

```
In [1]: import tensorflow as tf

In [2]: import os
import math
import numpy as np

In [3]: from tensorflow import keras
from tensorflow.keras import layers

In [4]: from tensorflow.keras.utils import load_img

In [5]: from tensorflow.keras.utils import array_to_img

In [6]: from tensorflow.keras.utils import img_to_array

In [7]: from tensorflow.keras.utils import image_dataset_from_directory

In [8]: from IPython.display import display

In [9]: root_dir = os.path.join("./data")

In [10]: crop_size = 300
upscale_factor = 3
input_size = crop_size // upscale_factor
batch_size = 4

train_ds = image_dataset_from_directory(
    root_dir,
    batch_size=batch_size,
    image_size=(crop_size, crop_size),
    validation_split=0.2,
    subset="training",
    seed=1337,
    label_mode=None,
)

valid_ds = image_dataset_from_directory(
    root_dir,
    batch_size=batch_size,
    image_size=(crop_size, crop_size),
    validation_split=0.2,
    subset="validation",
    seed=1337,
    label_mode=None,
)
```

Found 17186 files belonging to 1 classes.
Using 13749 files for training.
Found 17186 files belonging to 1 classes.
Using 3437 files for validation.

```
In [11]:  
def scaling(input_image):  
    input_image = input_image / 255.0  
    return input_image  
  
# Scale from (0, 255) to (0, 1)  
train_ds = train_ds.map(scaling)  
valid_ds = valid_ds.map(scaling)
```

```
In [12]:  
# import pillow  
# import PIL.Image  
import sys  
from PIL import Image  
sys.modules['Image'] = Image  
from PIL import Image  
print(Image.__file__)  
import Image  
print(Image.__file__)
```

C:\Users\anura\anaconda3\lib\site-packages\PIL\Image.py
C:\Users\anura\anaconda3\lib\site-packages\PIL\Image.py

```
In [13]:  
for batch in train_ds.take(1):  
    for img in batch:  
        display(array_to_img(img))
```





```
In [14]:  
dataset = os.path.join(root_dir, "images")  
test_path = os.path.join(dataset, "test")  
  
test_img_paths = sorted(  
    [  
        os.path.join(test_path, fname)  
    ]
```

```

        for fname in os.listdir(test_path)
            if fname.endswith(".jpg")
        )
    ]

```

In [15]:

```

# Use TF Ops to process.
def process_input(input, input_size, upscale_factor):
    input = tf.image.rgb_to_yuv(input)
    last_dimension_axis = len(input.shape) - 1
    y, u, v = tf.split(input, 3, axis=last_dimension_axis)
    return tf.image.resize(y, [input_size, input_size], method="area")

def process_target(input):
    input = tf.image.rgb_to_yuv(input)
    last_dimension_axis = len(input.shape) - 1
    y, u, v = tf.split(input, 3, axis=last_dimension_axis)
    return y

train_ds = train_ds.map(
    lambda x: (process_input(x, input_size, upscale_factor), process_target(x))
)

valid_ds = valid_ds.map(
    lambda x: (process_input(x, input_size, upscale_factor), process_target(x))
)

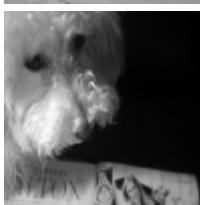
```

In [16]:

```

for batch in train_ds.take(1):
    for img in batch[0]:
        display(array_to_img(img))
    for img in batch[1]:
        display(array_to_img(img))

```







In [17]:

```
def get_model(upscale_factor=3, channels=1):
    conv_args = {
        "activation": "relu",
        "kernel_initializer": "Orthogonal",
        "padding": "same",
    }
    inputs = keras.Input(shape=(None, None, channels))
    x = layers.Conv2D(64, 5, **conv_args)(inputs)
    x = layers.Conv2D(64, 3, **conv_args)(x)
    x = layers.Conv2D(32, 3, **conv_args)(x)
    x = layers.Conv2D(channels * (upscale_factor ** 2), 3, **conv_args)(x)
    outputs = tf.nn.depth_to_space(x, upscale_factor)

    return keras.Model(inputs, outputs)
```

In [18]:

```
import matplotlib.pyplot as plt
from mpl_toolkits.axes_grid1.inset_locator import zoomed_inset_axes
from mpl_toolkits.axes_grid1.inset_locator import mark_inset
import PIL

def plot_results(img, prefix, title):
    """Plot the result with zoom-in area."""
    img_array = img_to_array(img)
    img_array = img_array.astype("float32") / 255.0

    # Create a new figure with a default 111 subplot.
    fig, ax = plt.subplots()
    im = ax.imshow(img_array[::-1], origin="lower")

    plt.title(title)
    # zoom-factor: 2.0, location: upper-left
    axins = zoomed_inset_axes(ax, 2, loc=2)
    axins.imshow(img_array[::-1], origin="lower")

    # Specify the limits.
    x1, x2, y1, y2 = 200, 300, 100, 200
    # Apply the x-limits.
    axins.set_xlim(x1, x2)
    # Apply the y-limits.
    axins.set_ylim(y1, y2)

    plt.yticks(visible=False)
```

```

plt.xticks(visible=False)

# Make the Line.
mark_inset(ax, axins, loc1=1, loc2=3, fc="none", ec="blue")
plt.savefig(str(prefix) + "-" + title + ".png")
plt.show()

def get_lowres_image(img, upscale_factor):
    """Return low-resolution image to use as model input."""
    return img.resize(
        (img.size[0] // upscale_factor, img.size[1] // upscale_factor),
        PIL.Image.BICUBIC,
    )

def upscale_image(model, img):
    """Predict the result based on input image and restore the image as RGB."""
    ycbcr = img.convert("YCbCr")
    y, cb, cr = ycbcr.split()
    y = img_to_array(y)
    y = y.astype("float32") / 255.0

    input = np.expand_dims(y, axis=0)
    out = model.predict(input)

    out_img_y = out[0]
    out_img_y *= 255.0

    # Restore the image in RGB color space.
    out_img_y = out_img_y.clip(0, 255)
    out_img_y = out_img_y.reshape((np.shape(out_img_y)[0], np.shape(out_img_y)[1]))
    out_img_y = PIL.Image.fromarray(np.uint8(out_img_y), mode="L")
    out_img_cb = cb.resize(out_img_y.size, PIL.Image.BICUBIC)
    out_img_cr = cr.resize(out_img_y.size, PIL.Image.BICUBIC)
    out_img = PIL.Image.merge("YCbCr", (out_img_y, out_img_cb, out_img_cr)).convert(
        "RGB"
    )
    return out_img

```

In [19]:

```

class ESPCNCallback(keras.callbacks.Callback):
    def __init__(self):
        super(ESPCNCallback, self).__init__()
        self.test_img = get_lowres_image(load_img(test_img_paths[0]), upscale_factor)

    # Store PSNR value in each epoch.
    def on_epoch_begin(self, epoch, logs=None):
        self.psnr = []

    def on_epoch_end(self, epoch, logs=None):
        print("Mean PSNR for epoch: %.2f" % (np.mean(self.psnr)))
        prediction = upscale_image(self.model, self.test_img)
        plot_results(prediction, "epoch-" + str(epoch), "prediction")

    def on_test_batch_end(self, batch, logs=None):
        self.psnr.append(10 * math.log10(1 / logs["loss"]))

```

In [20]:

```

early_stopping_callback = keras.callbacks.EarlyStopping(monitor="loss", patience=10)

checkpoint_filepath = "./tmp/checkpoint"

model_checkpoint_callback = keras.callbacks.ModelCheckpoint(
    filepath=checkpoint_filepath,
    save_weights_only=True,

```

```

        monitor="loss",
        mode="min",
        save_best_only=True,
    )

model = get_model(upscale_factor=upscale_factor, channels=1)
model.summary()

callbacks = [ESPCNCallback(), early_stopping_callback, model_checkpoint_callback]
loss_fn = keras.losses.MeanSquaredError()
optimizer = keras.optimizers.Adam(learning_rate=0.001)

```

Model: "model"

Layer (type)	Output Shape	Param #
<hr/>		
input_1 (InputLayer)	[None, None, None, 1]	0
conv2d (Conv2D)	(None, None, None, 64)	1664
conv2d_1 (Conv2D)	(None, None, None, 64)	36928
conv2d_2 (Conv2D)	(None, None, None, 32)	18464
conv2d_3 (Conv2D)	(None, None, None, 9)	2601
tf.nn.depth_to_space (TFOpL ambda)	(None, None, None, 1)	0
<hr/>		
Total params: 59,657		
Trainable params: 59,657		
Non-trainable params: 0		

In [21]:

```

# epochs = 10
epochs=50

model.compile(
    optimizer=optimizer, loss=loss_fn,
)

model.fit(
    train_ds, epochs=epochs, callbacks=callbacks, validation_data=valid_ds, verbose=
)

# The model weights (that are considered the best) are loaded into the model.
model.load_weights(checkpoint_filepath)

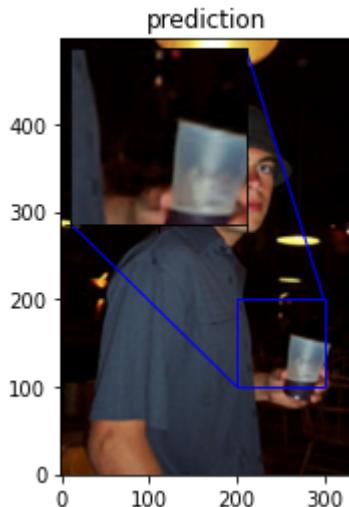
model.save('./model_final')

```

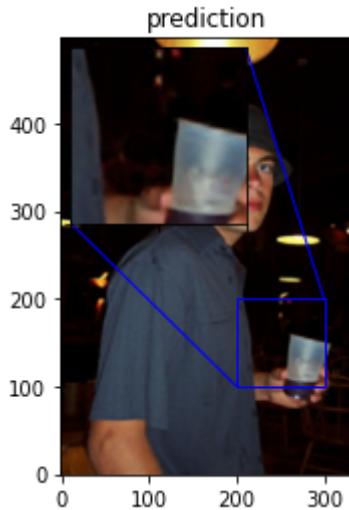
```

Epoch 1/50
3438/3438 [=====] - ETA: 0s - loss: 0.0037Mean PSNR for epoch: 25.35

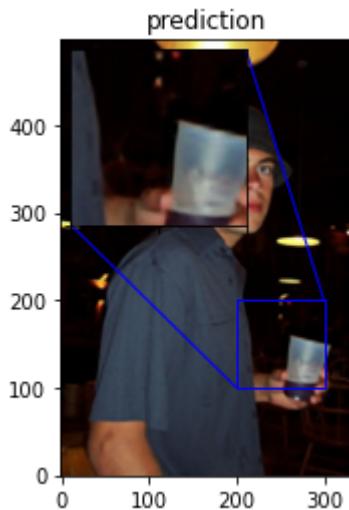
```



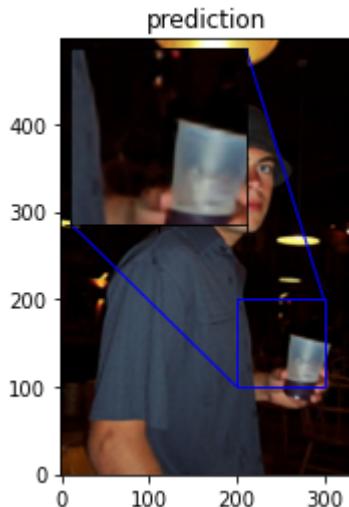
```
3438/3438 [=====] - 695s 202ms/step - loss: 0.0037 - val_loss: 0.0030  
Epoch 2/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0029Mean PSNR for epoch: 25.48
```



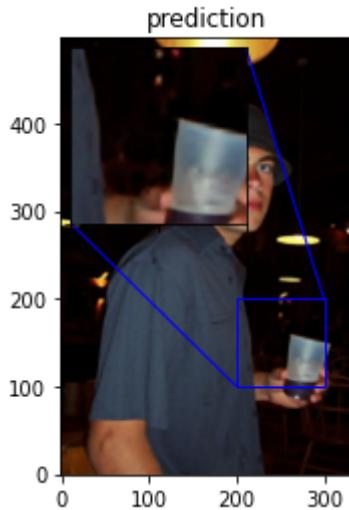
```
3438/3438 [=====] - 703s 205ms/step - loss: 0.0029 - val_loss: 0.0029  
Epoch 3/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0029Mean PSNR for epoch: 25.54
```



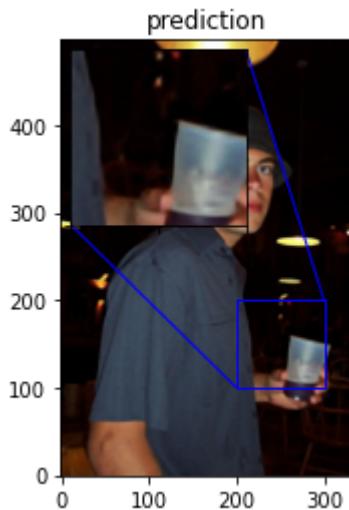
```
3438/3438 [=====] - 705s 205ms/step - loss: 0.0029 - val_loss: 0.0029  
Epoch 4/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0028Mean PSNR for epoch: 25.55
```



