

In [1]:

```
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-python
# For example, here's several helpful packages to load

import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)

# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all files
under the input directory

import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))

# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserve
d as output when you create a version using "Save & Run All"
# You can also write temporary files to /kaggle/temp/, but they won't be saved outside of
the current session
```

```
/kaggle/input/preprocesseddata/prepared_data/test/images/48.png
/kaggle/input/preprocesseddata/prepared_data/test/images/61.png
/kaggle/input/preprocesseddata/prepared_data/test/images/37.png
/kaggle/input/preprocesseddata/prepared_data/test/images/35.png
/kaggle/input/preprocesseddata/prepared_data/test/images/11.png
/kaggle/input/preprocesseddata/prepared_data/test/images/31.png
/kaggle/input/preprocesseddata/prepared_data/test/images/4.png
/kaggle/input/preprocesseddata/prepared_data/test/images/43.png
/kaggle/input/preprocesseddata/prepared_data/test/images/40.png
/kaggle/input/preprocesseddata/prepared_data/test/images/33.png
/kaggle/input/preprocesseddata/prepared_data/test/images/9.png
/kaggle/input/preprocesseddata/prepared_data/test/images/56.png
/kaggle/input/preprocesseddata/prepared_data/test/images/55.png
/kaggle/input/preprocesseddata/prepared_data/test/images/50.png
/kaggle/input/preprocesseddata/prepared_data/test/images/49.png
/kaggle/input/preprocesseddata/prepared_data/test/images/59.png
/kaggle/input/preprocesseddata/prepared_data/test/images/14.png
/kaggle/input/preprocesseddata/prepared_data/test/images/65.png
/kaggle/input/preprocesseddata/prepared_data/test/images/1.png
/kaggle/input/preprocesseddata/prepared_data/test/images/39.png
/kaggle/input/preprocesseddata/prepared_data/test/images/20.png
/kaggle/input/preprocesseddata/prepared_data/test/images/38.png
/kaggle/input/preprocesseddata/prepared_data/test/images/2.png
/kaggle/input/preprocesseddata/prepared_data/test/images/10.png
/kaggle/input/preprocesseddata/prepared_data/test/images/36.png
/kaggle/input/preprocesseddata/prepared_data/test/images/66.png
/kaggle/input/preprocesseddata/prepared_data/test/images/58.png
/kaggle/input/preprocesseddata/prepared_data/test/images/18.png
/kaggle/input/preprocesseddata/prepared_data/test/images/63.png
/kaggle/input/preprocesseddata/prepared_data/test/images/21.png
/kaggle/input/preprocesseddata/prepared_data/test/images/12.png
/kaggle/input/preprocesseddata/prepared_data/test/images/28.png
/kaggle/input/preprocesseddata/prepared_data/test/images/22.png
/kaggle/input/preprocesseddata/prepared_data/test/images/34.png
/kaggle/input/preprocesseddata/prepared_data/test/images/41.png
/kaggle/input/preprocesseddata/prepared_data/test/images/7.png
/kaggle/input/preprocesseddata/prepared_data/test/images/29.png
/kaggle/input/preprocesseddata/prepared_data/test/images/17.png
/kaggle/input/preprocesseddata/prepared_data/test/images/53.png
/kaggle/input/preprocesseddata/prepared_data/test/images/52.png
/kaggle/input/preprocesseddata/prepared_data/test/images/5.png
/kaggle/input/preprocesseddata/prepared_data/test/images/3.png
/kaggle/input/preprocesseddata/prepared_data/test/images/64.png
/kaggle/input/preprocesseddata/prepared_data/test/images/16.png
/kaggle/input/preprocesseddata/prepared_data/test/images/8.png
```

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

/kaggle/input/preprocesseddata/prepared_data/train/masks/230.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/44.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/189.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/68.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/180.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/46.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/57.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/84.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/15.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/210.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/234.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/252.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/183.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/30.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/60.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/235.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/99.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/155.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/228.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/27.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/237.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/201.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/78.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/212.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/47.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/207.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/149.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/13.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/260.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/253.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/206.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/134.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/259.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/25.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/54.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/111.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/62.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/205.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/0.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/141.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/19.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/110.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/26.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/79.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/257.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/42.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/175.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/226.png
/kaggle/input/preprocesseddata/prepared_data/train/masks/217.png

In [2]:

```
import cv2
import matplotlib.pyplot as plt
import os
from PIL import Image
import numpy as np
import pandas as pd
from glob import glob

import os
import random
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
plt.style.use("ggplot")
%matplotlib inline

from tqdm import tqdm_notebook, trange
from itertools import chain
from skimage.io import imread, imshow, concatenate_images
from skimage.transform import resize
```

```

from skimage.morphology import label
from sklearn.model_selection import train_test_split

import tensorflow as tf

from keras.models import Model, load_model
from keras.layers import Input, BatchNormalization, Activation, Dense, Dropout
from keras.layers.core import Lambda, RepeatVector, Reshape
from keras.layers.convolutional import Conv2D, Conv2DTranspose
from keras.layers.pooling import MaxPooling2D, GlobalMaxPool2D
from keras.layers.merge import concatenate, add
from keras.callbacks import EarlyStopping, ModelCheckpoint, ReduceLROnPlateau
from tensorflow.keras.optimizers import Adam
from keras.preprocessing.image import ImageDataGenerator, array_to_img, img_to_array, load_img

