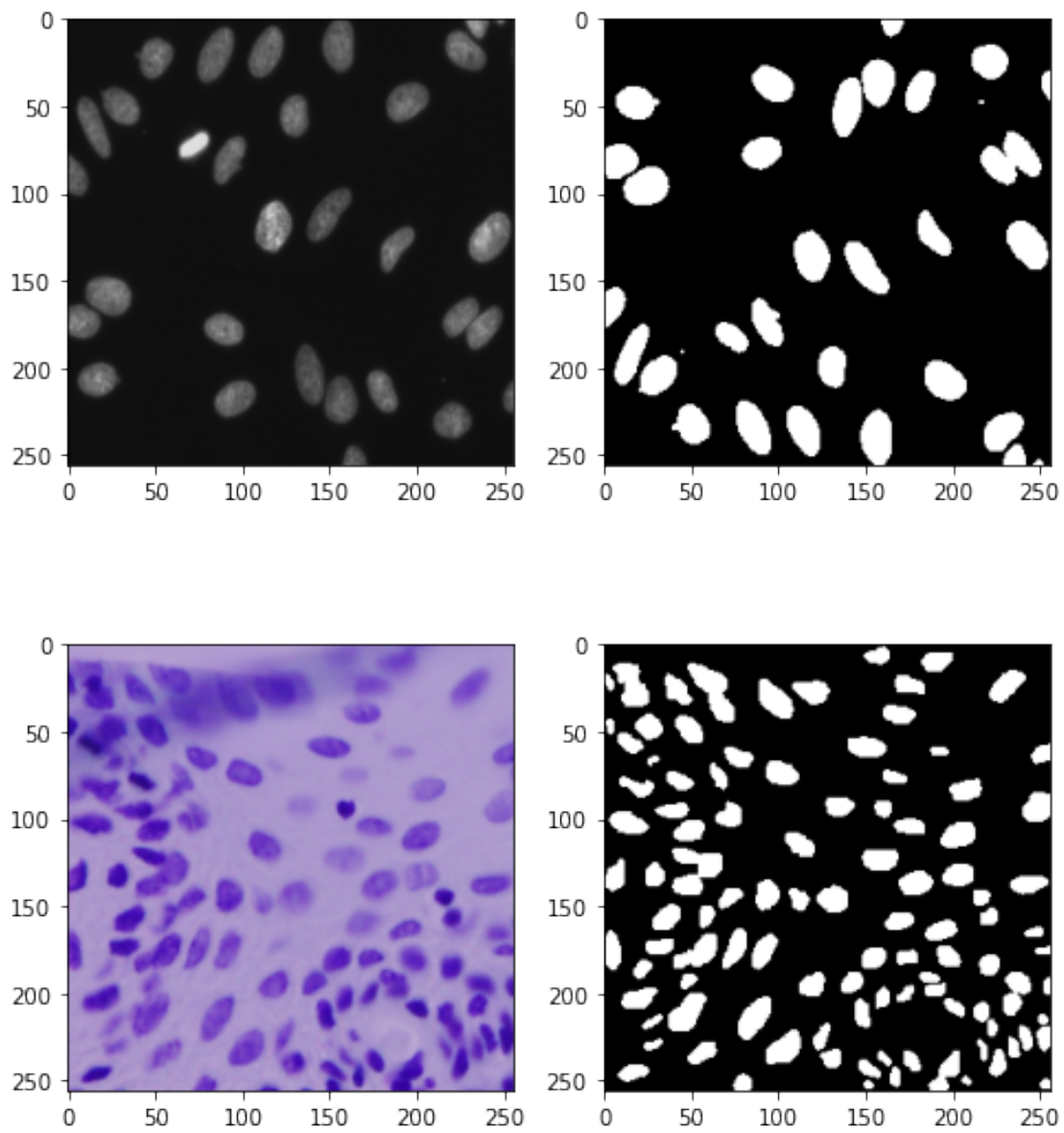
```

[18]: # Sample of training dataset
for image, mask in train_ds.take(1):
    for i in range(BATCH_SIZE):
        plt.figure(figsize=(8,4))
        plt.subplot(121)
        plt.imshow(image[i])
        plt.subplot(122)
        plt.imshow(mask[i], cmap = 'gray')
        plt.show()