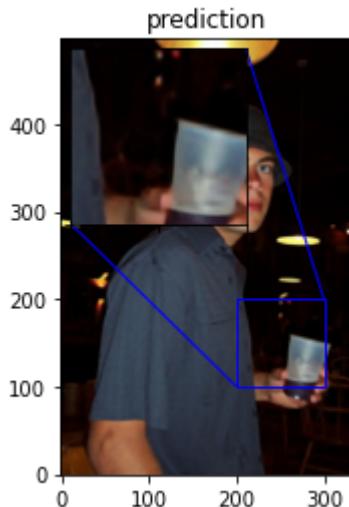
```
3438/3438 [=====] - 706s 205ms/step - loss: 0.0028 - val_loss: 0.0029  
Epoch 5/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0028Mean PSNR for epoch: 25.58
```



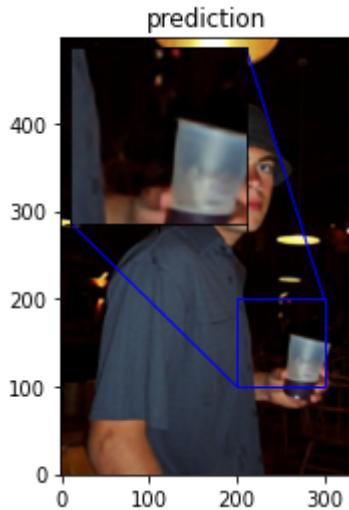
```
3438/3438 [=====] - 704s 205ms/step - loss: 0.0028 - val_loss: 0.0028  
Epoch 6/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0028Mean PSNR for epoch: 25.61
```



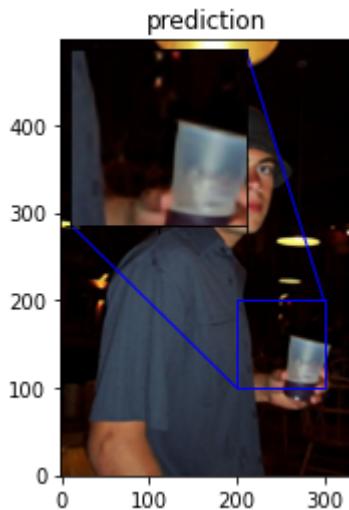
```
3438/3438 [=====] - 704s 205ms/step - loss: 0.0028 - val_loss: 0.0028  
Epoch 7/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0028Mean PSNR for epoch: 25.64
```



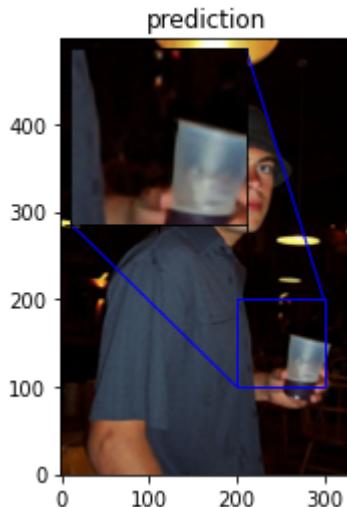
```
3438/3438 [=====] - 704s 205ms/step - loss: 0.0028 - val_loss: 0.0028  
Epoch 8/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0028Mean PSNR for epoch: 25.65
```



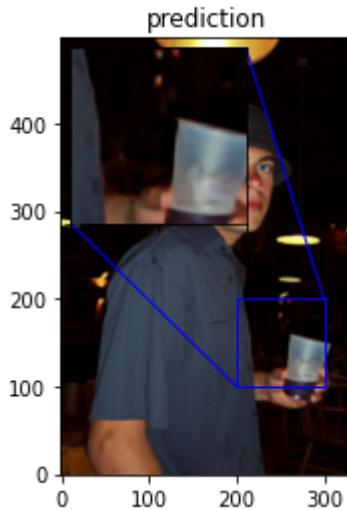
```
3438/3438 [=====] - 704s 205ms/step - loss: 0.0028 - val_loss: 0.0028  
Epoch 9/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0028Mean PSNR for epoch: 25.62
```



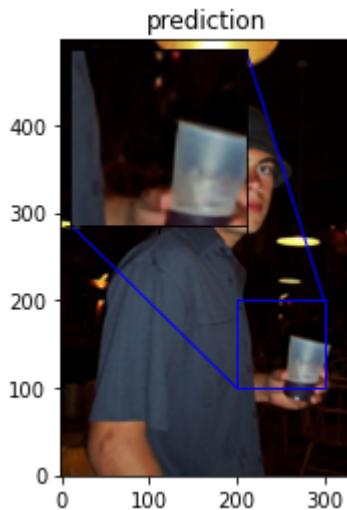
```
3438/3438 [=====] - 704s 205ms/step - loss: 0.0028 - val_loss: 0.0028  
Epoch 10/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.68
```



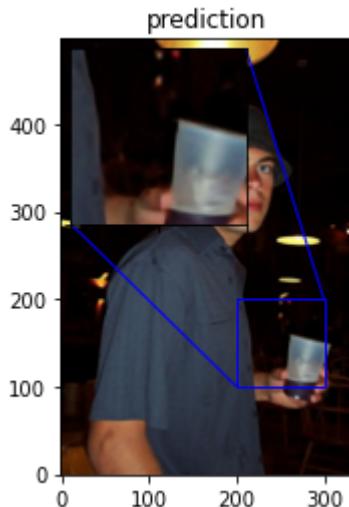
```
3438/3438 [=====] - 705s 205ms/step - loss: 0.0027 - val_loss: 0.0028  
Epoch 11/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.70
```



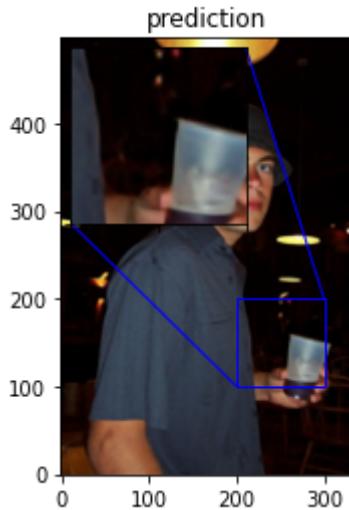
```
3438/3438 [=====] - 714s 208ms/step - loss: 0.0027 - val_loss: 0.0028  
Epoch 12/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.69
```



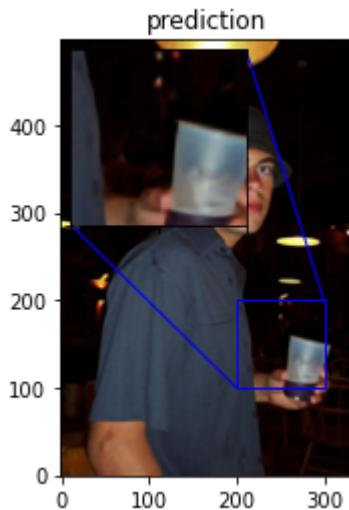
```
3438/3438 [=====] - 704s 205ms/step - loss: 0.0027 - val_loss: 0.0028  
Epoch 13/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.72
```



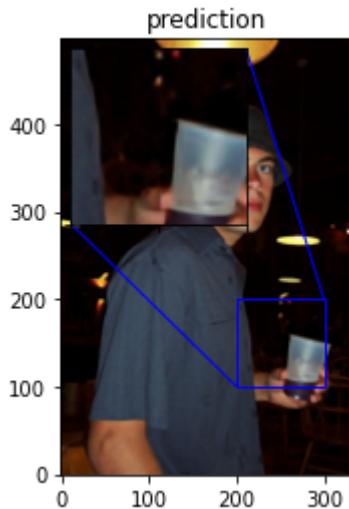
```
3438/3438 [=====] - 707s 206ms/step - loss: 0.0027 - val_loss: 0.0028  
Epoch 14/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.70
```



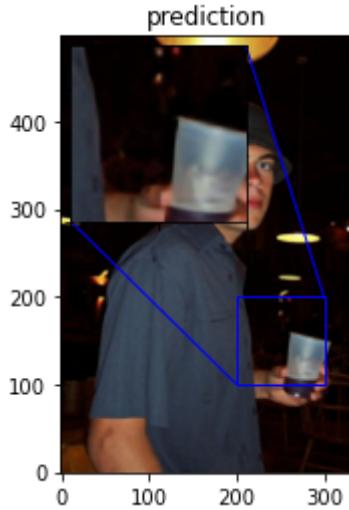
```
3438/3438 [=====] - 708s 206ms/step - loss: 0.0027 - val_loss: 0.0028  
Epoch 15/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.72
```



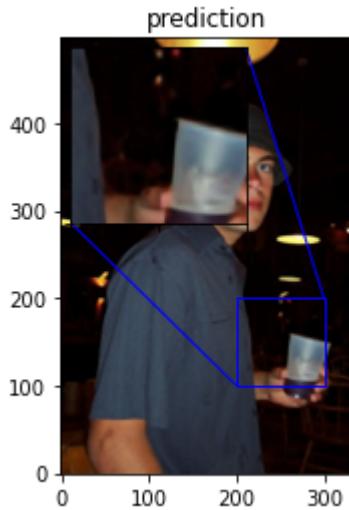
```
3438/3438 [=====] - 710s 207ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 16/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.73
```



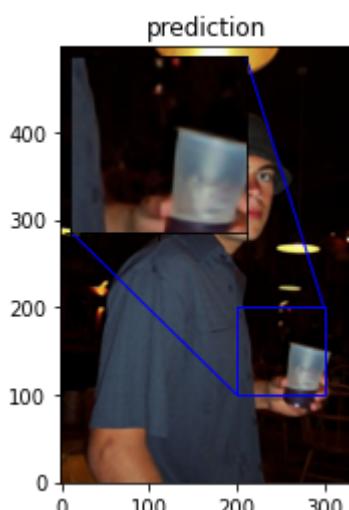
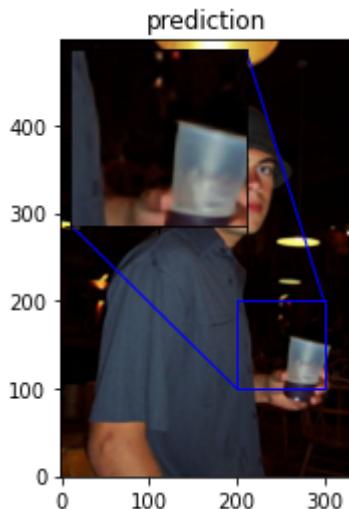
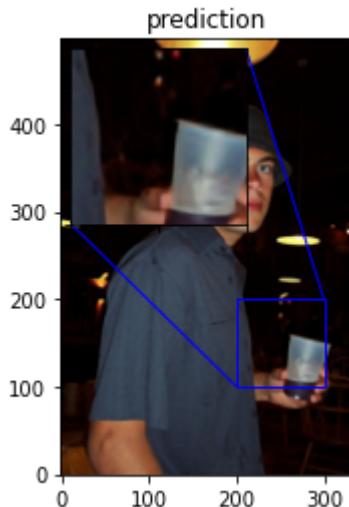
```
3438/3438 [=====] - 708s 206ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 17/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.73
```

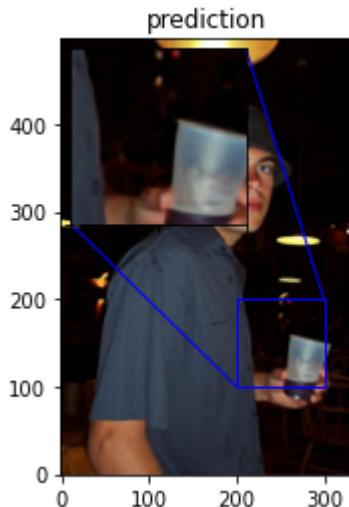


```
3438/3438 [=====] - 705s 205ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 18/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.74
```

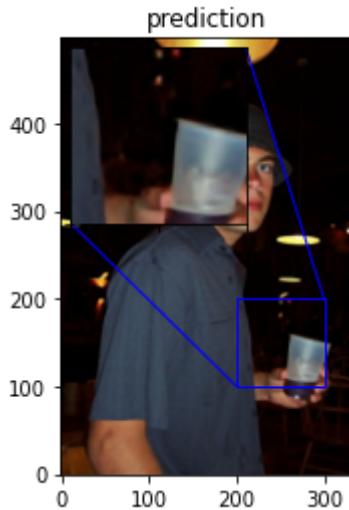


```
3438/3438 [=====] - 711s 207ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 19/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.73
```

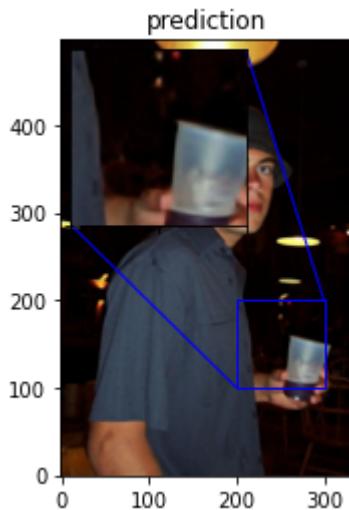




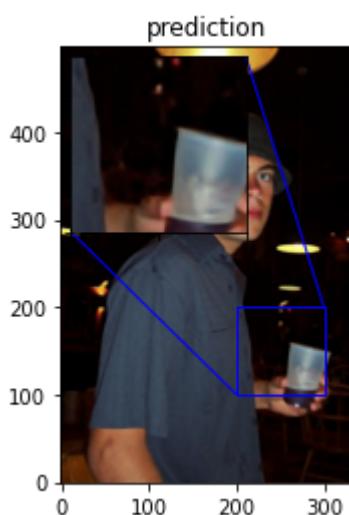
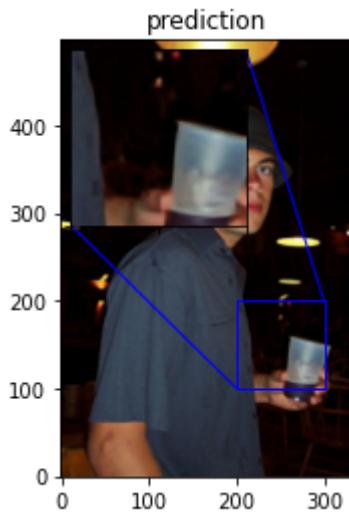
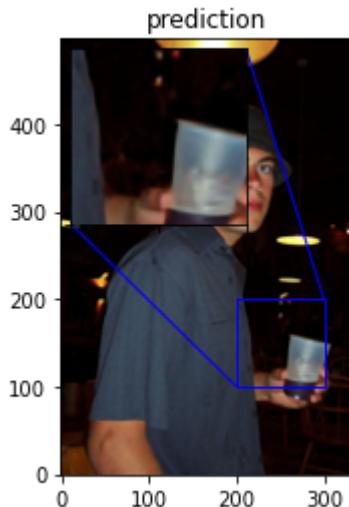
```
3438/3438 [=====] - 709s 206ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 23/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.73
```