```

In [3]:

```

train_images = glob('../input/preprocesseddata/prepared_data/train/images/*.png')
train_masks = glob('../input/preprocesseddata/prepared_data/train/masks/*.png')

```

In [4]:

```

img = []
for filename in train_images:
    img.append(filename)
img.sort()

```

In [5]:

```

mask = []
for filename in train_masks:
    mask.append(filename)
mask.sort()

```

In [6]:

```

# Set some parameters
im_width = 512
im_height = 512
border = 5
ids = next(os.walk("../input/preprocesseddata/prepared_data/train/images"))[2] # list of
names all images in the given path
print("No. of images = ", len(ids))

```

No. of images = 268

In [7]:

```

X = np.zeros((len(ids), im_height, im_width, 1), dtype=np.float32)
y = np.zeros((len(ids), im_height, im_width, 1), dtype=np.float32)

```

In [8]:

```

# tqdm is used to display the progress bar
for n, id_ in tqdm_notebook(enumerate(ids), total=len(ids)):
    # Load images
    img = load_img("../input/preprocesseddata/prepared_data/train/images/"+id_, grayscale=True)
    x_img = img_to_array(img)
    x_img = resize(x_img, (512, 512, 1), mode = 'constant', preserve_range = True)
    # Load masks
    mask = img_to_array(load_img("../input/preprocesseddata/prepared_data/train/masks/"+id_, grayscale=True))
    mask = resize(mask, (512, 512, 1), mode = 'constant', preserve_range = True)
    # Save images
    X[n] = x_img/255.0
    y[n] = mask/255.0

```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:2: TqdmDeprecationWarning: This function will be removed in tqdm==5.0.0

Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`

/opt/conda/lib/python3.7/site-packages/keras_preprocessing/image/utils.py:107: UserWarning: grayscale is deprecated. Please use color_mode = "grayscale"
warnings.warn('grayscale is deprecated. Please use ')

In [9]:

```
# Split train and valid
X_train, X_valid, y_train, y_valid = train_test_split(X, y, test_size=0.1, random_state=42)
```

In [10]:

```
# Visualize any random image along with the mask
ix = random.randint(0, len(X_train))
has_mask = y_train[ix].max() > 0

fig, (ax1, ax2) = plt.subplots(1, 2, figsize = (20, 15))

ax1.imshow(X_train[ix, ..., 0], cmap = 'seismic', interpolation = 'bilinear')
if has_mask:
    # draw a boundary(contour) in the original image
    ax1.contour(y_train[ix].squeeze(), colors = 'k', linewidths = 5, levels = [0.5])
ax1.set_title('Lung CT Scan')

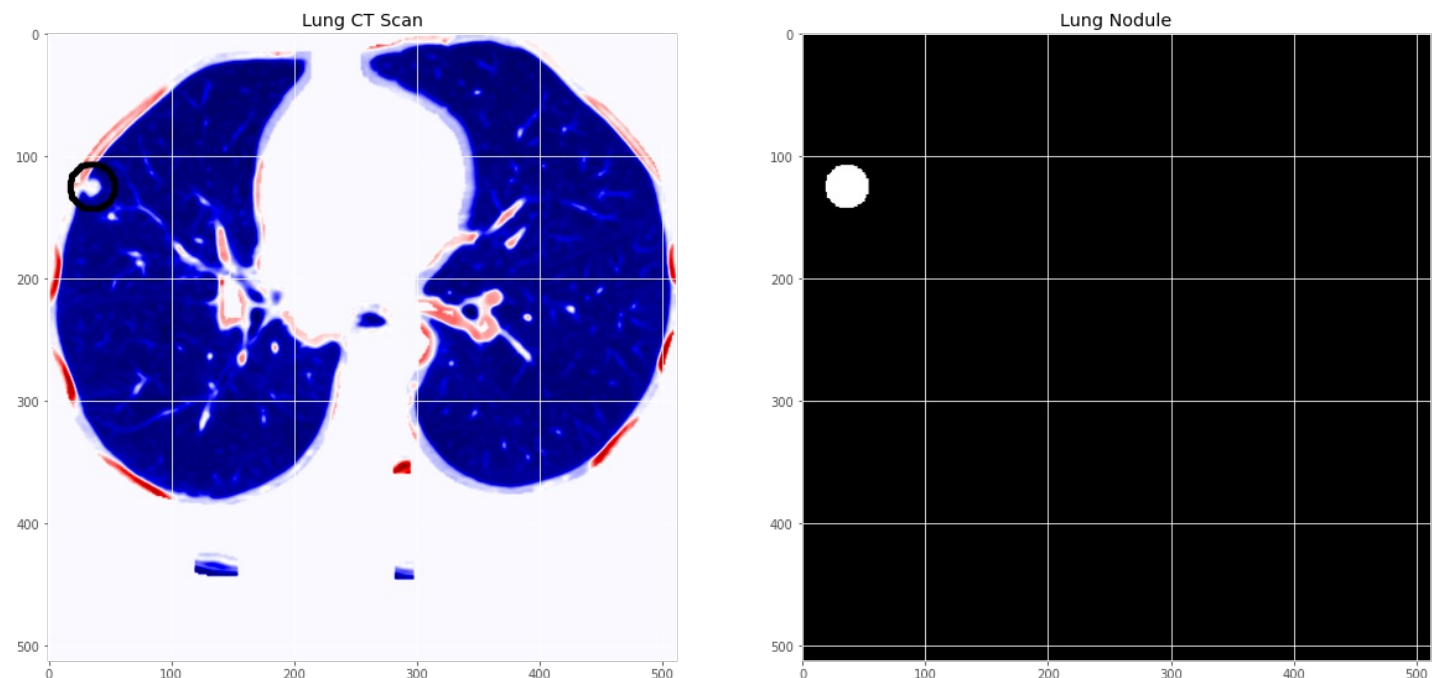
ax2.imshow(y_train[ix].squeeze(), cmap = 'gray', interpolation = 'bilinear')
ax2.set_title('Lung Nodule')
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:10: UserWarning: No contour levels were found within the data range.

Remove the CWD from sys.path while we load stuff.

Out[10]:

Text(0.5, 1.0, 'Lung Nodule')



In [11]:

```
def conv2d_block(input_tensor, n_filters, kernel_size = 3, batchnorm = True):
    """Function to add 2 convolutional layers with the parameters passed to it"""
    # first layer
    x = Conv2D(filters = n_filters, kernel_size = (kernel_size, kernel_size),\
               kernel_initializer = 'he_normal', padding = 'same')(input_tensor)
    if batchnorm:
        x = BatchNormalization()(x)
    x = Activation('relu')(x)
```

```

# second layer
x = Conv2D(filters = n_filters, kernel_size = (kernel_size, kernel_size),\
          kernel_initializer = 'he_normal', padding = 'same')(input_tensor)
if batchnorm:
    x = BatchNormalization()(x)
x = Activation('relu')(x)

return x

```

In [12]:

```

def get_unet(input_img, n_filters = 16, dropout = 0.1, batchnorm = True):
    """Function to define the UNET Model"""
    # Contracting Path
    c1 = conv2d_block(input_img, n_filters * 1, kernel_size = 3, batchnorm = batchnorm)
    p1 = MaxPooling2D((2, 2), padding='same')(c1)
    p1 = Dropout(dropout)(p1)

    c2 = conv2d_block(p1, n_filters * 2, kernel_size = 3, batchnorm = batchnorm)
    p2 = MaxPooling2D((2, 2), padding='same')(c2)
    p2 = Dropout(dropout)(p2)

    c3 = conv2d_block(p2, n_filters * 4, kernel_size = 3, batchnorm = batchnorm)
    p3 = MaxPooling2D((2, 2), padding='same')(c3)
    p3 = Dropout(dropout)(p3)

    c4 = conv2d_block(p3, n_filters * 8, kernel_size = 3, batchnorm = batchnorm)
    p4 = MaxPooling2D((2, 2), padding='same')(c4)
    p4 = Dropout(dropout)(p4)

    c5 = conv2d_block(p4, n_filters = n_filters * 16, kernel_size = 3, batchnorm = batchnorm)

    # Expansive Path
    u6 = Conv2DTranspose(n_filters * 8, (3, 3), strides = (2, 2), padding = 'same')(c5)
    u6 = concatenate([u6, c4])
    u6 = Dropout(dropout)(u6)
    c6 = conv2d_block(u6, n_filters * 8, kernel_size = 3, batchnorm = batchnorm)

    u7 = Conv2DTranspose(n_filters * 4, (3, 3), strides = (2, 2), padding = 'same')(c6)
    u7 = concatenate([u7, c3])
    u7 = Dropout(dropout)(u7)
    c7 = conv2d_block(u7, n_filters * 4, kernel_size = 3, batchnorm = batchnorm)

    u8 = Conv2DTranspose(n_filters * 2, (3, 3), strides = (2, 2), padding = 'same')(c7)
    u8 = concatenate([u8, c2])
    u8 = Dropout(dropout)(u8)
    c8 = conv2d_block(u8, n_filters * 2, kernel_size = 3, batchnorm = batchnorm)

    u9 = Conv2DTranspose(n_filters * 1, (3, 3), strides = (2, 2), padding = 'same')(c8)
    u9 = concatenate([u9, c1])
    u9 = Dropout(dropout)(u9)
    c9 = conv2d_block(u9, n_filters * 1, kernel_size = 3, batchnorm = batchnorm)

    outputs = Conv2D(1, (1, 1), activation='sigmoid')(c9)
    model = Model(inputs=[input_img], outputs=[outputs])
    return model

```

In [13]:

```

input_img = Input((im_height, im_width, 1), name='img')
model = get_unet(input_img, n_filters=16, dropout=0.05, batchnorm=True)
model.compile(optimizer=Adam(), loss="binary_crossentropy", metrics=["accuracy"])

```

2022-05-07 16:24:53.781264: I tensorflow/core/common_runtime/process_util.cc:146] Creating new thread pool with default inter op setting: 2. Tune using inter_op_parallelism_threads for best performance.

In [14]:

```

model.summary()

```

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
=====			
img (InputLayer)	[(None, 512, 512, 1)]	0	
conv2d_1 (Conv2D)	(None, 512, 512, 16)	160	img[0][0]
batch_normalization_1 (BatchNormalizati	(None, 512, 512, 16)	64	conv2d_1[0][0]
activation_1 (Activation)	(None, 512, 512, 16)	0	batch_normalization_1[0][0]
max_pooling2d (MaxPooling2D)	(None, 256, 256, 16)	0	activation_1[0][0]
dropout (Dropout)	(None, 256, 256, 16)	0	max_pooling2d[0][0]
conv2d_3 (Conv2D)	(None, 256, 256, 32)	4640	dropout[0][0]
batch_normalization_3 (BatchNormalizati	(None, 256, 256, 32)	128	conv2d_3[0][0]
activation_3 (Activation)	(None, 256, 256, 32)	0	batch_normalization_3[0][0]
max_pooling2d_1 (MaxPooling2D)	(None, 128, 128, 32)	0	activation_3[0][0]
dropout_1 (Dropout)	(None, 128, 128, 32)	0	max_pooling2d_1[0][0]
conv2d_5 (Conv2D)	(None, 128, 128, 64)	18496	dropout_1[0][0]
batch_normalization_5 (BatchNormalizati	(None, 128, 128, 64)	256	conv2d_5[0][0]
activation_5 (Activation)	(None, 128, 128, 64)	0	batch_normalization_5[0][0]
max_pooling2d_2 (MaxPooling2D)	(None, 64, 64, 64)	0	activation_5[0][0]
dropout_2 (Dropout)	(None, 64, 64, 64)	0	max_pooling2d_2[0][0]

conv2d_7 (Conv2D)	(None, 64, 64, 128)	73856	dropout_2[0][0]
batch_normalization_7 (BatchNormalizati	(None, 64, 64, 128)	512	conv2d_7[0][0]
activation_7 (Activation)	(None, 64, 64, 128)	0	batch_normalization_7[0][0]
max_pooling2d_3 (MaxPooling2D)	(None, 32, 32, 128)	0	activation_7[0][0]
dropout_3 (Dropout)	(None, 32, 32, 128)	0	max_pooling2d_3[0][0]
conv2d_9 (Conv2D)	(None, 32, 32, 256)	295168	dropout_3[0][0]
batch_normalization_9 (BatchNormalizati	(None, 32, 32, 256)	1024	conv2d_9[0][0]
activation_9 (Activation)	(None, 32, 32, 256)	0	batch_normalization_9[0][0]
conv2d_transpose (Conv2DTranspose)	(None, 64, 64, 128)	295040	activation_9[0][0]
concatenate (Concatenate)	(None, 64, 64, 256)	0	conv2d_transpose[0][0] activation_7[0][0]
dropout_4 (Dropout)	(None, 64, 64, 256)	0	concatenate[0][0]
conv2d_11 (Conv2D)	(None, 64, 64, 128)	295040	dropout_4[0][0]
batch_normalization_11 (BatchNormalizati	(None, 64, 64, 128)	512	conv2d_11[0][0]
activation_11 (Activation)	(None, 64, 64, 128)	0	batch_normalization_11[0][0]
conv2d_transpose_1 (Conv2DTranspose)	(None, 128, 128, 64)	73792	activation_11[0][0]
concatenate_1 (Concatenate)	(None, 128, 128, 128)	0	conv2d_transpose_1[0][0] activation_5[0][0]
dropout_5 (Dropout)	(None, 128, 128, 128)	0	concatenate_1[0][0]

conv2d_13 (Conv2D)	(None, 128, 128, 64) 73792	dropout_5[0][0]
batch_normalization_13 (Batch Normalization)	(None, 128, 128, 64) 256	conv2d_13[0][0]
activation_13 (Activation)	(None, 128, 128, 64) 0	batch_normalization_13[0][0]
conv2d_transpose_2 (Conv2DTranspose)	(None, 256, 256, 32) 18464	activation_13[0][0]
concatenate_2 (Concatenate)	(None, 256, 256, 64) 0	conv2d_transpose_2[0][0] activation_3[0][0]
dropout_6 (Dropout)	(None, 256, 256, 64) 0	concatenate_2[0][0]
conv2d_15 (Conv2D)	(None, 256, 256, 32) 18464	dropout_6[0][0]
batch_normalization_15 (Batch Normalization)	(None, 256, 256, 32) 128	conv2d_15[0][0]
activation_15 (Activation)	(None, 256, 256, 32) 0	batch_normalization_15[0][0]
conv2d_transpose_3 (Conv2DTranspose)	(None, 512, 512, 16) 4624	activation_15[0][0]
concatenate_3 (Concatenate)	(None, 512, 512, 32) 0	conv2d_transpose_3[0][0] activation_1[0][0]
dropout_7 (Dropout)	(None, 512, 512, 32) 0	concatenate_3[0][0]
conv2d_17 (Conv2D)	(None, 512, 512, 16) 4624	dropout_7[0][0]
batch_normalization_17 (Batch Normalization)	(None, 512, 512, 16) 64	conv2d_17[0][0]
activation_17 (Activation)	(None, 512, 512, 16) 0	batch_normalization_17[0][0]
conv2d_18 (Conv2D)	(None, 512, 512, 1) 17	activation_17[0][0]
=====		
=====		
Total params: 1,179,121		
Trainable params: 1,177,649		
Non-trainable params: 1,472		

In [15]:

```
callbacks = [  
    EarlyStopping(patience=10, verbose=1),  
    ReduceLROnPlateau(factor=0.1, patience=5, min_lr=0.00001, verbose=1),  
    ModelCheckpoint('./lung_nodules_model.h5', verbose=1, save_best_only=True, save_weights_only=True)  
]
```

In [16]:

```
results = model.fit(X_train, y_train, batch_size=32, epochs=10, callbacks=callbacks,\n                    validation_data=(X_valid, y_valid))
```

```
2022-05-07 16:24:55.175948: I tensorflow/compiler/mlir/mlir_graph_optimization_pass.cc:185] None of the MLIR Optimization Passes are enabled (registered 2)
```

Epoch 1/10

8/8 [=====] - 175s 21s/step - loss: 0.7473 - accuracy: 0.5841 - val_loss: 13.7697 - val_accuracy: 0.1505

Epoch 00001: val_loss improved from inf to 13.76968, saving model to ./lung_nodules_model.h5

Epoch 2/10

8/8 [=====] - 168s 21s/step - loss: 0.5487 - accuracy: 0.9413 - val_loss: 54.9287 - val_accuracy: 0.0020

Epoch 00002: val_loss did not improve from 13.76968

Epoch 3/10

8/8 [=====] - 167s 21s/step - loss: 0.4664 - accuracy: 0.9753 - val_loss: 16.9327 - val_accuracy: 0.0074

Epoch 00003: val_loss did not improve from 13.76968

Epoch 4/10

8/8 [=====] - 168s 21s/step - loss: 0.4234 - accuracy: 0.9859 - val_loss: 8.1438 - val_accuracy: 0.0309

Epoch 00004: val_loss improved from 13.76968 to 8.14383, saving model to ./lung_nodules_model.h5

Epoch 5/10

8/8 [=====] - 166s 21s/step - loss: 0.3941 - accuracy: 0.9899 - val_loss: 0.7280 - val_accuracy: 0.7962

Epoch 00005: val_loss improved from 8.14383 to 0.72802, saving model to ./lung_nodules_model.h5

Epoch 6/10

8/8 [=====] - 168s 21s/step - loss: 0.3663 - accuracy: 0.9930 - val_loss: 0.5230 - val_accuracy: 0.9361

Epoch 00006: val_loss improved from 0.72802 to 0.52298, saving model to ./lung_nodules_model.h5

Epoch 7/10

8/8 [=====] - 167s 21s/step - loss: 0.3449 - accuracy: 0.9936 - val_loss: 0.5837 - val_accuracy: 0.8423

Epoch 00007: val_loss did not improve from 0.52298

Epoch 8/10

8/8 [=====] - 168s 21s/step - loss: 0.3247 - accuracy: 0.9944 - val_loss: 0.4533 - val_accuracy: 0.9945

Epoch 00008: val_loss improved from 0.52298 to 0.45335, saving model to ./lung_nodules_model.h5

Epoch 9/10

8/8 [=====] - 167s 21s/step - loss: 0.3061 - accuracy: 0.9950 - val_loss: 0.4272 - val_accuracy: 0.9958

Epoch 00009: val_loss improved from 0.45335 to 0.42715, saving model to ./lung_nodules_model.h5

Epoch 10/10

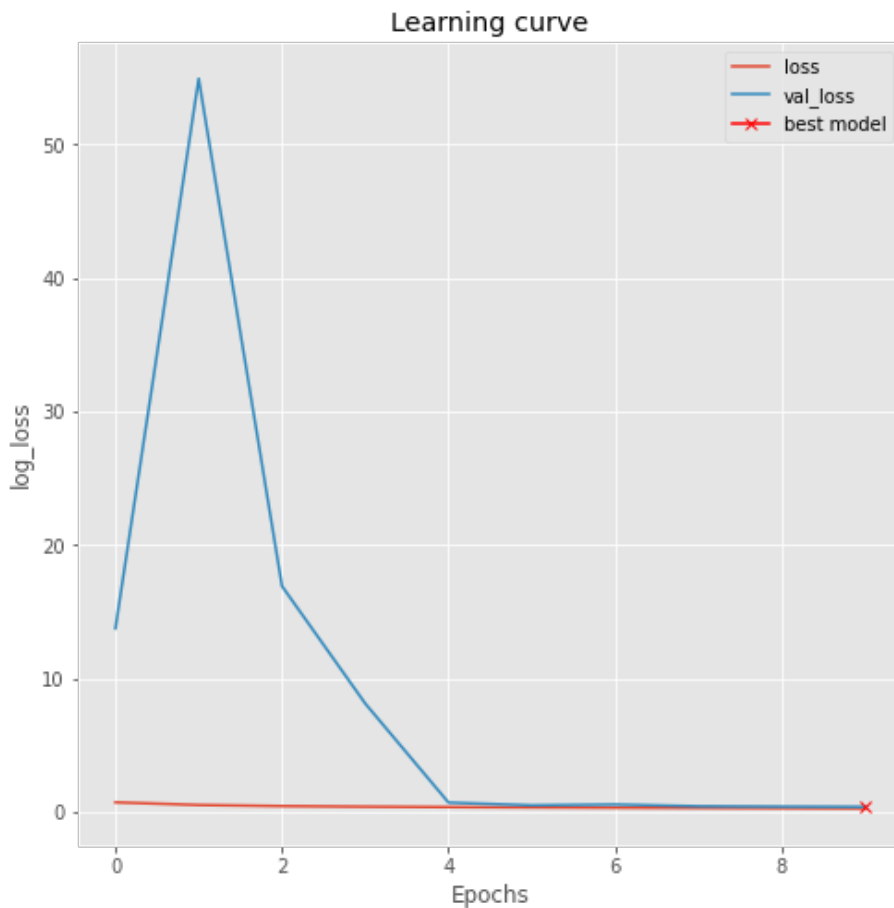
8/8 [=====] - 168s 21s/step - loss: 0.2901 - accuracy: 0.9953 -

val_loss: 0.4100 - val_accuracy: 0.9958

Epoch 00010: val_loss improved from 0.42715 to 0.40997, saving model to ./lung_nodules_model.h5

In [17]:

```
plt.figure(figsize=(8, 8))
plt.title("Learning curve")
plt.plot(results.history["loss"], label="loss")
plt.plot(results.history["val_loss"], label="val_loss")
plt.plot(np.argmax(results.history["val_loss"]), np.min(results.history["val_loss"]), marker="x", color="r", label="best model")
plt.xlabel("Epochs")
plt.ylabel("log_loss")
plt.legend();
```



In [18]:

```
# load the best model
model.load_weights('./lung_nodules_model.h5')
```

In [19]:

```
model.save('./UNET_model.h5')
```

In [20]:

```
from keras.models import model_from_json
model_json = model.to_json()

with open("./UNET_model_weights.json", "w") as json_file:
    json_file.write(model_json)
model.save_weights("./UNET_model_weights.h5")
print("Saved model to disk")
```

Saved model to disk

In [21]:

```
# Evaluate on validation set (this must be equals to the best log_loss)
model.evaluate(X_valid, y_valid, verbose=1)
```

```
1/1 [=====] - 4s 4s/step - loss: 0.4100 - accuracy: 0.9958
```

Out[21]:

```
[0.409972220659256, 0.9957706332206726]
```

In [22]:

```
# Predict on train, val and test
preds_train = model.predict(X_train, verbose=1)
preds_val = model.predict(X_valid, verbose=1)
```

```
8/8 [=====] - 33s 4s/step
```

```
1/1 [=====] - 4s 4s/step
```

In [23]:

```
# Threshold predictions
preds_train_t = (preds_train > 0.5).astype(np.uint8)
preds_val_t = (preds_val > 0.5).astype(np.uint8)
```

In [24]:

```
def plot_sample(X, y, preds, binary_preds, ix=None):
    """Function to plot the results"""
    if ix is None:
        ix = random.randint(0, len(X))

    has_mask = y[ix].max() > 0

    fig, ax = plt.subplots(1, 4, figsize=(20, 10))
    ax[0].imshow(X[ix, ..., 0], cmap='seismic')
    if has_mask:
        ax[0].contour(y[ix].squeeze(), colors='k', levels=[0.5])
    ax[0].set_title('Lung CT Scan ')

    ax[1].imshow(y[ix].squeeze())
    ax[1].set_title('Ground Truth')

    ax[2].imshow(preds[ix].squeeze(), vmin=0, vmax=1)
    if has_mask:
        ax[2].contour(y[ix].squeeze(), colors='k', levels=[0.5])
    ax[2].set_title('Predicted Nodule')