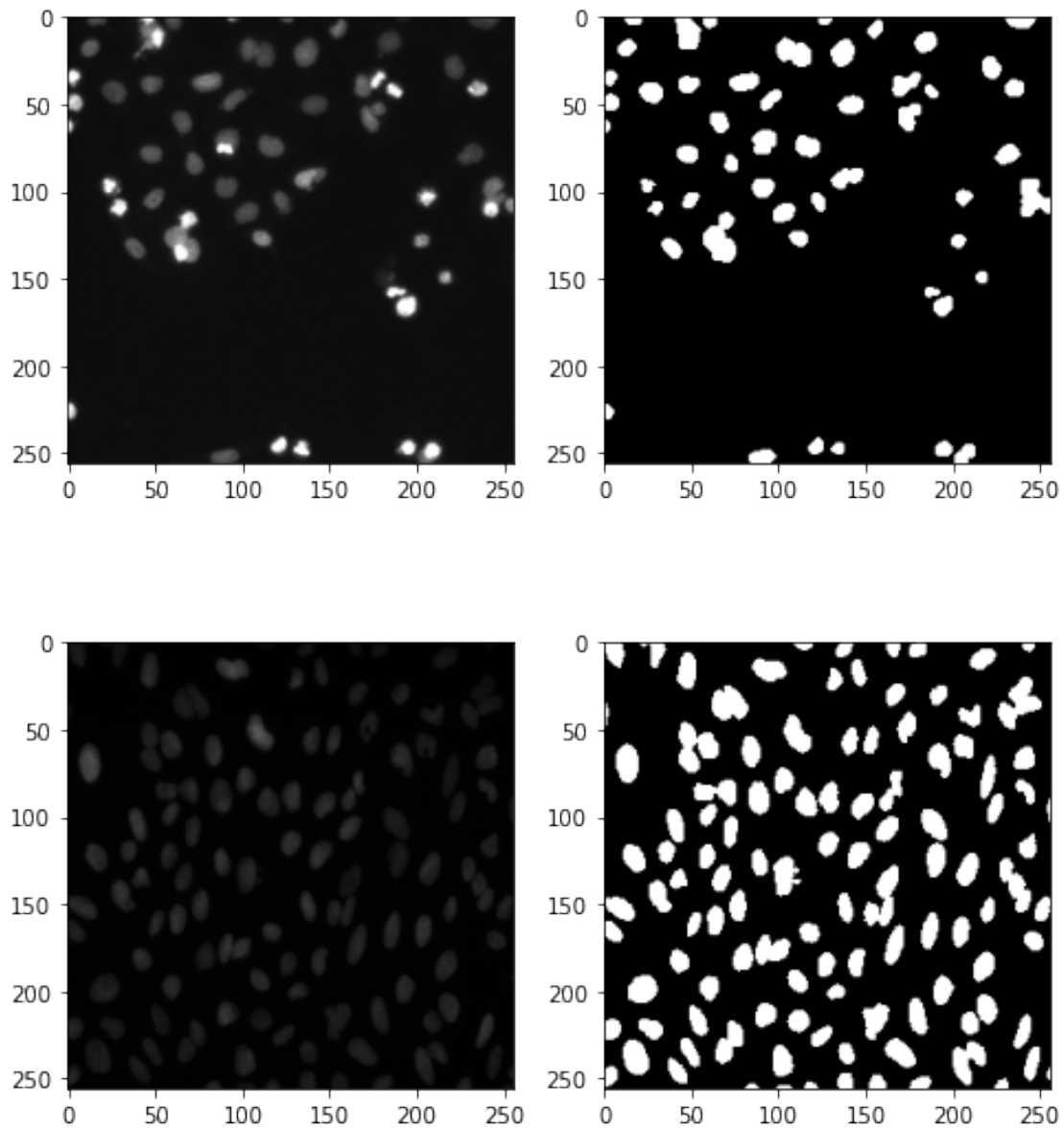
```











```
[19]: # Sample of validation dataset
for image, mask in val_ds.take(1):
    for i in range(BATCH_SIZE):
        plt.figure(figsize=(8,4))
        plt.subplot(121)
        plt.imshow(image[i])
        plt.subplot(122)
        plt.imshow(mask[i], cmap = 'gray')
        plt.show()
```