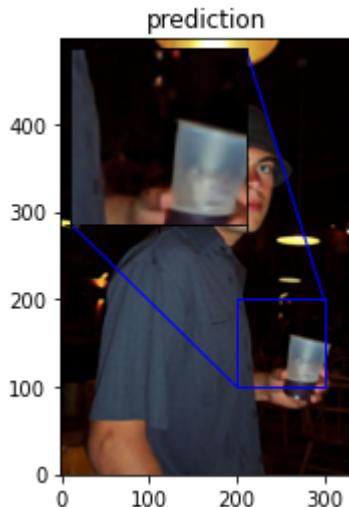


```
3438/3438 [=====] - 728s 212ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 24/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.76
```

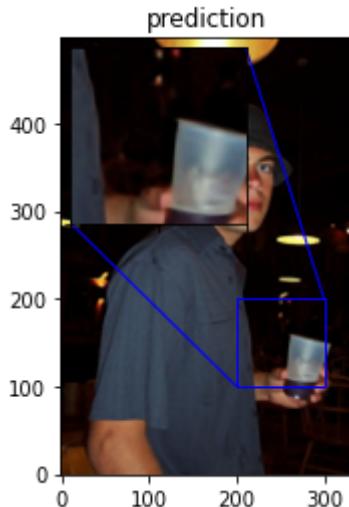


```
3438/3438 [=====] - 707s 206ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 25/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.74
```

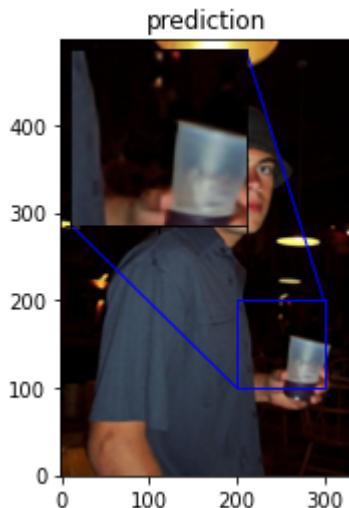




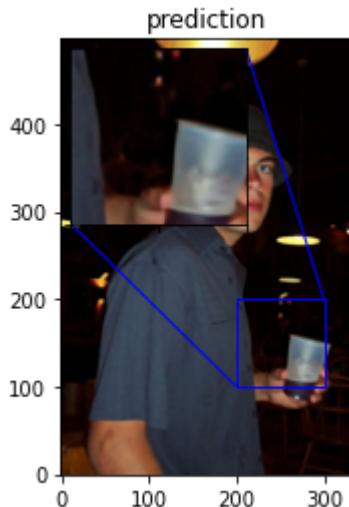
```
3438/3438 [=====] - 705s 205ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 29/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.76
```



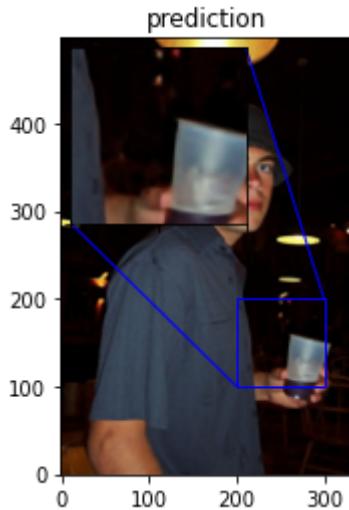
```
3438/3438 [=====] - 712s 207ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 30/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.73
```



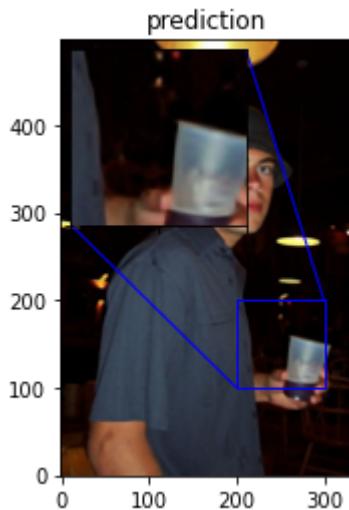
```
3438/3438 [=====] - 713s 207ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 31/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.75
```



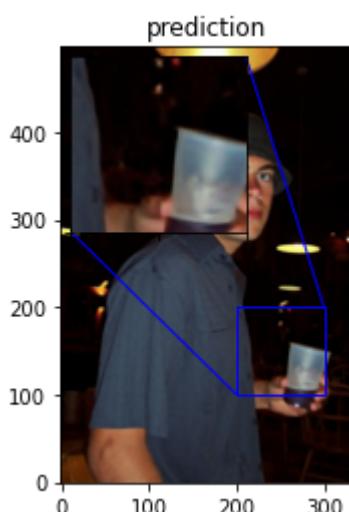
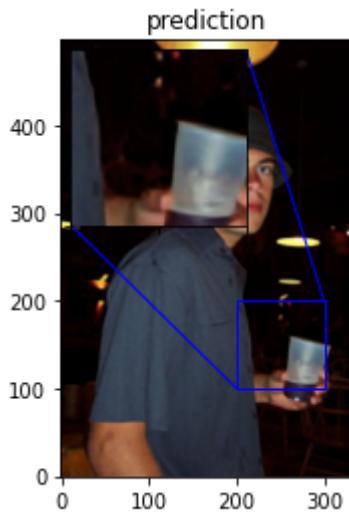
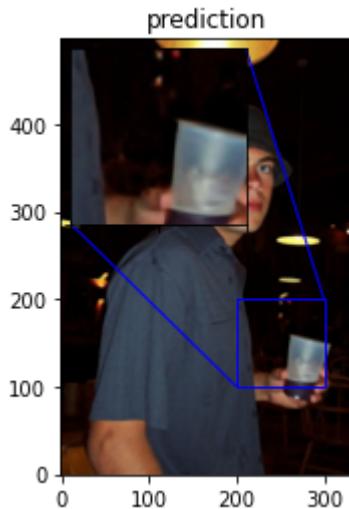
```
3438/3438 [=====] - 706s 205ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 32/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.76
```

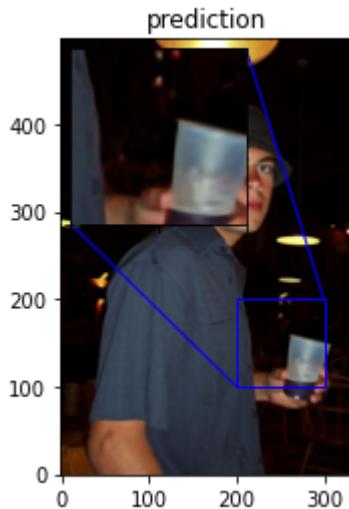


```
3438/3438 [=====] - 707s 206ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 33/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.78
```

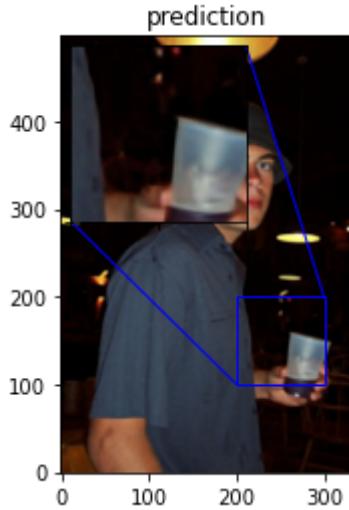


```
3438/3438 [=====] - 720s 209ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 34/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.75
```

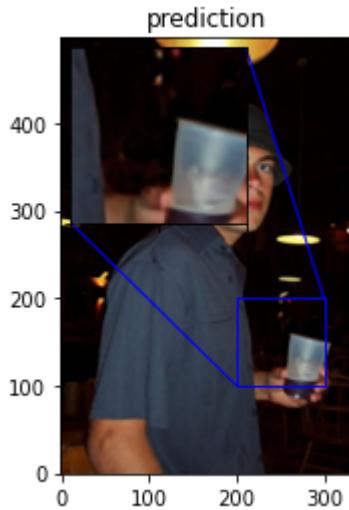




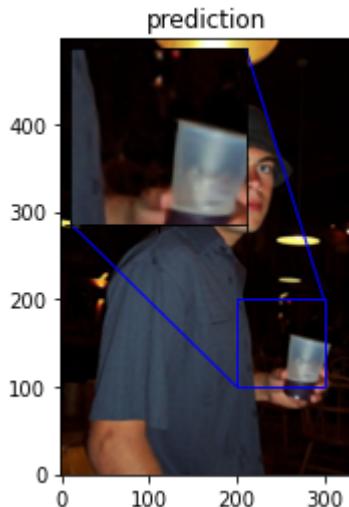
```
3438/3438 [=====] - 706s 205ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 38/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.77
```



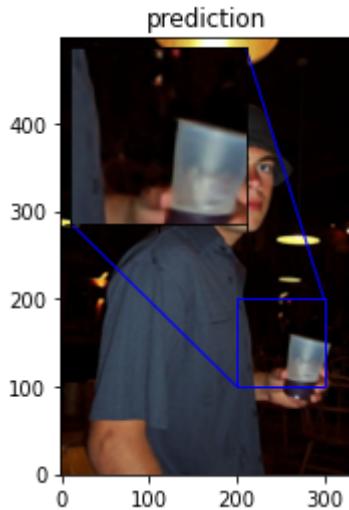
```
3438/3438 [=====] - 706s 205ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 39/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.75
```



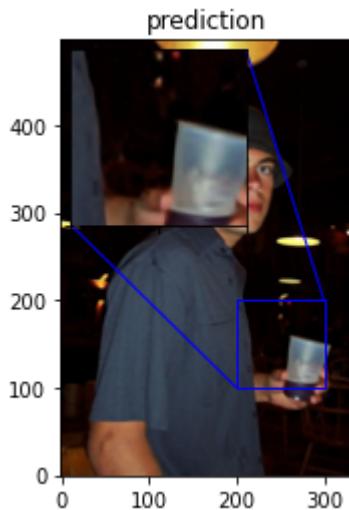
```
3438/3438 [=====] - 714s 208ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 40/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.77
```



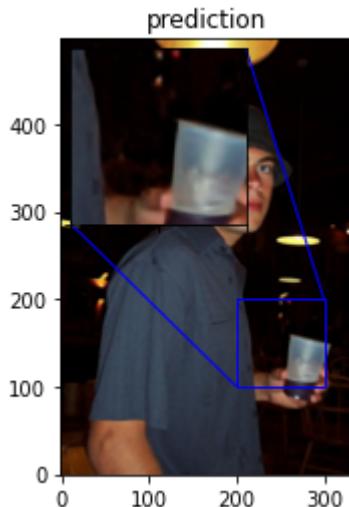
```
3438/3438 [=====] - 707s 206ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 41/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.77
```



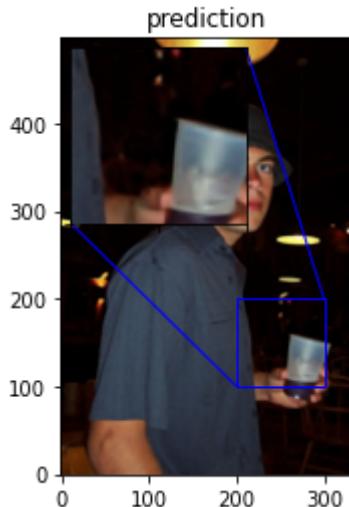
```
3438/3438 [=====] - 720s 210ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 42/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.77
```



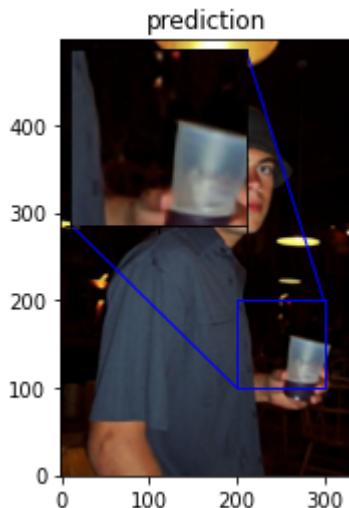
```
3438/3438 [=====] - 713s 207ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 43/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.79
```



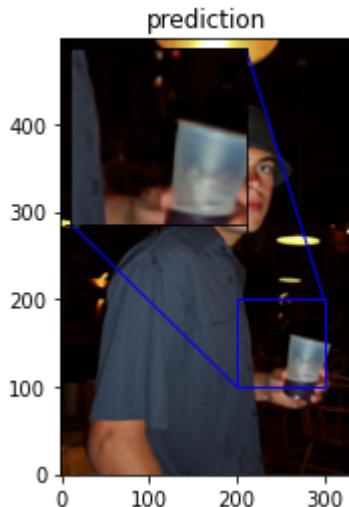
```
3438/3438 [=====] - 706s 205ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 44/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.78
```