    ax[3].imshow(binary_preds[ix].squeeze(), vmin=0, vmax=1)
    if has_mask:
        ax[3].contour(y[ix].squeeze(), colors='k', levels=[0.5])
    ax[3].set_title('Predicted Nodule Binary')
```

In [25]:

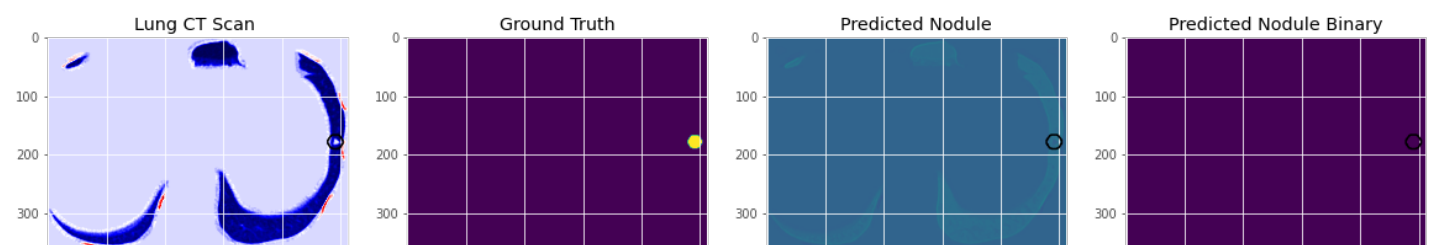
```
# Check if training data looks all right
plot_sample(X_train, y_train, preds_train, preds_train_t, ix=14)
```

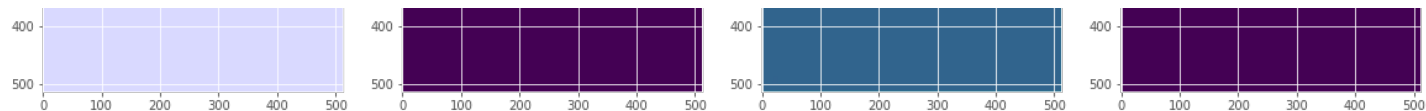
```
/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:11: UserWarning: No contour
levels were found within the data range.
```

```
# This is added back by InteractiveShellApp.init_path()
```

```
/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:19: UserWarning: No contour
levels were found within the data range.
```

```
/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:24: UserWarning: No contour
levels were found within the data range.
```





In [26]:

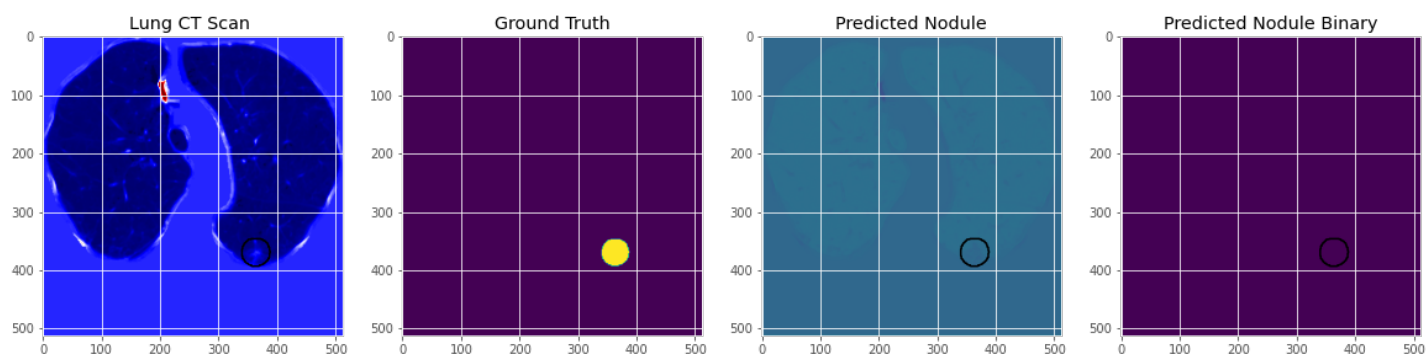
```
plot_sample(X_train, y_train, preds_train, preds_train_t)
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:11: UserWarning: No contour levels were found within the data range.

This is added back by InteractiveShellApp.init_path()

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:19: UserWarning: No contour levels were found within the data range.

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:24: UserWarning: No contour levels were found within the data range.



In [27]:

```
# Check if valid data looks all right
```

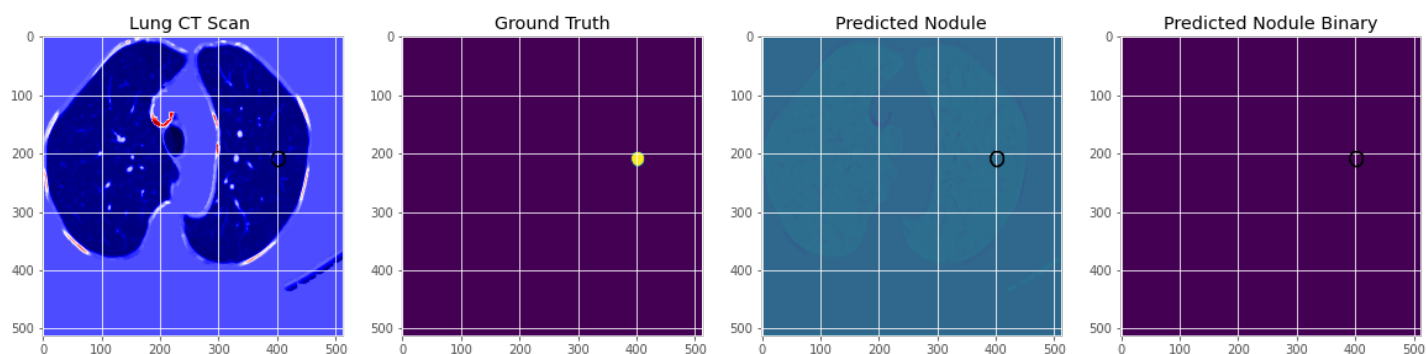
```
plot_sample(X_valid, y_valid, preds_val, preds_val_t, ix=19)
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:11: UserWarning: No contour levels were found within the data range.

This is added back by InteractiveShellApp.init_path()

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:19: UserWarning: No contour levels were found within the data range.

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:24: UserWarning: No contour levels were found within the data range.



In [28]:

```
# Check if valid data looks all right
```

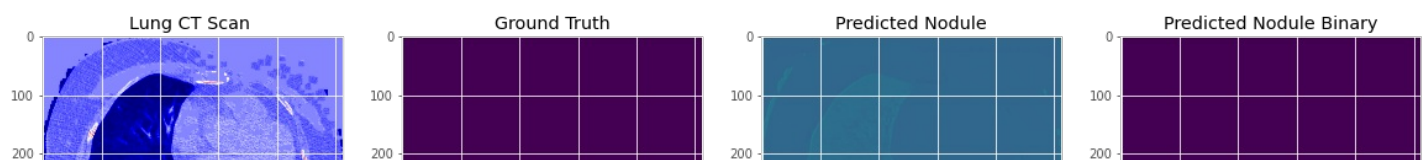
```
plot_sample(X_valid, y_valid, preds_val, preds_val_t)
```

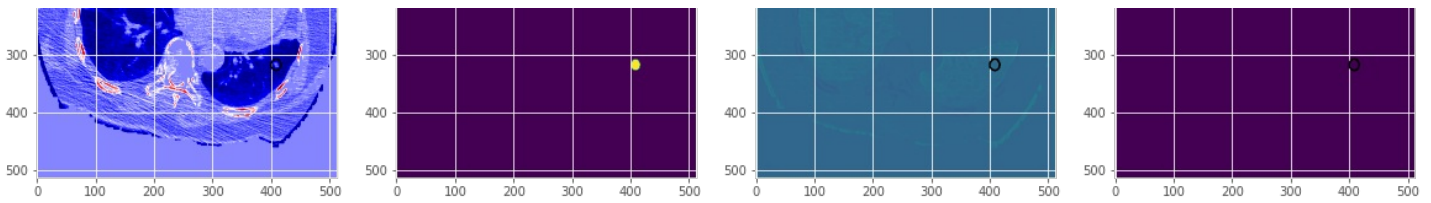
/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:11: UserWarning: No contour levels were found within the data range.

This is added back by InteractiveShellApp.init_path()

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:19: UserWarning: No contour levels were found within the data range.

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:24: UserWarning: No contour levels were found within the data range.





In [29]:

```
# LOAD MODEL
from keras.models import load_model
model = load_model('./UNET_model.h5')
```

In [30]:

```
test_images = glob('../input/preprocesseddata/prepared_data/test/images/*.png')
test_masks = glob('../input/preprocesseddata/prepared_data/test/masks/*.png')
```

In [31]:

```
img = []
for filename in test_images:
    img.append(filename)
img.sort()
mask = []
for filename in test_masks:
    mask.append(filename)
mask.sort()
```

In [32]:

```
# Set some parameters
im_width = 512
im_height = 512
border = 5
ids = next(os.walk("../input/preprocesseddata/prepared_data/test/images"))[2] # list of names all images in the given path
print("No. of images = ", len(ids))
```

No. of images = 67

In [33]:

```
X = np.zeros((1, im_height, im_width, 1), dtype=np.float32)
y = np.zeros((1, im_height, im_width, 1), dtype=np.float32)
```

In [34]:

```
# tqdm is used to display the progress bar
from tqdm import tqdm
images_path = "../input/preprocesseddata/prepared_data/test/images/"
masks_path = "../input/preprocesseddata/prepared_data/test/masks/"
img = load_img(images_path+'1.png', grayscale=True)
x_img = img_to_array(img)
x_img = resize(x_img, (512, 512, 1), mode = 'constant', preserve_range = True)
# Load masks
mask = img_to_array(load_img(masks_path+'1.png', grayscale=True))
mask = resize(mask, (512, 512, 1), mode = 'constant', preserve_range = True)
# Save images
X[0] = x_img/255.0
y[0] = mask/255.0
```

In [35]:

```
preds_val = model.predict(X, verbose=1)
```

1/1 [=====] - 1s 507ms/step

In [36]:

```
model.evaluate(X, y, verbose=1)
```

```
preds_val_t = (preds_val > 0.5).astype(np.uint8)
```

In [37]:

```
# Check if valid data looks all right  
plot_sample(X, y, preds_val, preds_val_t)
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:11: UserWarning: No contour levels were found within the data range.

```
# This is added back by InteractiveShellApp.init_path()
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:19: UserWarning: No contour levels were found within the data range.

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:24: UserWarning: No contour levels were found within the data range.

