

Group Name: Innovators Group 9

ML+CV Combined Project: Cell Segmentation

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Task Performed:

We implemented five layered UNet architecture as a part of hyper parameter training and implemented all the activation functions on the architecture

```
c1 = Conv2D(16, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(c1)
p1 = MaxPooling2D((2, 2))(c1)

#
c2 = Conv2D(32, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(p1)
c2 = Dropout(0.1)(c2)
c2 = Conv2D(32, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(c2)
p2 = MaxPooling2D((2, 2))(c2)

#
c3 = Conv2D(64, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(p2)
c3 = Dropout(0.2)(c3)
c3 = Conv2D(64, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(c3)
p3 = MaxPooling2D((2, 2))(c3)

#
c4 = Conv2D(128, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(p3)
c4 = Dropout(0.2)(c4)
c4 = Conv2D(128, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(c4)
p4 = MaxPooling2D(pool_size=(2, 2))(c4)

#
c_x = Conv2D(128, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(p3)
c4 = Dropout(0.2)(c4)
c_x = Conv2D(128, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(c4)
p_x = MaxPooling2D(pool_size=(2, 2))(c4)

#
c5 = Conv2D(256, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(p_x)
c5 = Dropout(0.3)(c5)
c5 = Conv2D(256, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(c5)

#Expansive path
u_x = Conv2DTranspose(32, (2, 2), strides=(2, 2), padding='same')(c5)
u_x = concatenate([u_x, c_x])
c_x = Conv2D(32, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(u_x)
c8 = Dropout(0.1)(c8)
c_x = Conv2D(32, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(c_x)

#
u6 = Conv2DTranspose(128, (2, 2), strides=(2, 2), padding='same')(c5)
u6 = concatenate([u6, c4])
c6 = Conv2D(128, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(u6)
c6 = Dropout(0.2)(c6)
c6 = Conv2D(128, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(c6)

#
u7 = Conv2DTranspose(64, (2, 2), strides=(2, 2), padding='same')(c6)
u7 = concatenate([u7, c3])
c7 = Conv2D(64, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(u7)
c7 = Dropout(0.2)(c7)
c7 = Conv2D(64, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(c7)

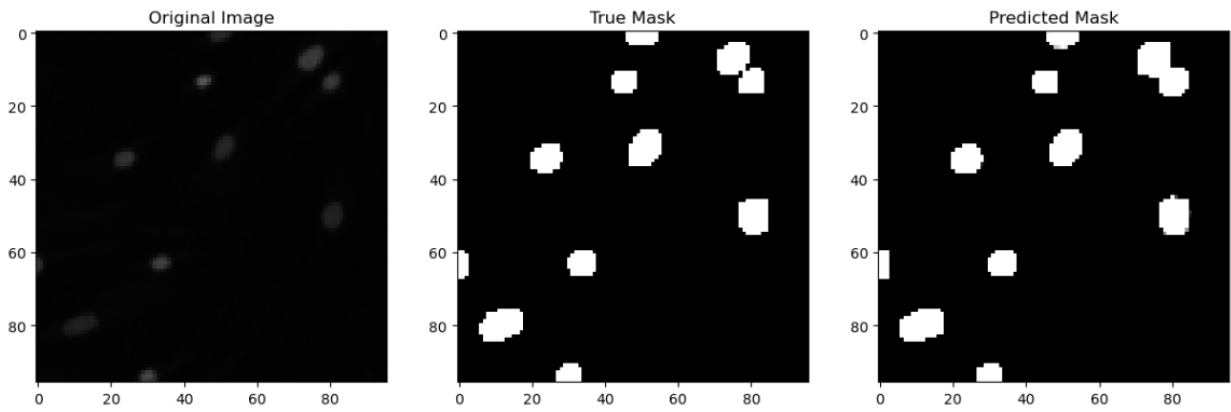
#
u8 = Conv2DTranspose(32, (2, 2), strides=(2, 2), padding='same')(c7)
u8 = concatenate([u8, c2])
c8 = Conv2D(32, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(u8)
c8 = Dropout(0.1)(c8)
c8 = Conv2D(32, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(c8)

#
u9 = Conv2DTranspose(16, (2, 2), strides=(2, 2), padding='same')(c8)
u9 = concatenate([u9, c1])
c9 = Conv2D(16, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(u9)
c9 = Dropout(0.1)(c9)
c9 = Conv2D(16, (3, 3), activation= self.activation, kernel_initializer=kernel_initializer, padding='same')(c9)
```

Outcomes:

We plotted predictions for every activation functions

Eg:Swish:



Eg: TrainableLeakyRelu