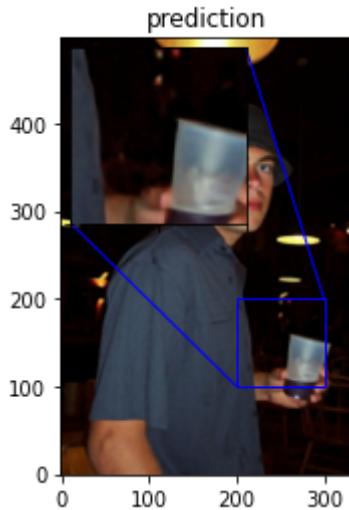
```
3438/3438 [=====] - 707s 206ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 45/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.76
```



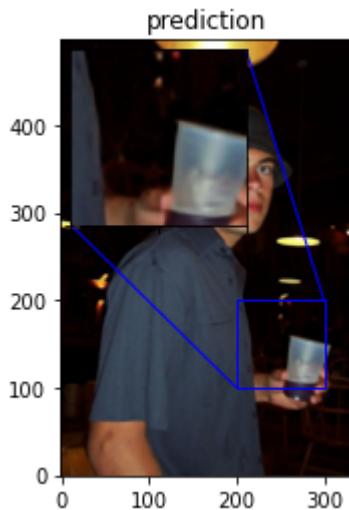
```
3438/3438 [=====] - 709s 206ms/step - loss: 0.0027 - val_loss: 0.0027  
Epoch 46/50  
3438/3438 [=====] - ETA: 0s - loss: 0.0026Mean PSNR for epoch: 25.78
```



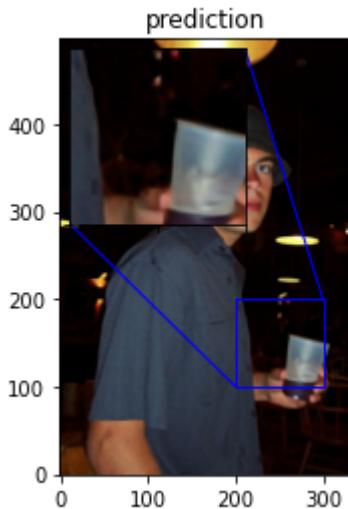
3438/3438 [=====] - 706s 205ms/step - loss: 0.0026 - val_loss: 0.0027
Epoch 47/50
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.77



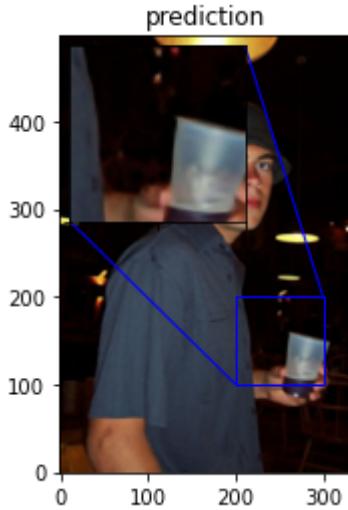
3438/3438 [=====] - 706s 205ms/step - loss: 0.0027 - val_loss: 0.0027
Epoch 48/50
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.77



3438/3438 [=====] - 707s 206ms/step - loss: 0.0027 - val_loss: 0.0027
Epoch 49/50
3438/3438 [=====] - ETA: 0s - loss: 0.0026Mean PSNR for epoch: 25.79



```
3438/3438 [=====] - 707s 206ms/step - loss: 0.0026 - val_loss: 0.0027
Epoch 50/50
3438/3438 [=====] - ETA: 0s - loss: 0.0027Mean PSNR for epoch: 25.78
```



```
3438/3438 [=====] - 713s 207ms/step - loss: 0.0027 - val_loss: 0.0027
INFO:tensorflow:Assets written to: ./model_final\assets
```

```
In [ ]: print("hello")
```

```
In [80]: total_bicubic_psnr = 0.0
total_test_psnr = 0.0

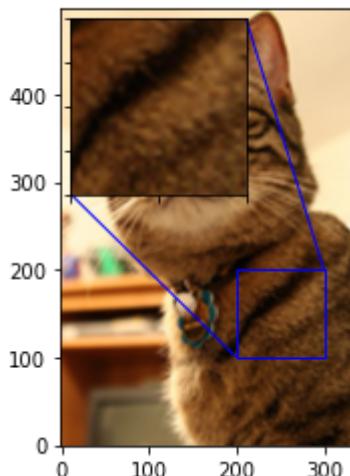
for index, test_img_path in enumerate(test_img_paths[50:60]):
    img = load_img(test_img_path)
    lowres_input = get_lowres_image(img, upscale_factor)
    w = lowres_input.size[0] * upscale_factor
    h = lowres_input.size[1] * upscale_factor
    highres_img = img.resize((w, h))
    prediction = upscale_image(model, lowres_input)
    lowres_img = lowres_input.resize((w, h))
    lowres_img_arr = img_to_array(lowres_img)
    highres_img_arr = img_to_array(highres_img)
    predict_img_arr = img_to_array(prediction)
    bicubic_psnr = tf.image.psnr(lowres_img_arr, highres_img_arr, max_val=255)
    test_psnr = tf.image.psnr(predict_img_arr, highres_img_arr, max_val=255)

    total_bicubic_psnr += bicubic_psnr
    total_test_psnr += test_psnr
```

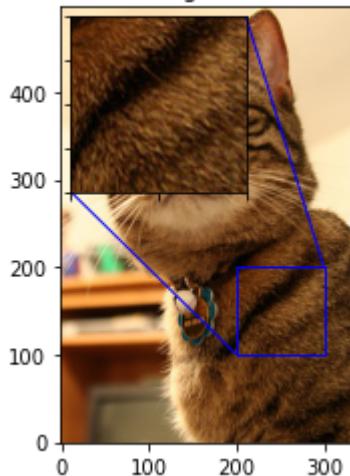
```
print(  
    "PSNR of low resolution image and high resolution image is %.4f" % bicubic_p  
)  
print("PSNR of predict and high resolution is %.4f" % test_psnr)  
plot_results(lowres_img, index, "lowres")  
plot_results(highres_img, index, "highres")  
plot_results(prediction, index, "prediction")  
  
print("Avg. PSNR of lowres images is %.4f" % (total_bicubic_psnr / 10))  
print("Avg. PSNR of reconstructions is %.4f" % (total_test_psnr / 10))
```

PSNR of low resolution image and high resolution image is 26.7161
PSNR of predict and high resolution is 27.1135

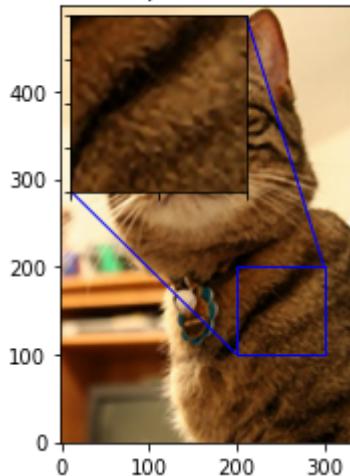
lowres



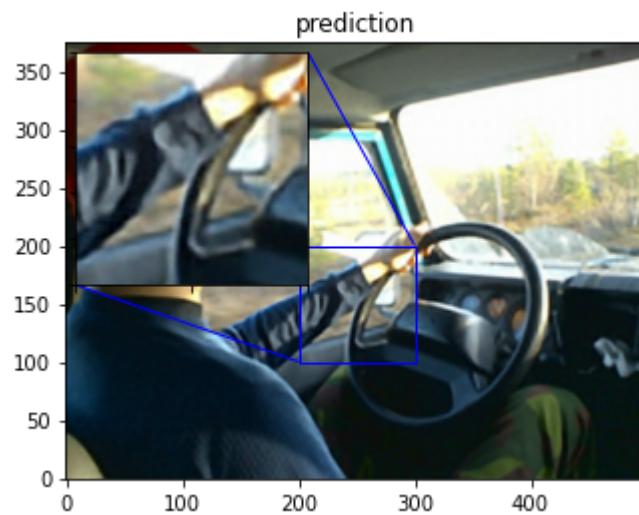
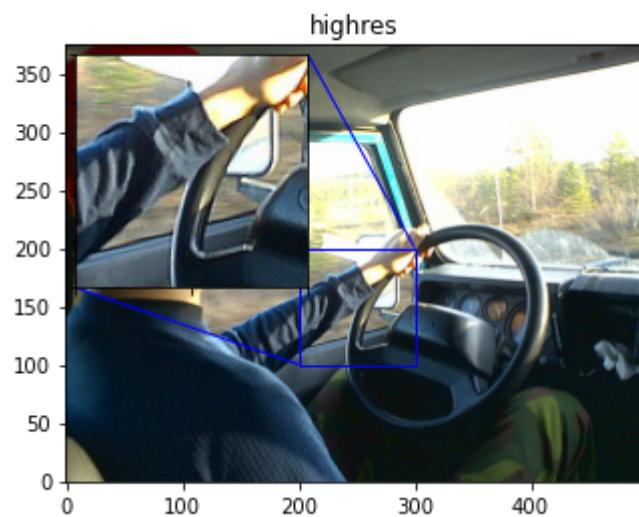
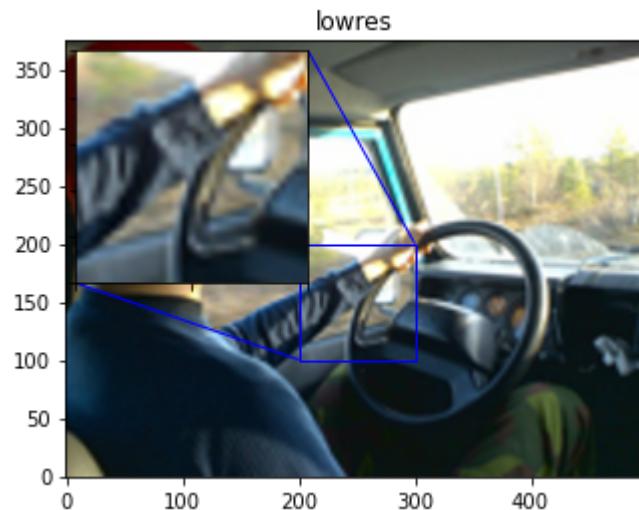
highres



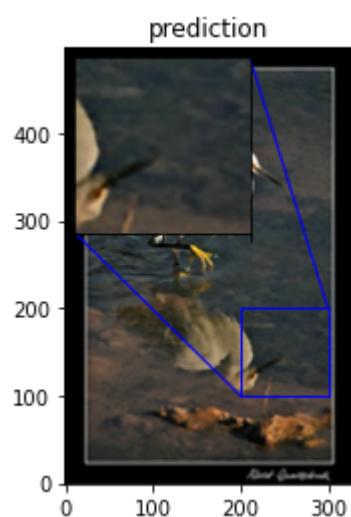
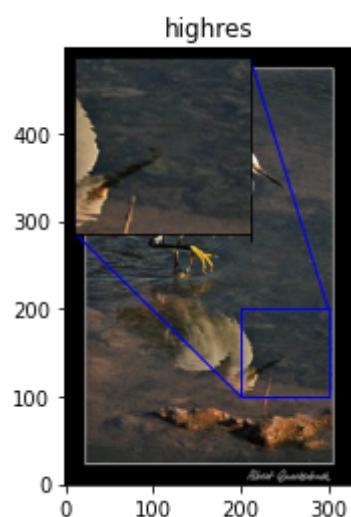
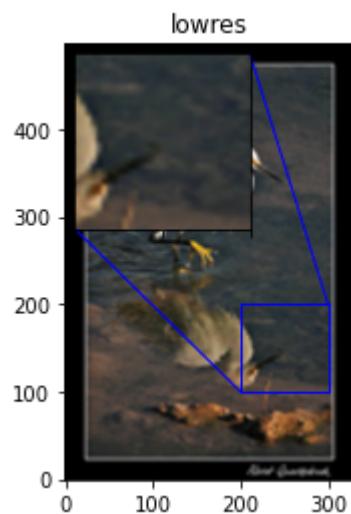
prediction



PSNR of low resolution image and high resolution image is 26.4780
PSNR of predict and high resolution is 27.6336



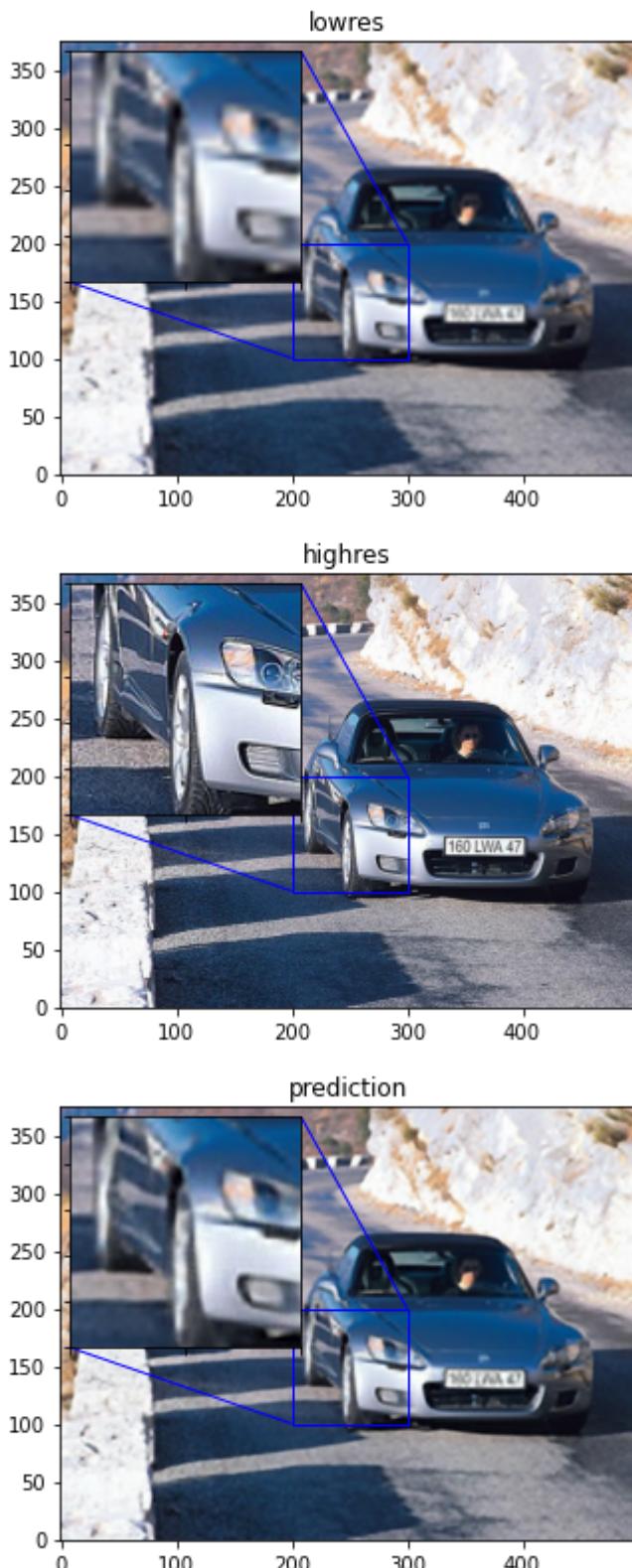
PSNR of low resolution image and high resolution image is 22.4373
PSNR of predict and high resolution is 23.4782



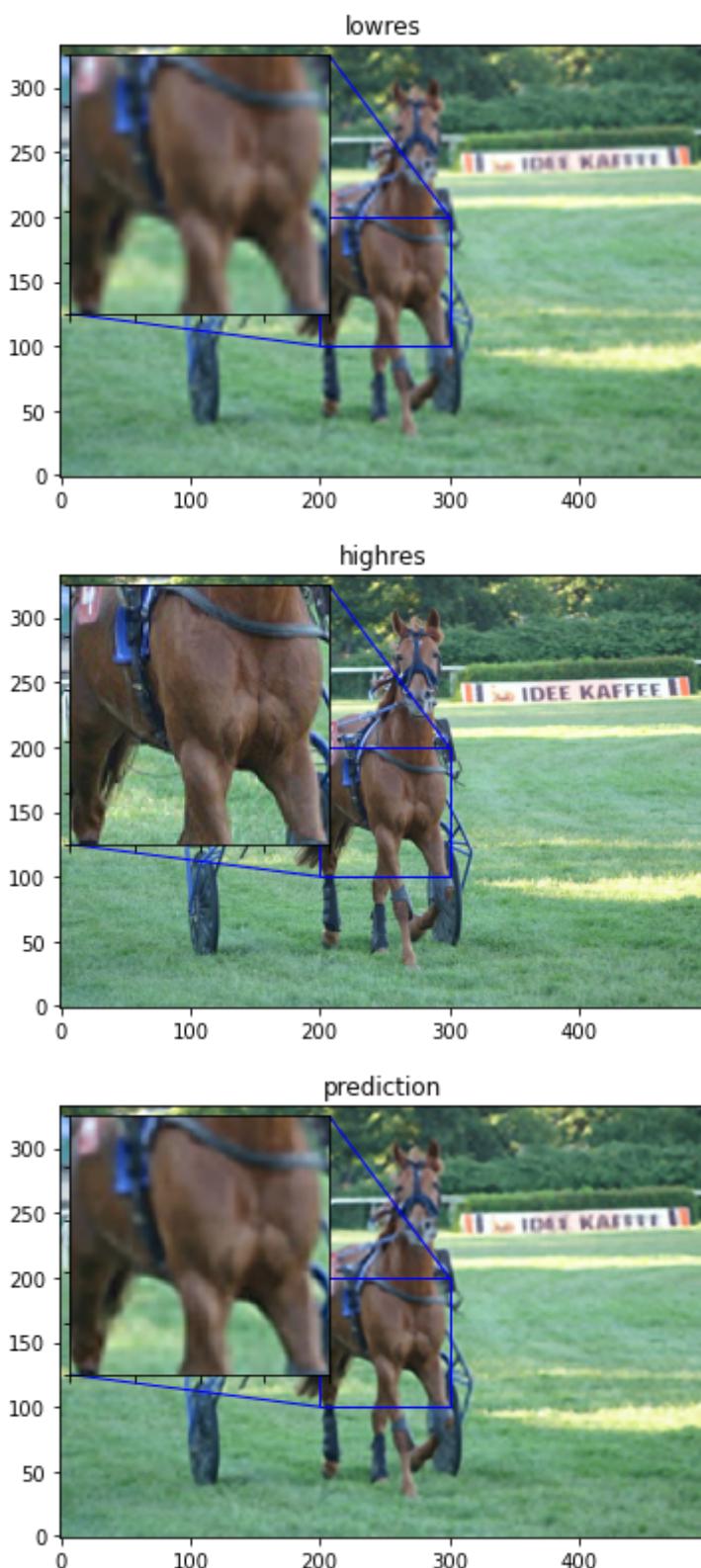
PSNR of low resolution image and high resolution image is 22.4946
PSNR of predict and high resolution is 22.8259



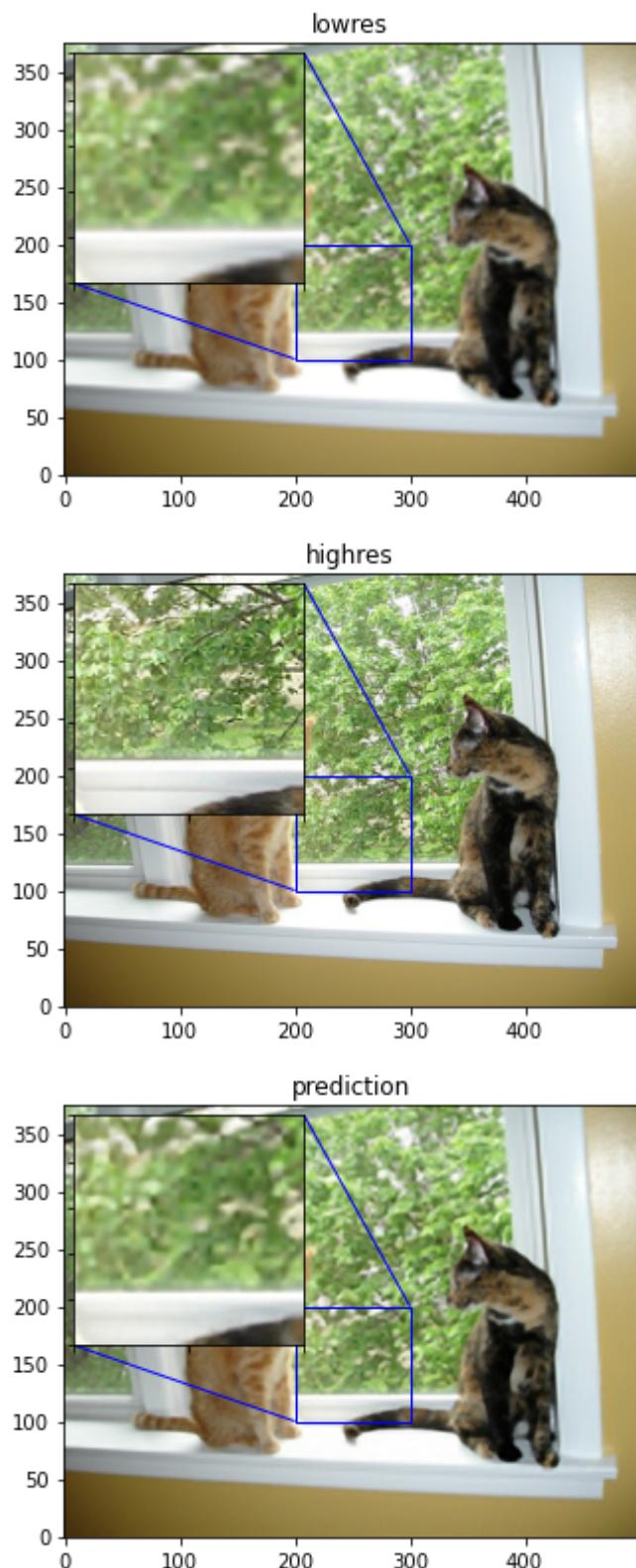
PSNR of low resolution image and high resolution image is 21.8302
PSNR of predict and high resolution is 22.4925



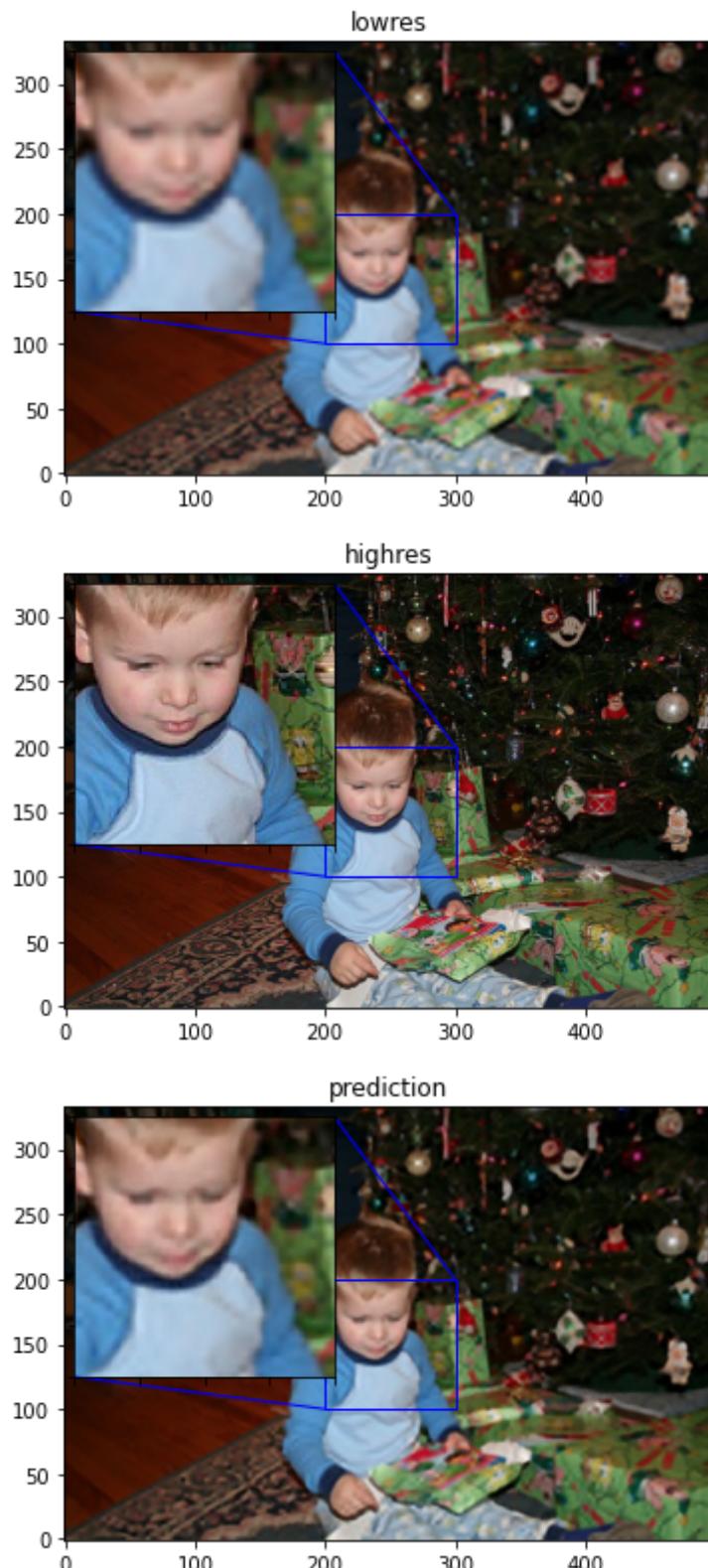
WARNING:tensorflow:5 out of the last 15 calls to <function Model.make_predict_function.<locals>.predict_function at 0x000002090351D0D0> triggered tf.function retracing. Tracing is expensive and the excessive number of tracings could be due to (1) creating @tf.function repeatedly in a loop, (2) passing tensors with different shapes, (3) passing Python objects instead of tensors. For (1), please define your @tf.function outside of the loop. For (2), @tf.function has experimental_relax_shapes=True option that relaxes argument shapes that can avoid unnecessary retracing. For (3), please refer to https://www.tensorflow.org/guide/function#controlling_retracing and https://www.tensorflow.org/api_docs/python/tf/function for more details.
 PSNR of low resolution image and high resolution image is 24.2919
 PSNR of predict and high resolution is 25.0663



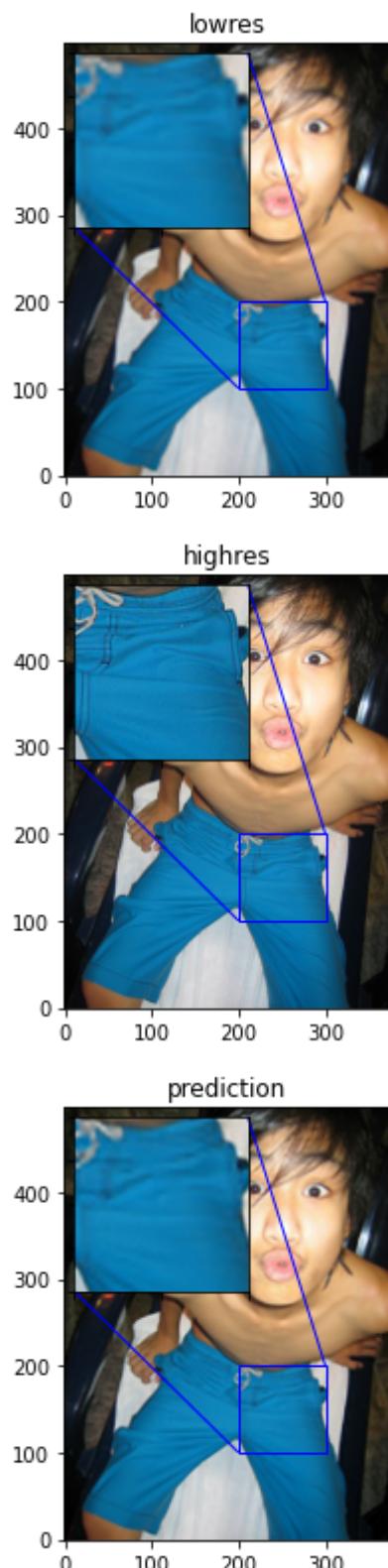
PSNR of low resolution image and high resolution image is 23.2296
PSNR of predict and high resolusion is 23.6755



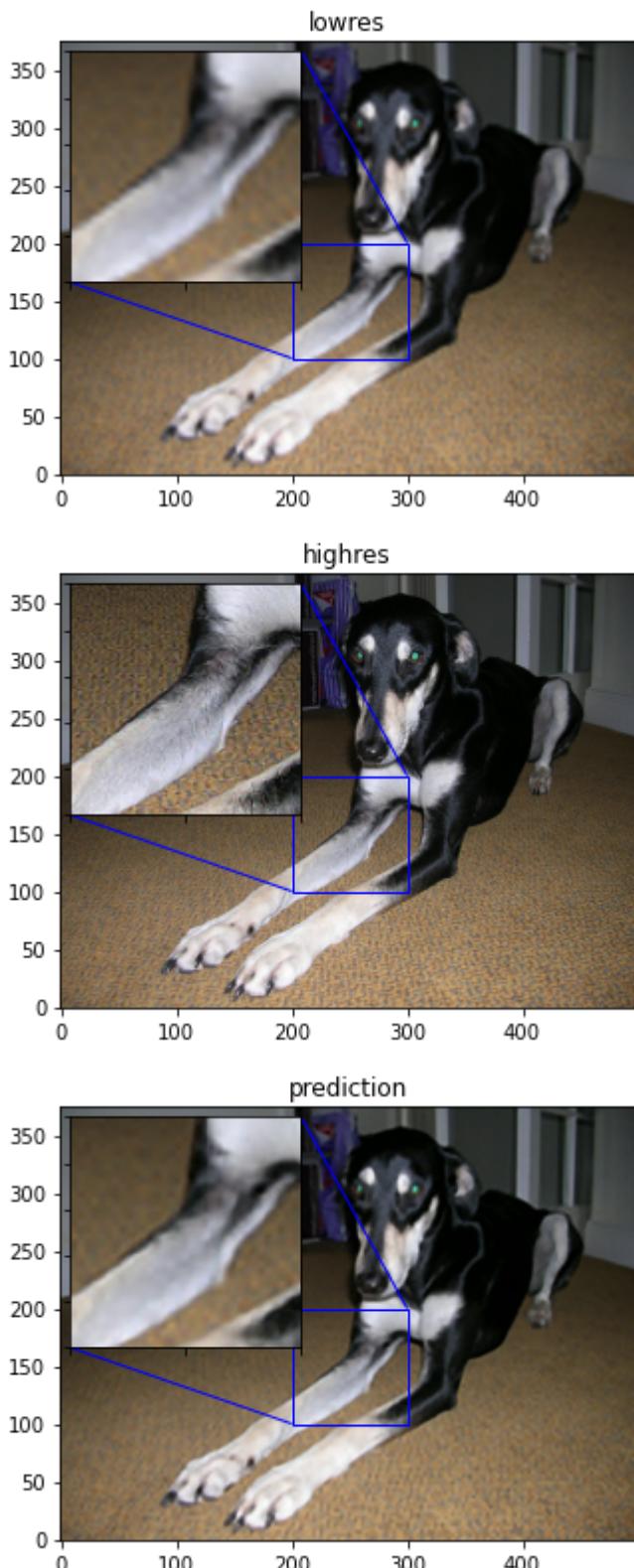
PSNR of low resolution image and high resolution image is 24.3504
PSNR of predict and high resolusion is 24.8538



PSNR of low resolution image and high resolution image is 28.1352
PSNR of predict and high resolusion is 28.9987



PSNR of low resolution image and high resolution image is 26.5851
PSNR of predict and high resolution is 26.9208



Avg. PSNR of lowres images is 24.6548
 Avg. PSNR of reconstructions is 25.3059

In [81]:

```
model.save('./model_epoch1')
```

INFO:tensorflow:Assets written to: ./model_epoch1/assets

In [44]:

'pwd' is not recognized as an internal or external command,
 operable program or batch file.

In [46]:

```
# from tensorflow import keras
# modell = keras.models.load_model('./model_')
```

In [47]:

```
# total_bicubic_psnr = 0.0
# total_test_psnr = 0.0

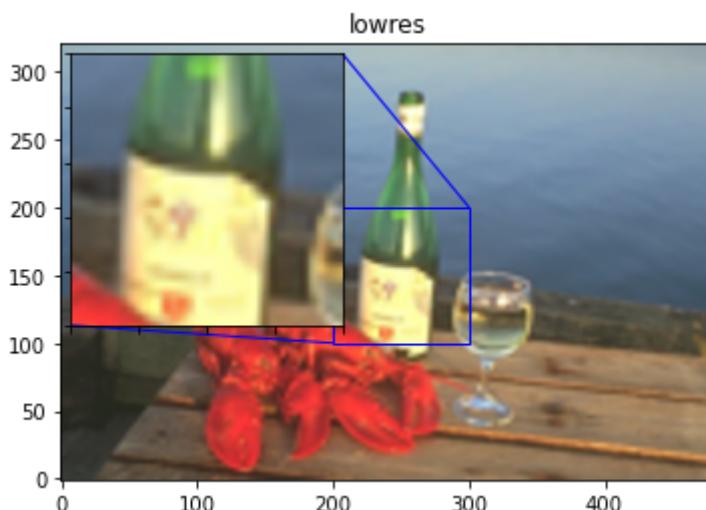
# for index, test_img_path in enumerate(test_img_paths[50:60]):
#     img = Load_img(test_img_path)
#     lowres_input = get_lowres_image(img, upscale_factor)
#     w = lowres_input.size[0] * upscale_factor
#     h = lowres_input.size[1] * upscale_factor
#     highres_img = img.resize((w, h))
#     prediction = upscale_image(modell, lowres_input)
#     lowres_img = lowres_input.resize((w, h))
#     lowres_img_arr = img_to_array(lowres_img)
#     highres_img_arr = img_to_array(highres_img)
#     predict_img_arr = img_to_array(prediction)
#     bicubic_psnr = tf.image.psnr(lowres_img_arr, highres_img_arr, max_val=255)
#     test_psnr = tf.image.psnr(predict_img_arr, highres_img_arr, max_val=255)

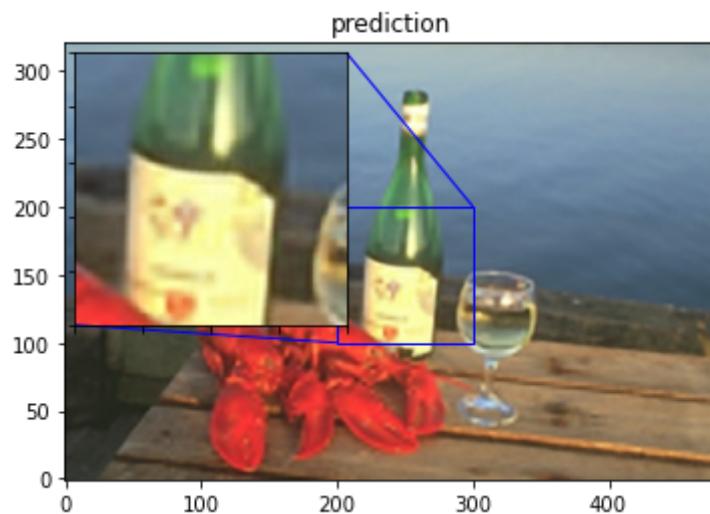
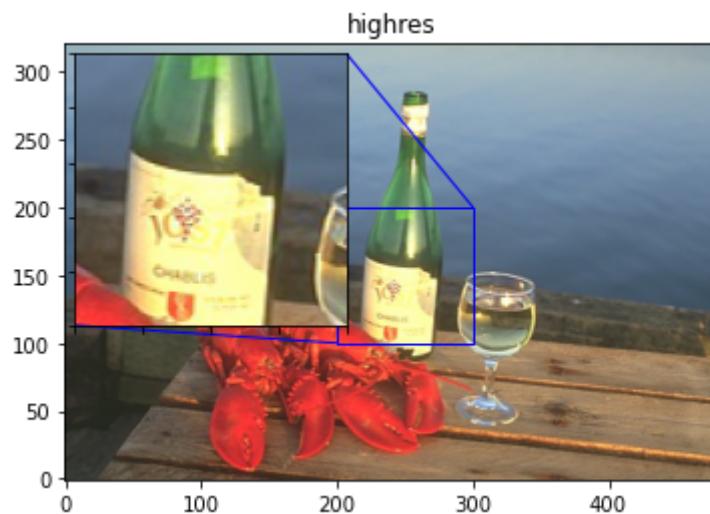
#     total_bicubic_psnr += bicubic_psnr
#     total_test_psnr += test_psnr

#     print(
#         "PSNR of Low resolution image and high resolution image is %.4f" % bicubic_psnr
#     )
#     print("PSNR of predict and high resolution is %.4f" % test_psnr)
#     plot_results(lowres_img, index, "Lowres")
#     plot_results(highres_img, index, "highres")
#     plot_results(prediction, index, "prediction")

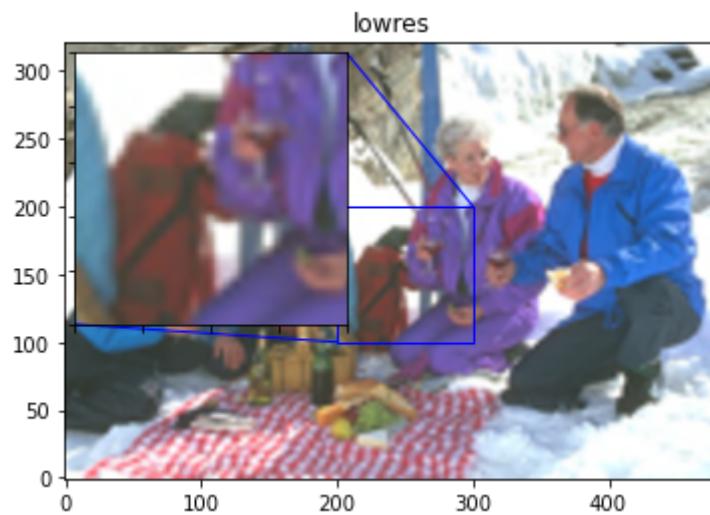
# print("Avg. PSNR of Lowres images is %.4f" % (total_bicubic_psnr / 10))
# print("Avg. PSNR of reconstructions is %.4f" % (total_test_psnr / 10))
```

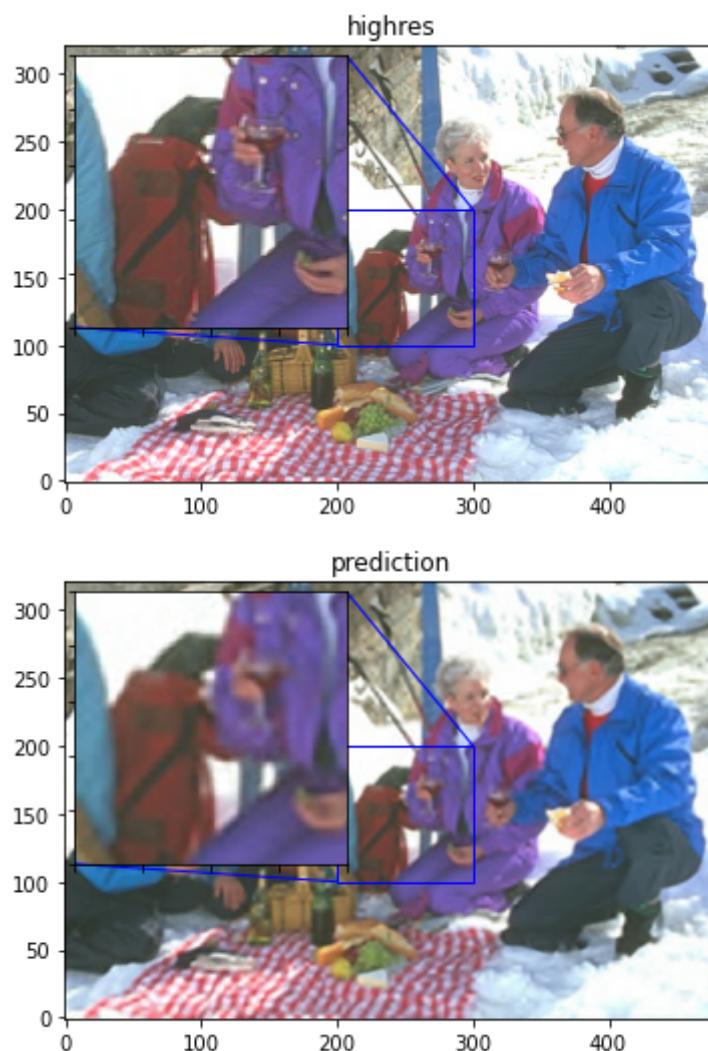
PSNR of low resolution image and high resolution image is 29.8502
 PSNR of predict and high resolution is 30.2815



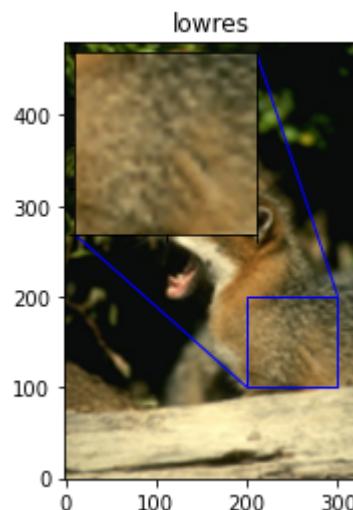


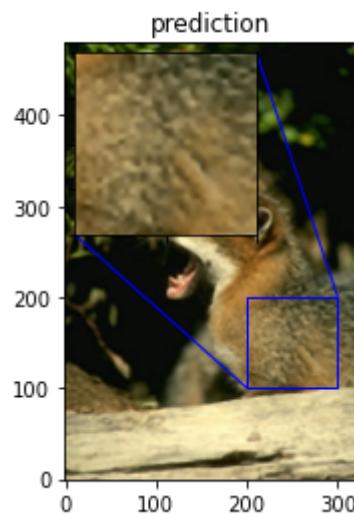
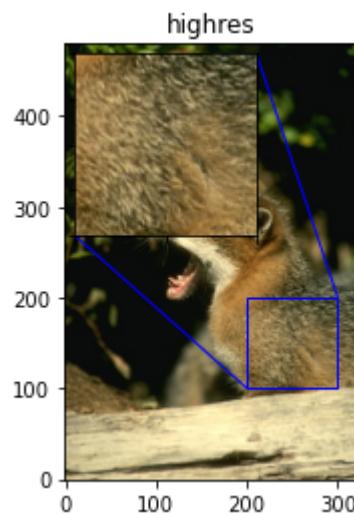
PSNR of low resolution image and high resolution image is 24.9783
PSNR of predict and high resolution is 25.8959



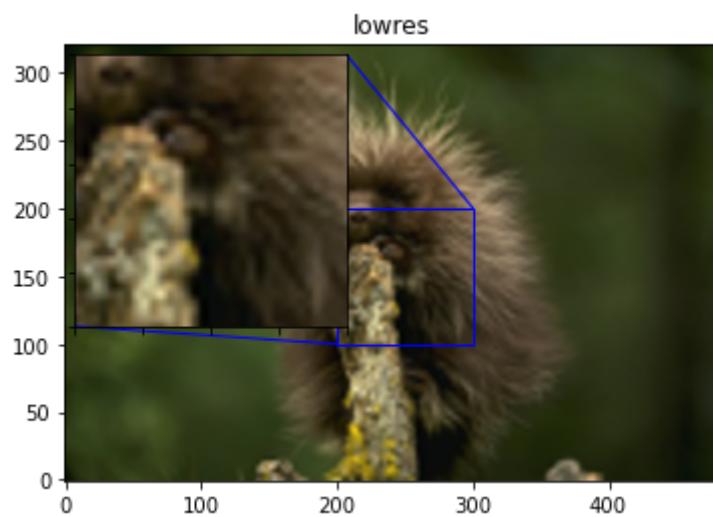


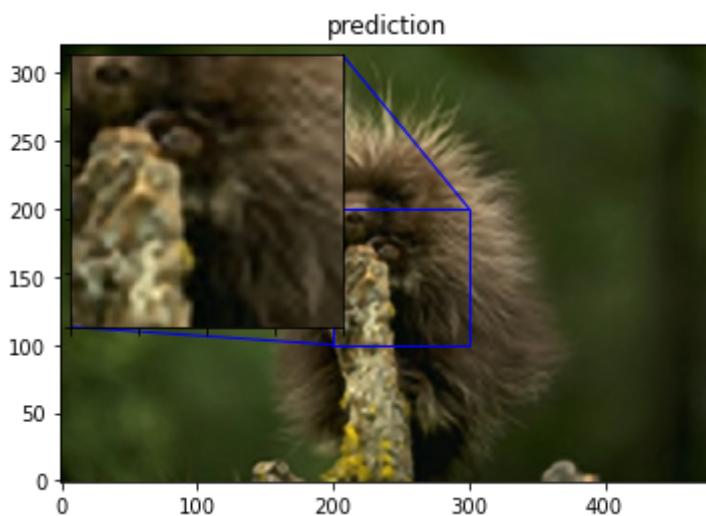
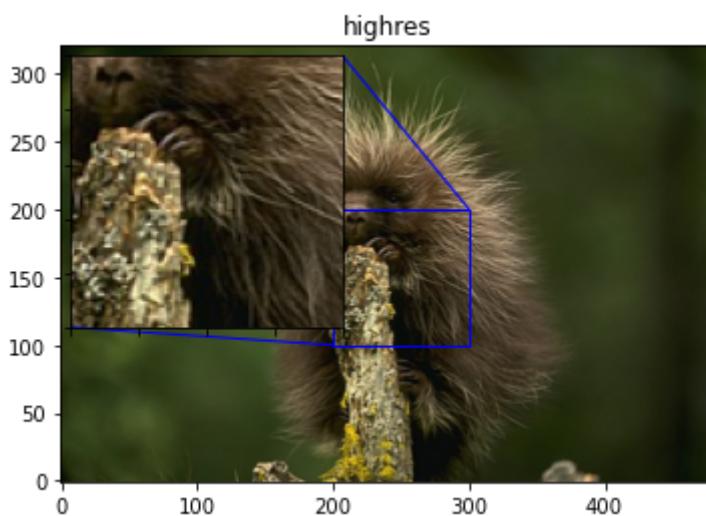
PSNR of low resolution image and high resolution image is 27.7724
PSNR of predict and high resoln is 28.3722



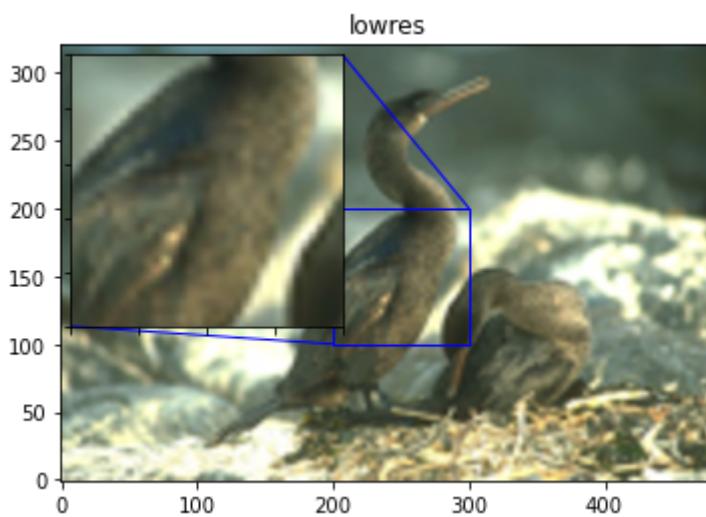


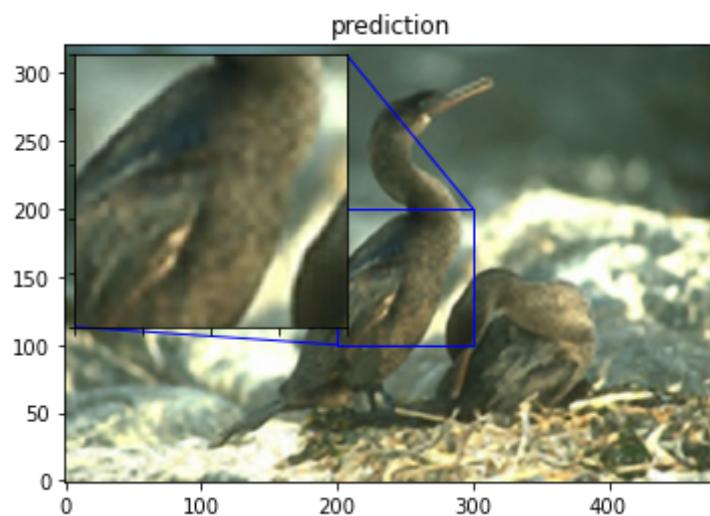
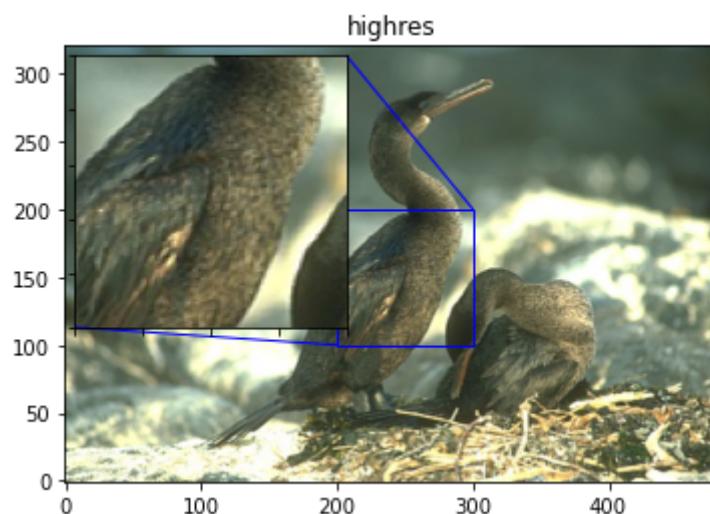
PSNR of low resolution image and high resolution image is 28.0314
PSNR of predict and high resolution is 28.2546



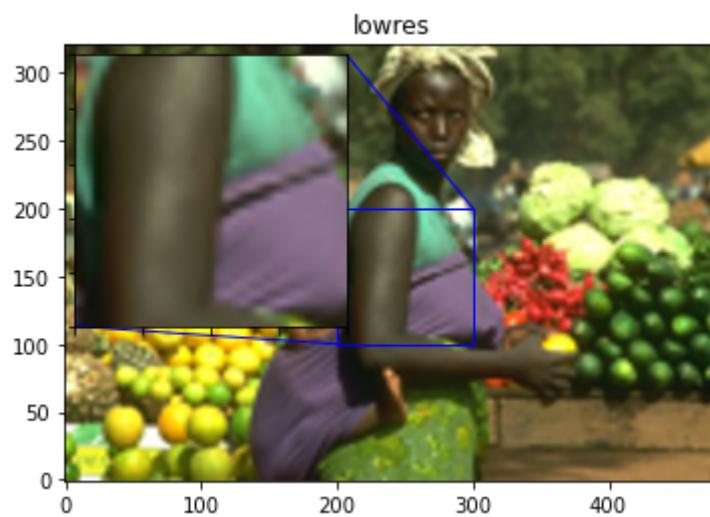


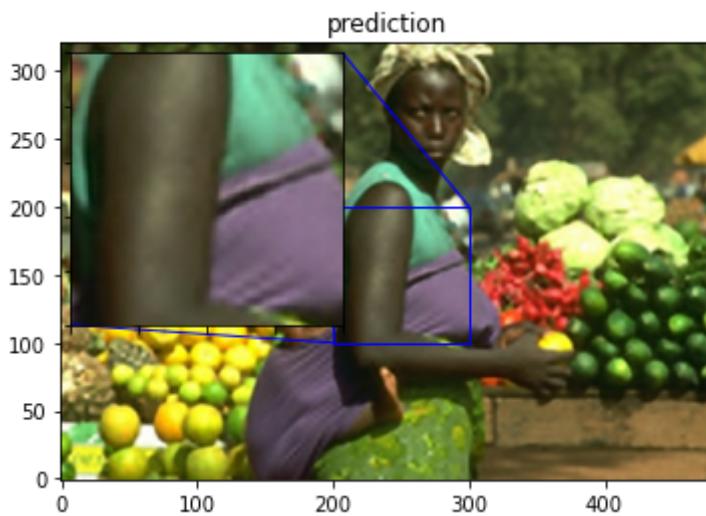
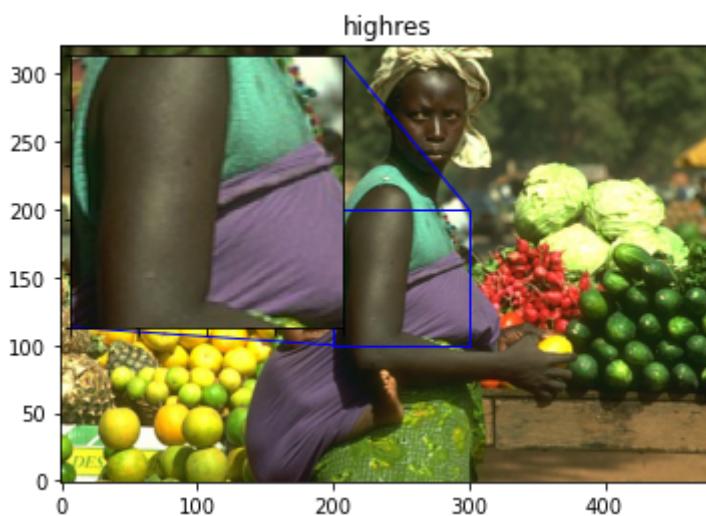
PSNR of low resolution image and high resolution image is 25.7630
PSNR of predict and high resolution is 26.3259



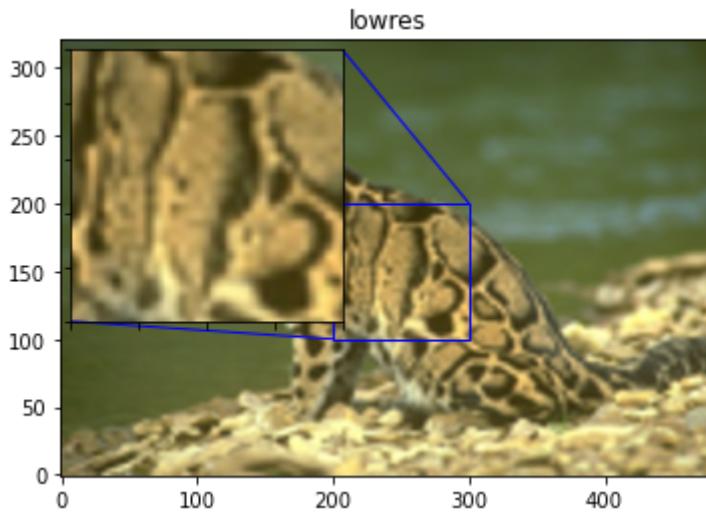


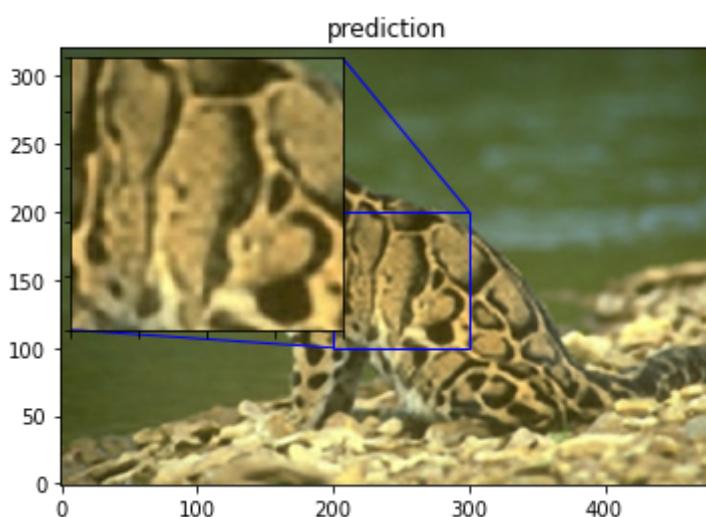
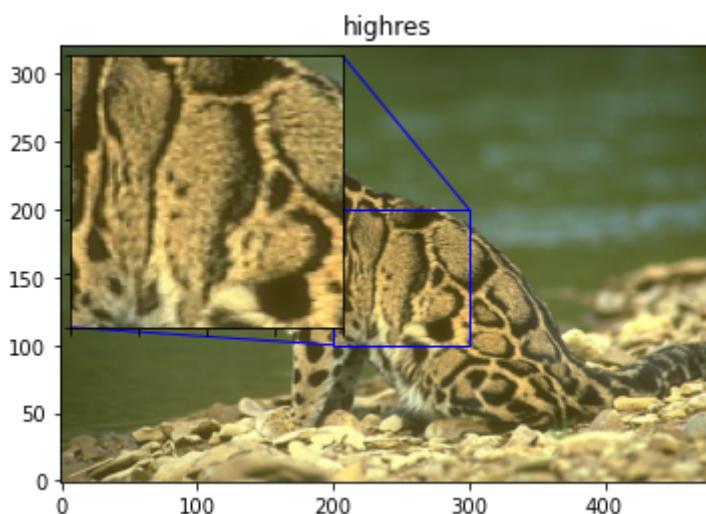
PSNR of low resolution image and high resolution image is 25.7874
PSNR of predict and high resolusion is 26.5285



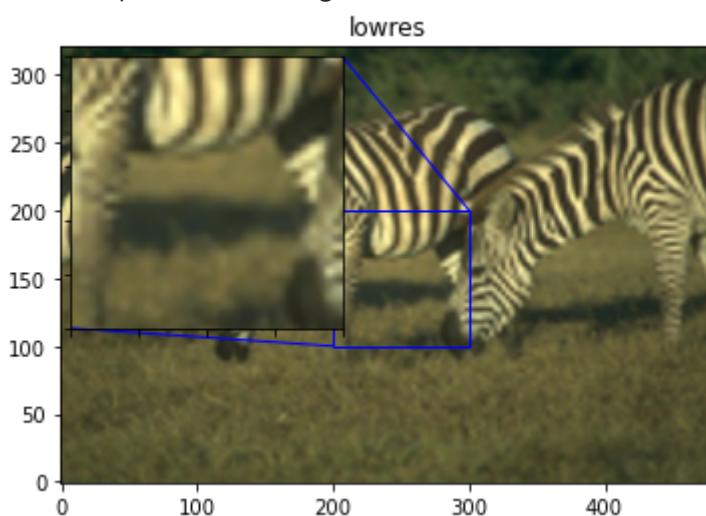


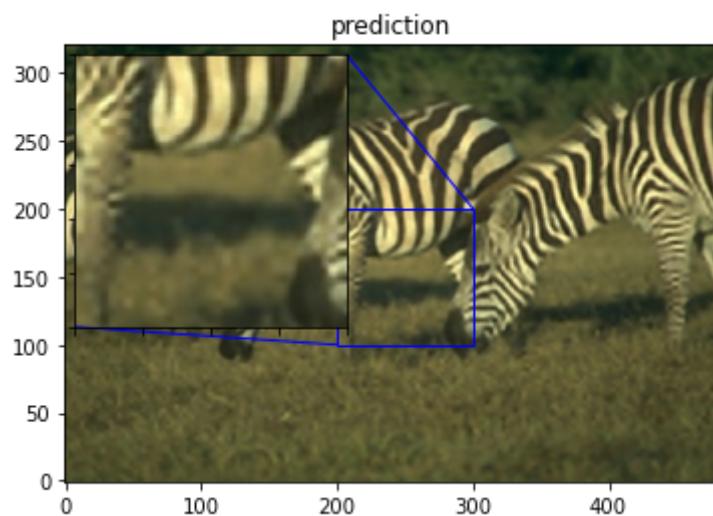
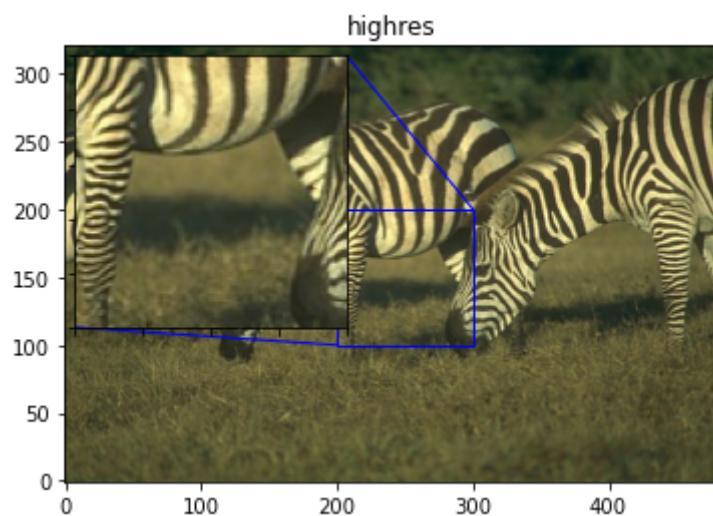
PSNR of low resolution image and high resolution image is 26.2512
PSNR of predict and high resolusion is 27.1253



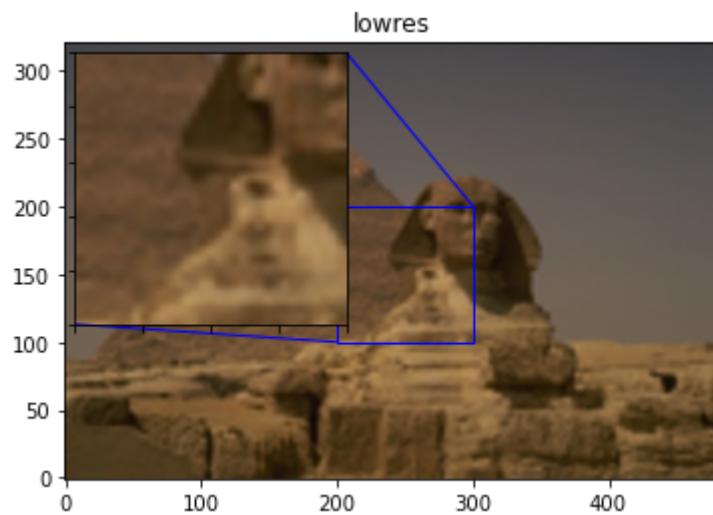


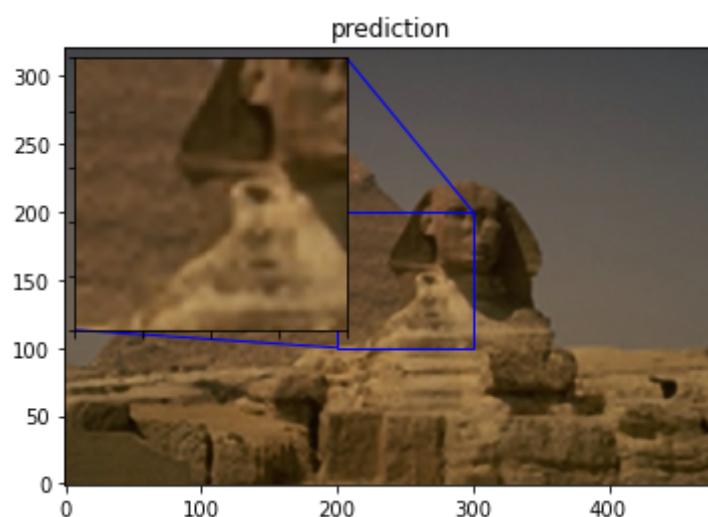
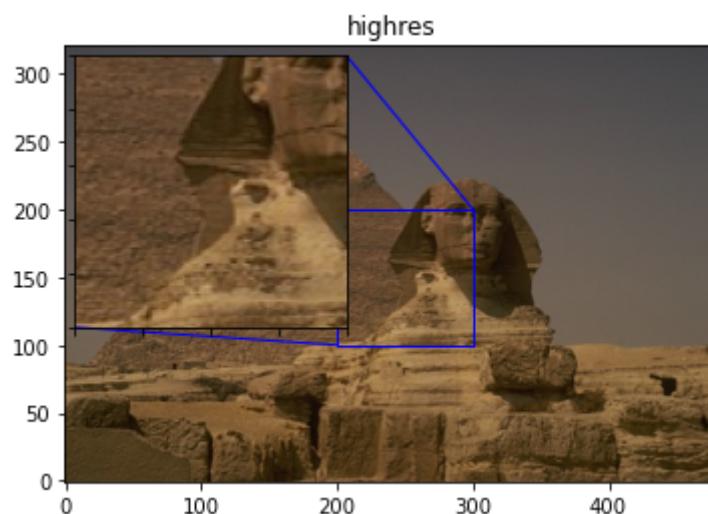
PSNR of low resolution image and high resolution image is 23.3820
PSNR of predict and high resolusion is 24.6839



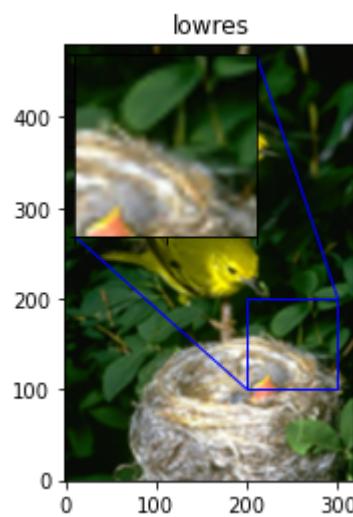


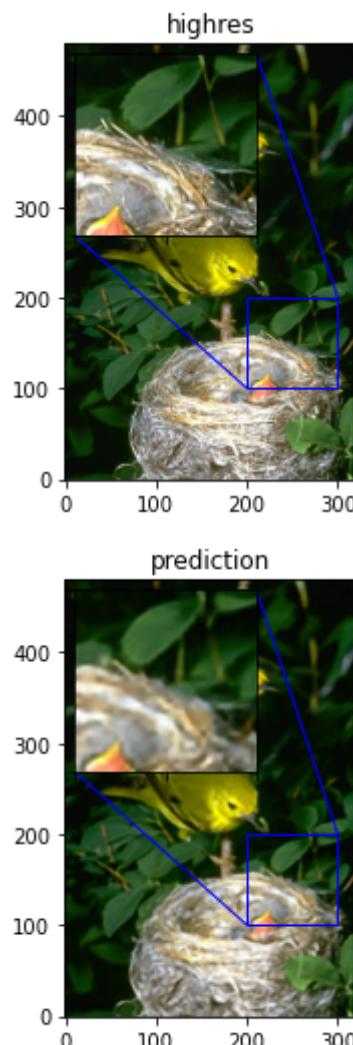
PSNR of low resolution image and high resolution image is 29.8914
PSNR of predict and high resoln is 30.0418





PSNR of low resolution image and high resolution image is 25.1712
PSNR of predict and high resolusion is 25.6863





Avg. PSNR of lowres images is 26.6879

Avg. PSNR of reconstructions is 27.3196

In []: