

```
!git clone https://github.com/ClawSwipe/lrp_toolbox.git  
↳ fatal: destination path 'lrp_toolbox' already exists and is not an empty directory.  
  
cd lrp_toolbox/python  
↳ /content/lrp_toolbox/python
```

> Model IO

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> Data IO

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> Render

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> Training Test MNIST

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▼ LRP Demo MNIST

```
...  
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@date: 14.08.2015  
@version: 1.0  
@copyright: Copyright (c) 2015-2017, Sebastian Lapuschkin, Alexander Binder, Gregoire Mor  
@license : BSD-2-Clause
```

The purpose of this module is to demonstrate the process of obtaining pixel-wise explanations.

The module first loads a pre-trained neural network model and the MNIST test set with labels. The data is then randomly permuted and for the first 10 samples due to the permuted order,

by attributing relevance values to each of the input pixels.

finally, the resulting heatmap is rendered as an image and (over)written out to disk and c
LRP works by:

Doing a standard forward pass on an input sample to make a prediction.

Starting with the output prediction as total relevance, then propagating that relevance

At each layer, relevance is redistributed to the layer below based on specific rules (

Finally, each input feature gets a relevance score that tells you how much it contribu

...

```
import matplotlib.pyplot as plt
import time
import numpy
import numpy as np
import importlib.util as imp
if imp.find_spec("cupy"): #use cupy for GPU support if available
    import cupy
    import cupy as np
na = np.newaxis
import render
from skimage.transform import resize

#load a neural network, as well as the MNIST test data and some labels
nn = model_io.read('../models/MNIST/long-tanh.nn') # 99.16% prediction accuracy
nn.drop_softmax_output_layer() #drop softmax output layer for analyses

X = data_io.read('../data/MNIST/test_images.npy')
Y = data_io.read('../data/MNIST/test_labels.npy')

# transfer pixel values from [0 255] to [-1 1] to satisfy the expected input / training pa
X = X / 127.5 - 1

# transform numeric class labels to vector indicator for uniformity. assume presence of al
I = Y[:,0].astype(int)
Y = np.zeros([X.shape[0],np.unique(Y).size])
Y[np.arange(Y.shape[0]),I] = 1

acc = np.mean(np.argmax(nn.forward(X), axis=1) == np.argmax(Y, axis=1))
if not np == numpy: # np=cupy
    acc = np.asarray(acc)
print('model test accuracy is: {:.4f}'.format(acc))

#permute data order for demonstration. or not. your choice.
I = np.arange(X.shape[0])
```

```

#I = np.random.permutation(I)

#predict and perform LRP for the 10 first samples
for i in I[:10]:
    x = X[na,i,:]

    #forward pass and prediction
    ypred = nn.forward(x)
    print('True Class: ', np.argmax(Y[i]))
    print('Predicted Class:', np.argmax(ypred), '\n')

    #prepare initial relevance to reflect the model's dominant prediction (ie depopulate r
    mask = np.zeros_like(ypred)
    mask[:,np.argmax(ypred)] = 1
    Rinit = ypred*mask

    #compute first layer relevance according to prediction
    #R = nn.lrp(Rinit)                      #as Eq(56) from DOI: 10.1371/journal.pone.0130146
    R = nn.lrp(Rinit,'epsilon',0.01)        #as Eq(58) from DOI: 10.1371/journal.pone.0130140
    #R = nn.lrp(Rinit,'alphabeta',2)         #as Eq(60) from DOI: 10.1371/journal.pone.0130140

    #R = nn.lrp(ypred*Y[na,i]) #compute first layer relevance according to the true class
    ...
    yselect = 3
    yselect = (np.arange(Y.shape[1])[na,:] == yselect)*1.
    R = nn.lrp(ypred*yselect) #compute first layer relvance for an arbitrarily selected cl
    ...

#undo input normalization for digit drawing. get it back to range [0,1] per pixel
x = (x+1.)/2.

if not np == numpy: # np=cupy
    x = np.asarray(x)
    R = np.asarray(R)

#render input and heatmap as rgb images
digit = render.digit_to_rgb(x, scaling=3)

# Similarly, reshape the relevance map if it's flattened (1D array)
R_numpy = R.reshape(28, 28)

# Display the relevance map as a heatmap
plt.imshow(R_numpy, cmap='seismic', interpolation='none')
plt.colorbar() # Show color bar to indicate relevance magnitude
plt.title("Relevance Heatmap")
plt.axis('off')
plt.show()

# Loop through all custom heatmaps

```

```
for cmap_name in render.custom_maps.keys():
    print(f'Applying custom heatmap: {cmap_name}')
    hm = render.hm_to_rgb(R, X=x, scaling=3, cmap=cmap_name, sigma=2)
    digit_hm = render.save_image([digit, hm], f'../heatmap_{i}_{cmap_name}.png')

    # Display (optional)
    plt.imshow(digit_hm, interpolation='none')
    plt.title(f'Sample #{i} - {cmap_name}')
    plt.axis('off')
    plt.show()

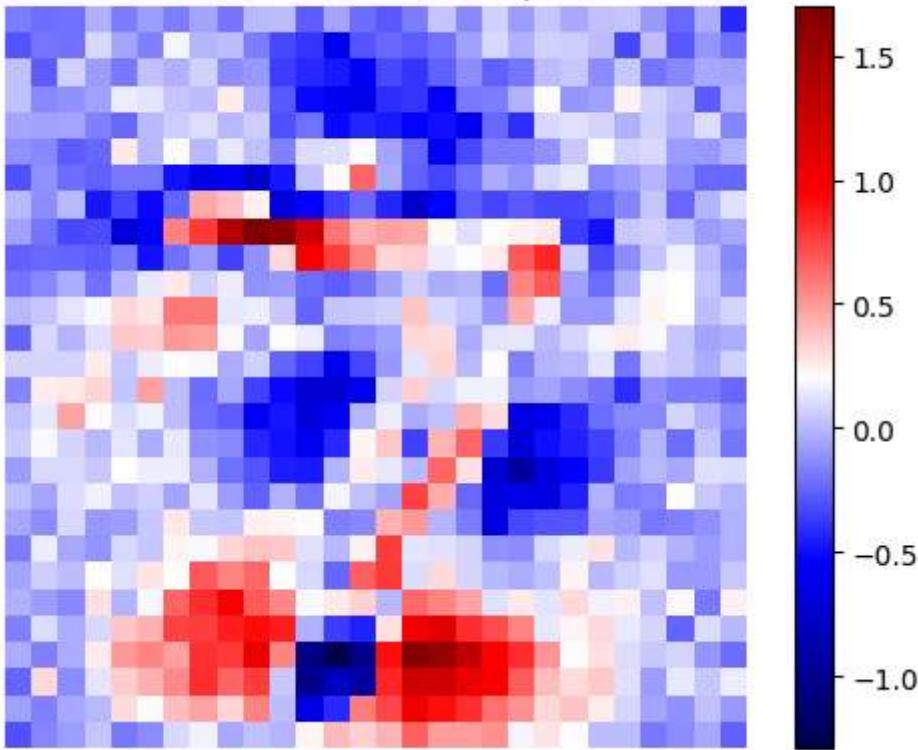
digit_hm = render.save_image([digit, hm], f'../heatmap_{i}.png')
data_io.write(R, '../heatmap.npy')

#display the image as written to file
plt.imshow(digit_hm, interpolation = 'none')
plt.axis('off')
plt.show()

#note that modules.Sequential allows for batch processing inputs
if True:
    N = 256
    t_start = time.time()
    x = X[:N,...]
    y = nn.forward(x)
    R = nn.lrp(y)
    data_io.write(R, '../Rbatch.npy')
    print('Computation of {} heatmaps using {} in {:.3f}s'.format(N, np.__name__, time.tim
```

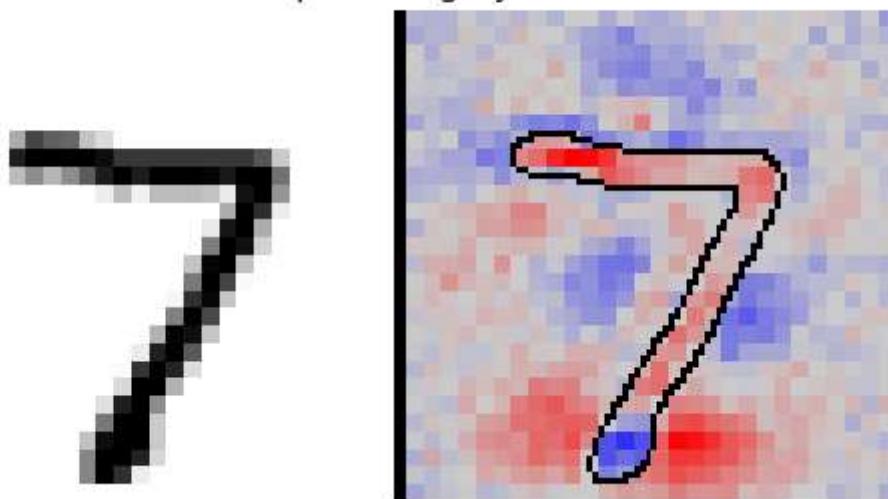
```
→ loading pickled model from ../models/MNIST/long-tanh.nn
removing softmax output mapping
loading np-formatted data from ../data/MNIST/test_images.npy
loading np-formatted data from ../data/MNIST/test_labels.npy
model test accuracy is: 0.9916
True Class:    7
Predicted Class: 7
```

Relevance Heatmap



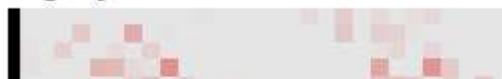
Applying custom heatmap: gray-red
saving image to ../heatmap_0_gray-red.png

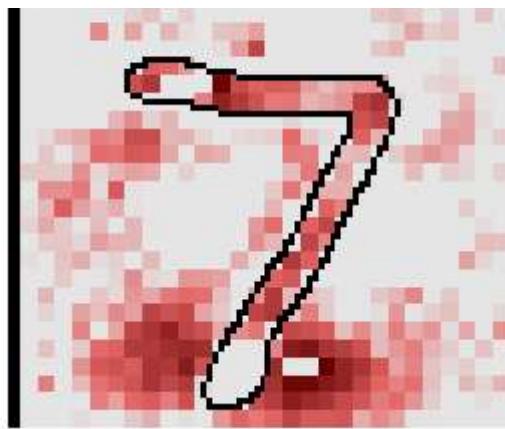
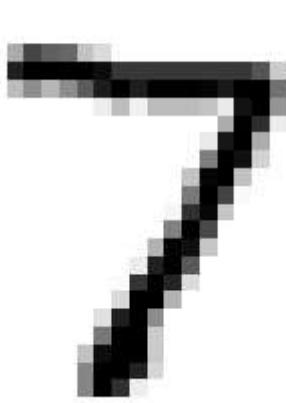
Sample #0 - gray-red



Applying custom heatmap: gray-red2
saving image to ../heatmap_0_gray-red2.png

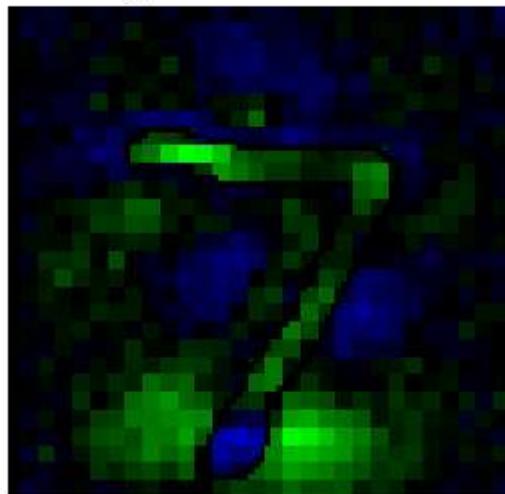
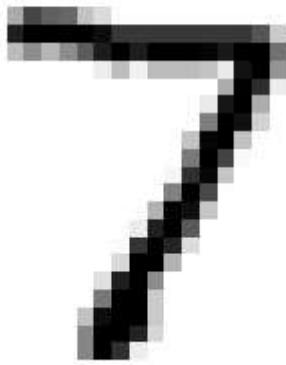
Sample #0 - gray-red2





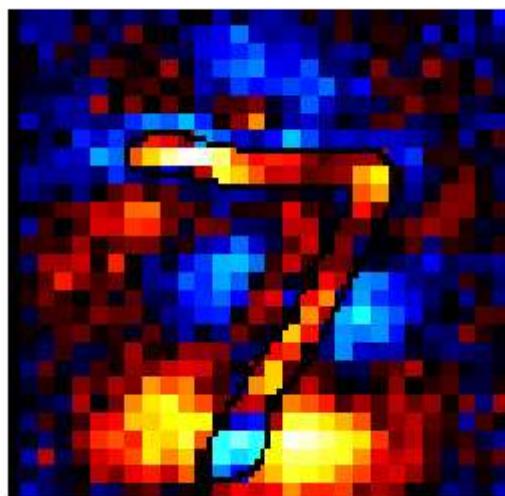
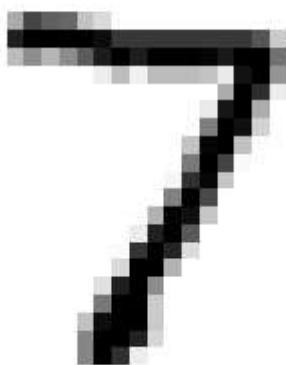
Applying custom heatmap: black-green
saving image to .../heatmap_0_black-green.png

Sample #0 - black-green



Applying custom heatmap: black-firered
saving image to .../heatmap_0_black-firered.png

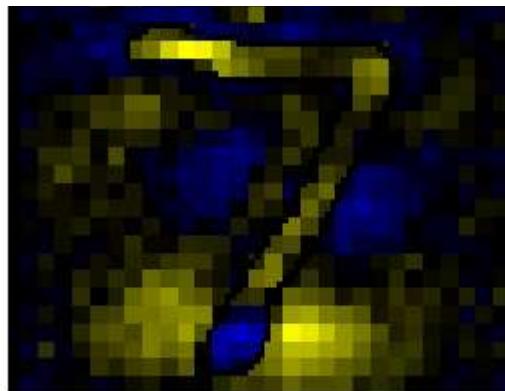
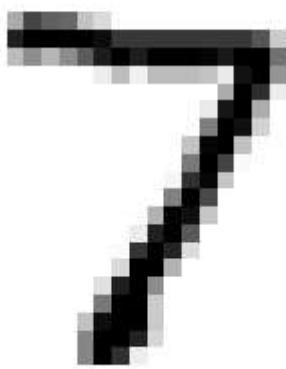
Sample #0 - black-firered



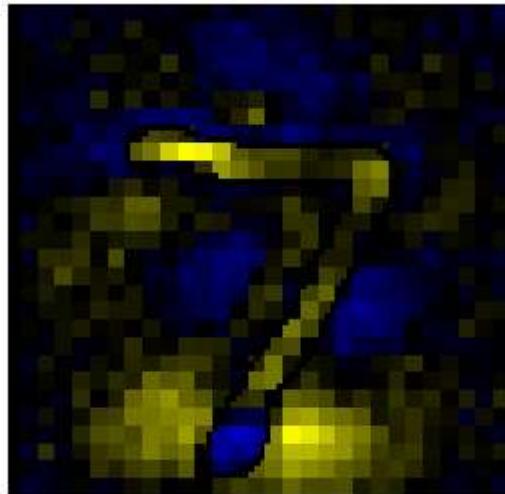
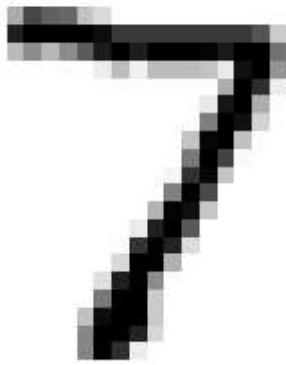
Applying custom heatmap: blue-black-yellow
saving image to .../heatmap_0_blue-black-yellow.png

Sample #0 - blue-black-yellow

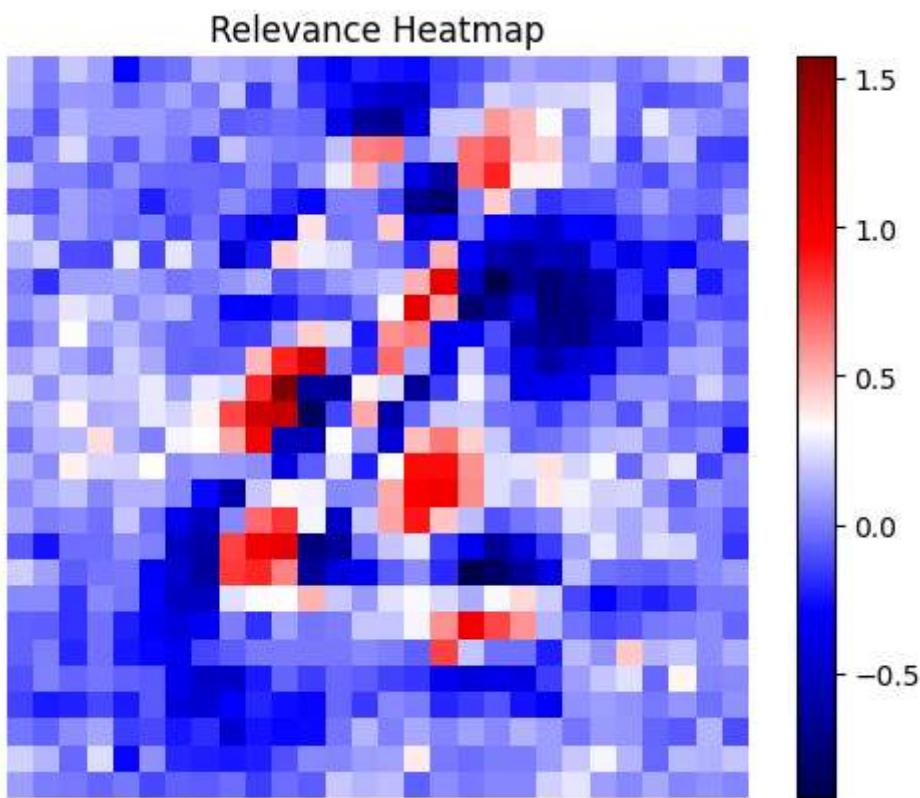




```
saving image to ../heatmap_0.png  
writing data in npy-format to ../heatmap.npy
```



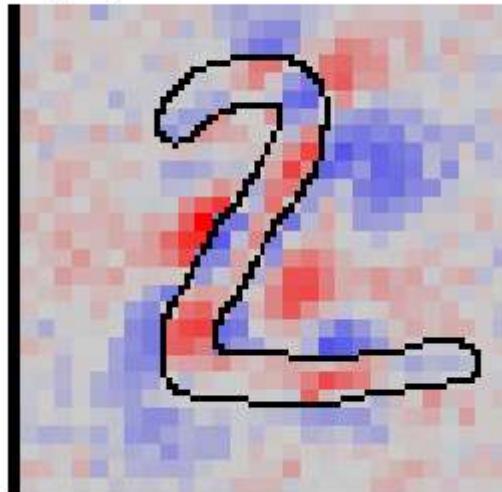
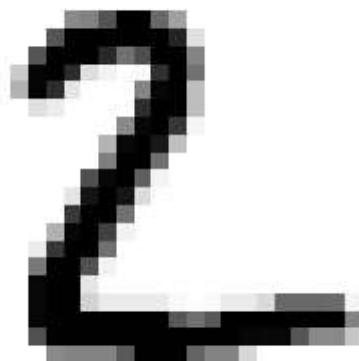
True Class: 2
Predicted Class: 2



Applying custom heatmap: gray-red

saving image to ../heatmap_1_gray-red.png

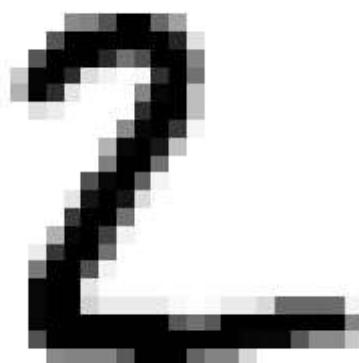
Sample #1 - gray-red



Applying custom heatmap: gray-red2

saving image to ../heatmap_1_gray-red2.png

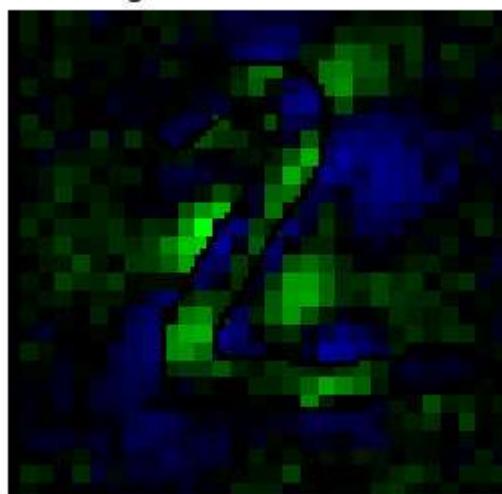
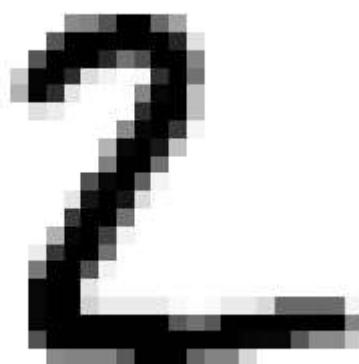
Sample #1 - gray-red2



Applying custom heatmap: black-green

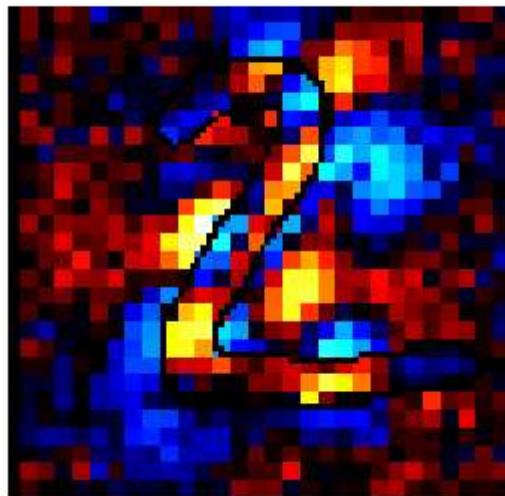
saving image to ../heatmap_1_black-green.png

Sample #1 - black-green

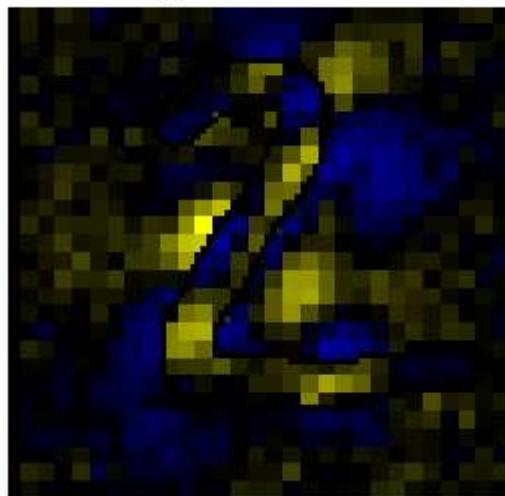


Applying custom heatmap: black-firered

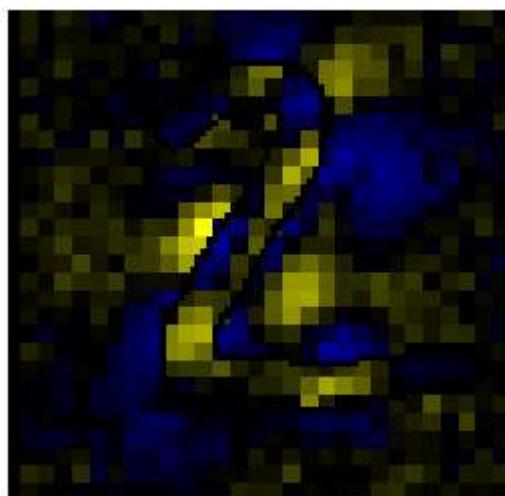
saving image to ../heatmap_1_black-firered.png

Sample #1 - black-firered

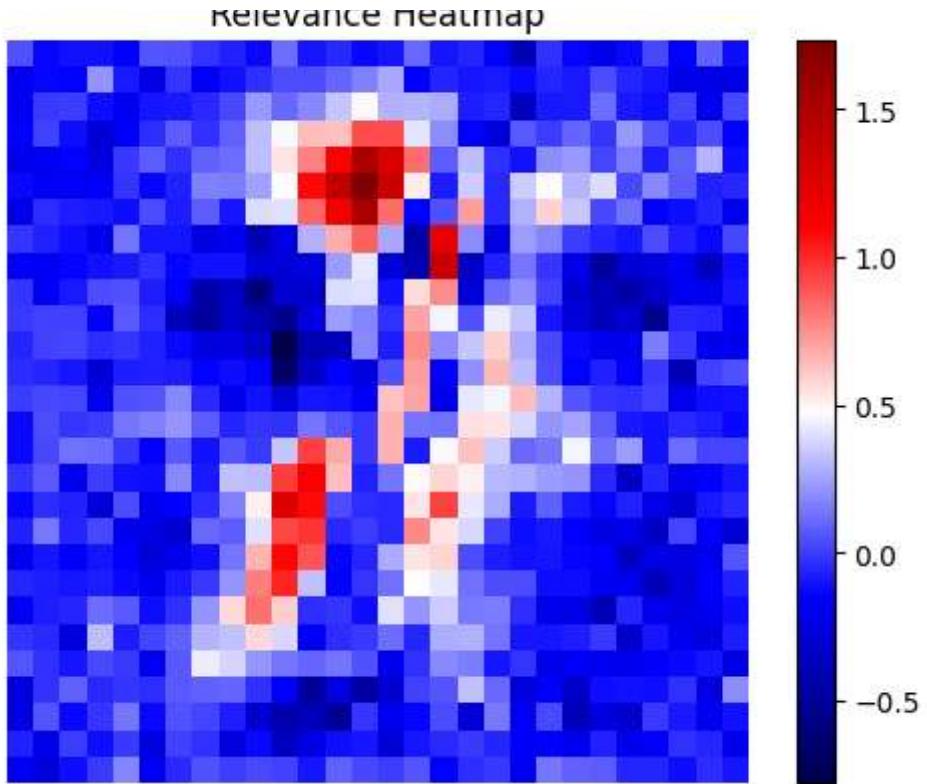
Applying custom heatmap: blue-black-yellow
saving image to ../heatmap_1_blue-black-yellow.png

Sample #1 - blue-black-yellow

saving image to ../heatmap_1.png
writing data in npy-format to ../heatmap.npy

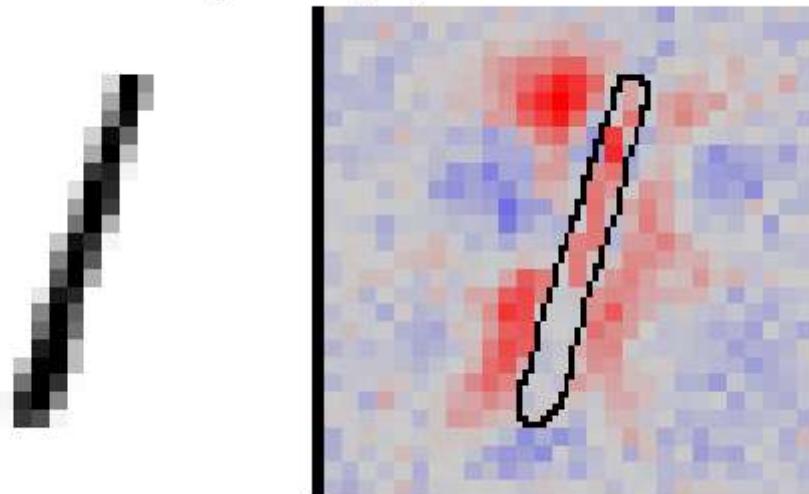


True Class: 1
Predicted Class: 1



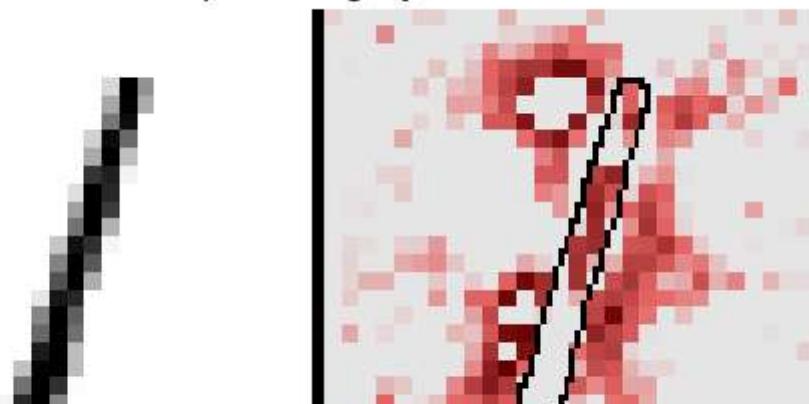
Applying custom heatmap: gray-red
saving image to .../heatmap_2_gray-red.png

Sample #2 - gray-red



Applying custom heatmap: gray-red2
saving image to .../heatmap_2_gray-red2.png

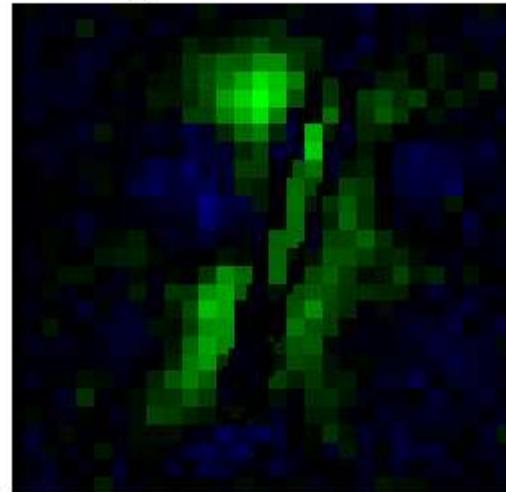
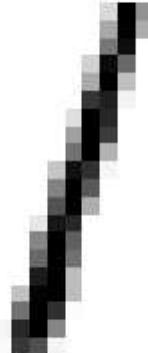
Sample #2 - gray-red2





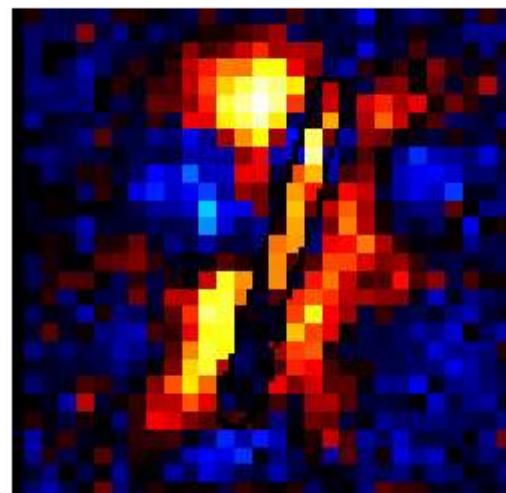
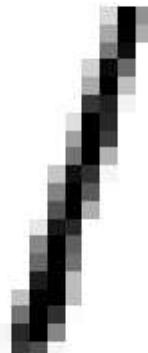
Applying custom heatmap: black-green
saving image to .../heatmap_2_black-green.png

Sample #2 - black-green



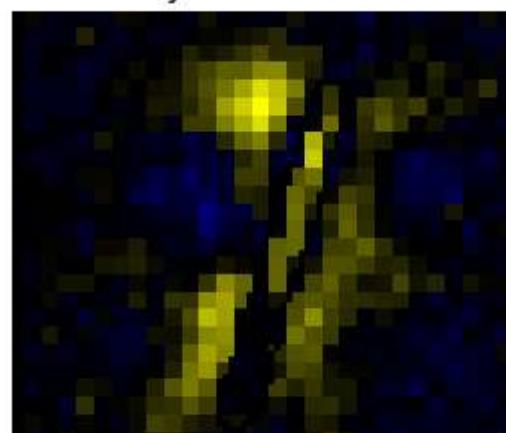
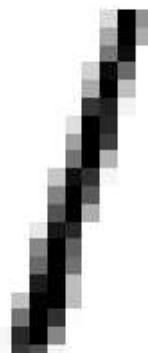
Applying custom heatmap: black-firered
saving image to .../heatmap_2_black-firered.png

Sample #2 - black-firered



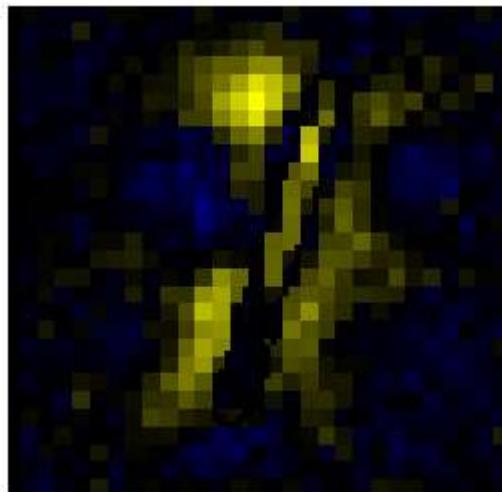
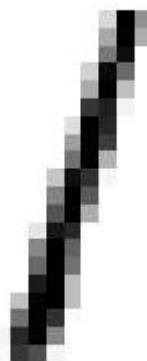
Applying custom heatmap: blue-black-yellow
saving image to .../heatmap_2_blue-black-yellow.png

Sample #2 - blue-black-yellow



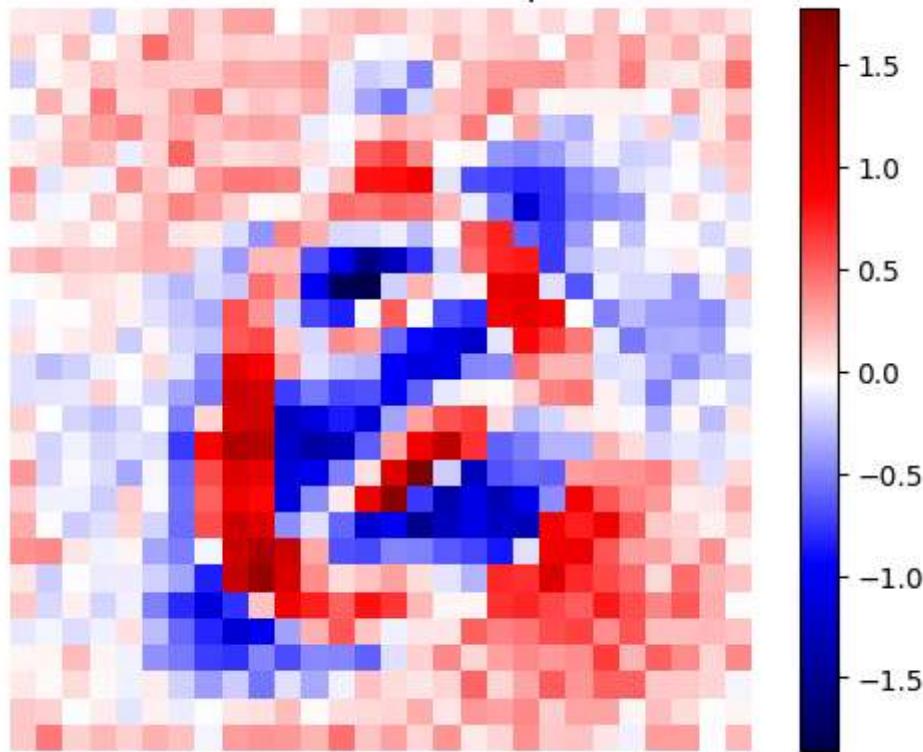


saving image to ../heatmap_2.png
writing data in npy-format to ../heatmap.npy



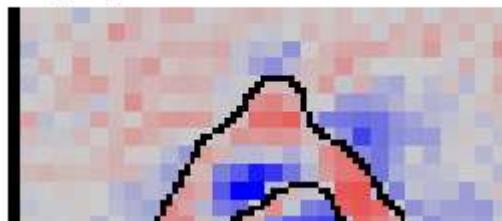
True Class: 0
Predicted Class: 0

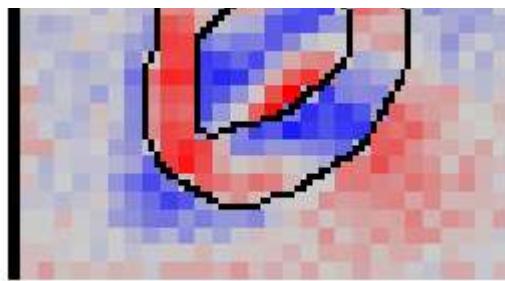
Relevance Heatmap



Applying custom heatmap: gray-red
saving image to ../heatmap_3_gray-red.png

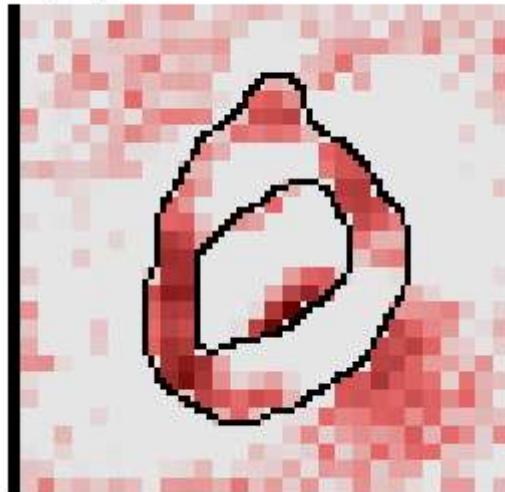
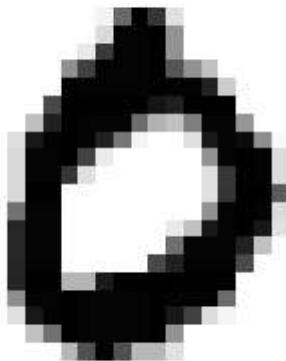
Sample #3 - gray-red





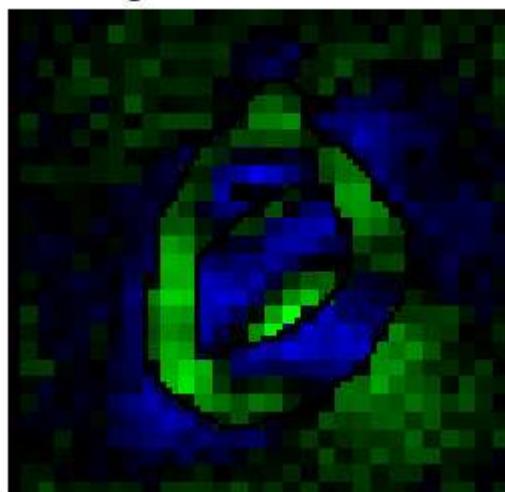
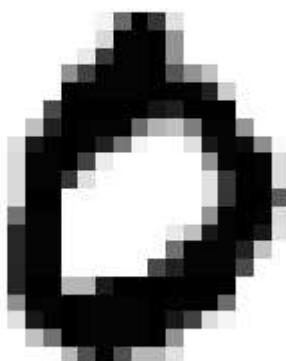
Applying custom heatmap: gray-red2
saving image to ../heatmap_3_gray-red2.png

Sample #3 - gray-red2



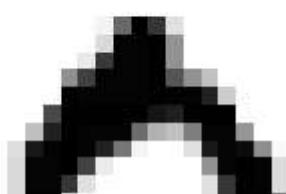
Applying custom heatmap: black-green
saving image to ../heatmap_3_black-green.png

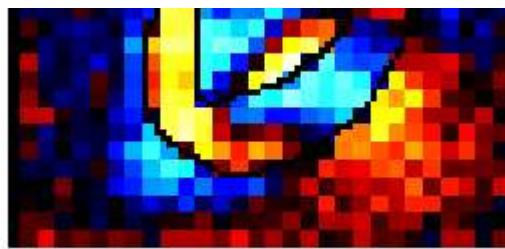
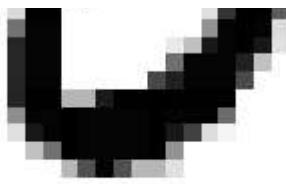
Sample #3 - black-green



Applying custom heatmap: black-firered
saving image to ../heatmap_3_black-firered.png

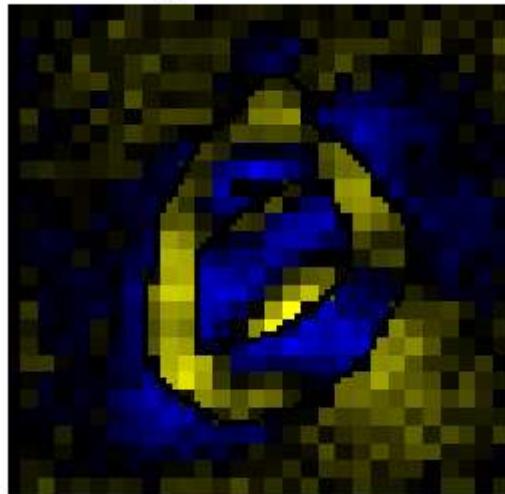
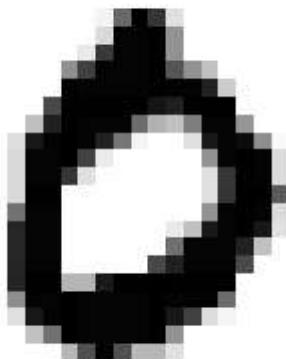
Sample #3 - black-firered



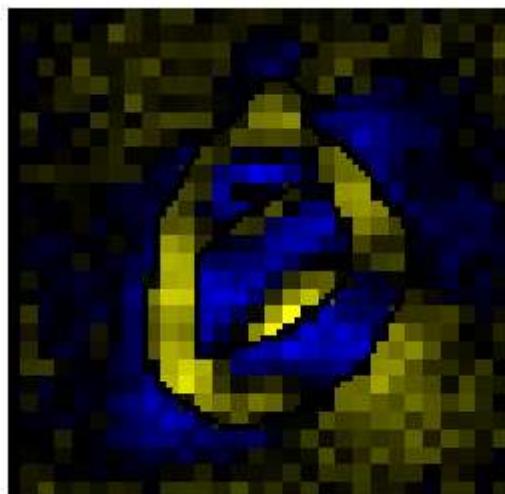
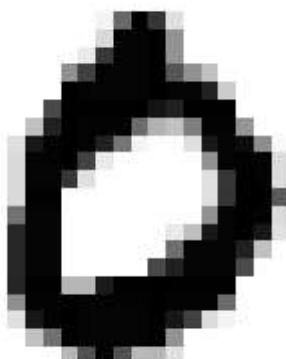


Applying custom heatmap: blue-black-yellow
saving image to ../heatmap_3_blue-black-yellow.png

Sample #3 - blue-black-yellow

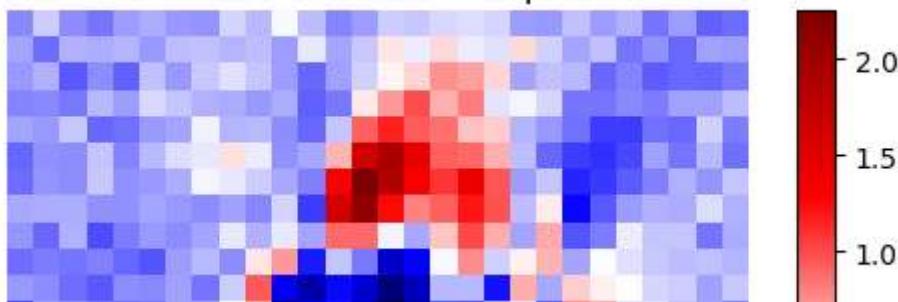


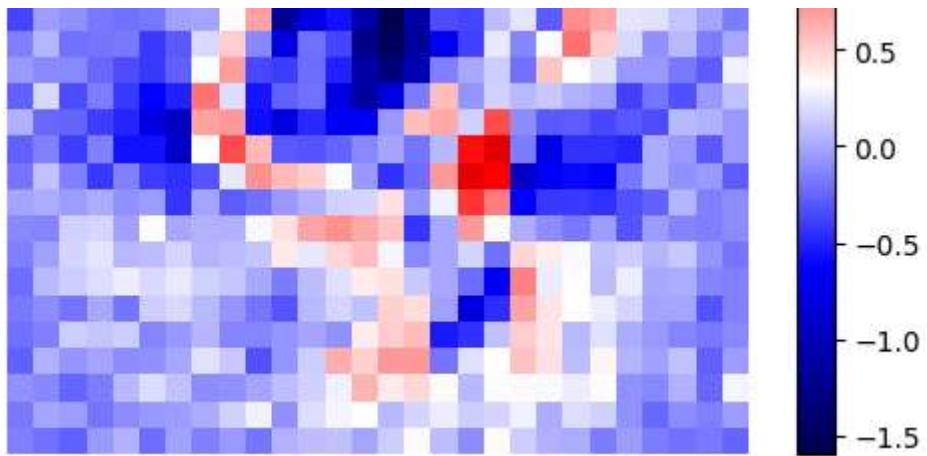
saving image to ../heatmap_3.png
writing data in npy-format to ../heatmap.npy



True Class: 4
Predicted Class: 4

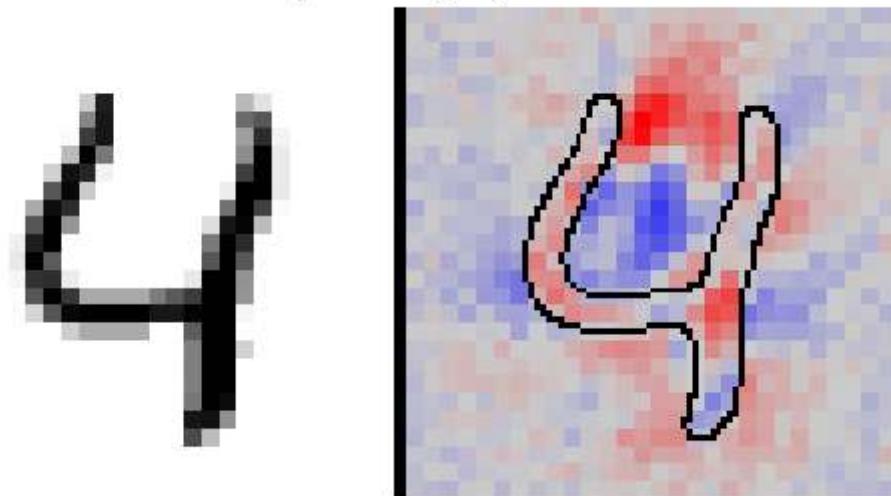
Relevance Heatmap





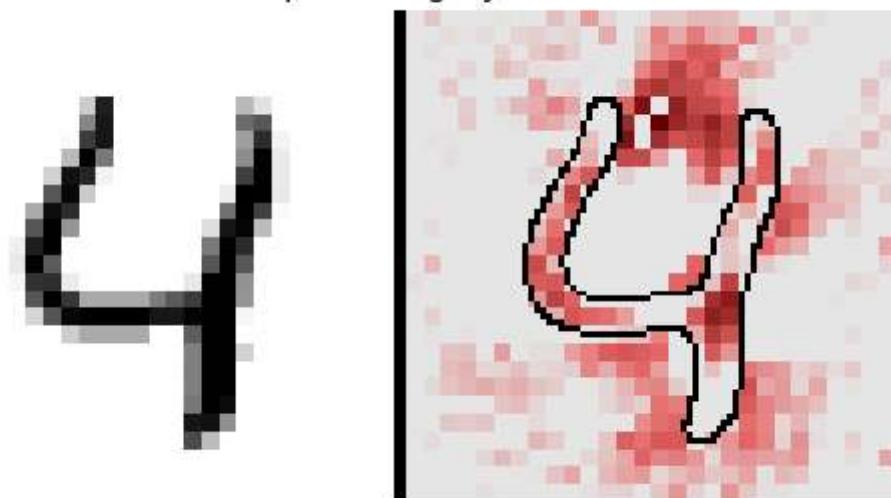
Applying custom heatmap: gray-red
saving image to .../heatmap_4_gray-red.png

Sample #4 - gray-red



Applying custom heatmap: gray-red2
saving image to .../heatmap_4_gray-red2.png

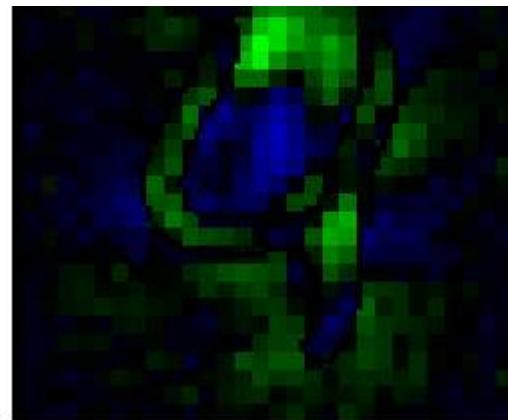
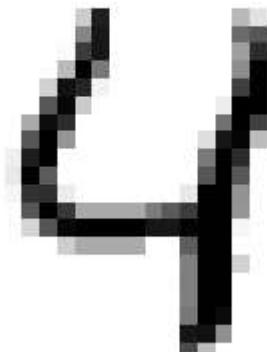
Sample #4 - gray-red2



Applying custom heatmap: black-green
saving image to .../heatmap_4_black-green.png

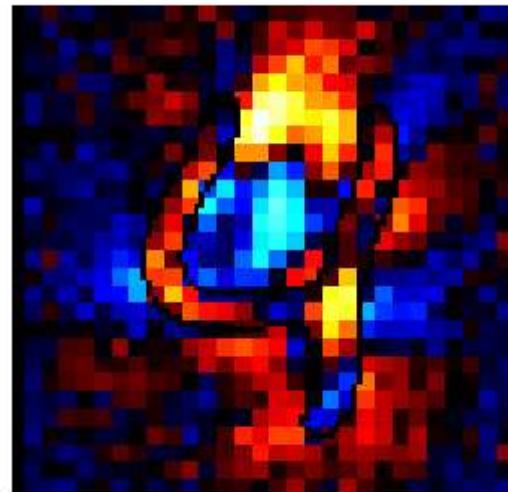
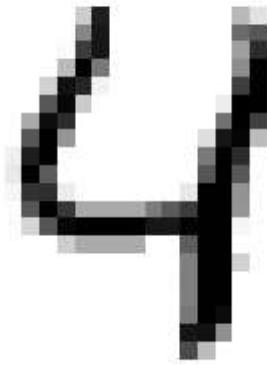
Sample #4 - black-green





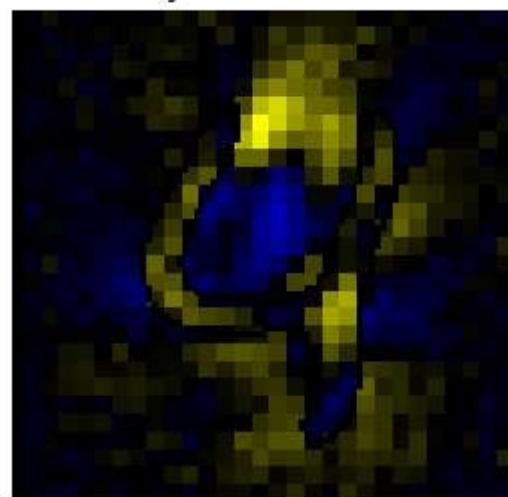
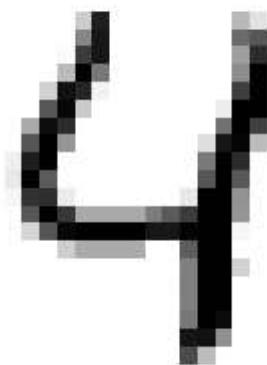
Applying custom heatmap: black-firered
saving image to .../heatmap_4_black-firered.png

Sample #4 - black-firered



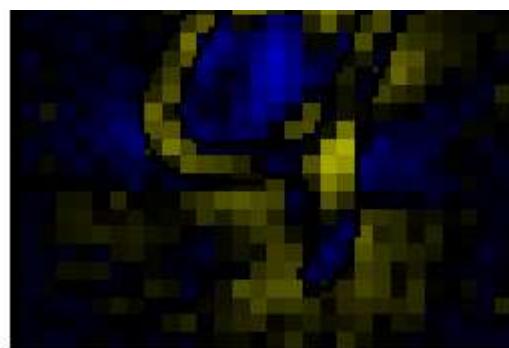
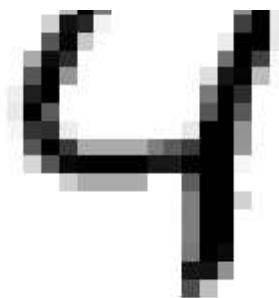
Applying custom heatmap: blue-black-yellow
saving image to .../heatmap_4_blue-black-yellow.png

Sample #4 - blue-black-yellow



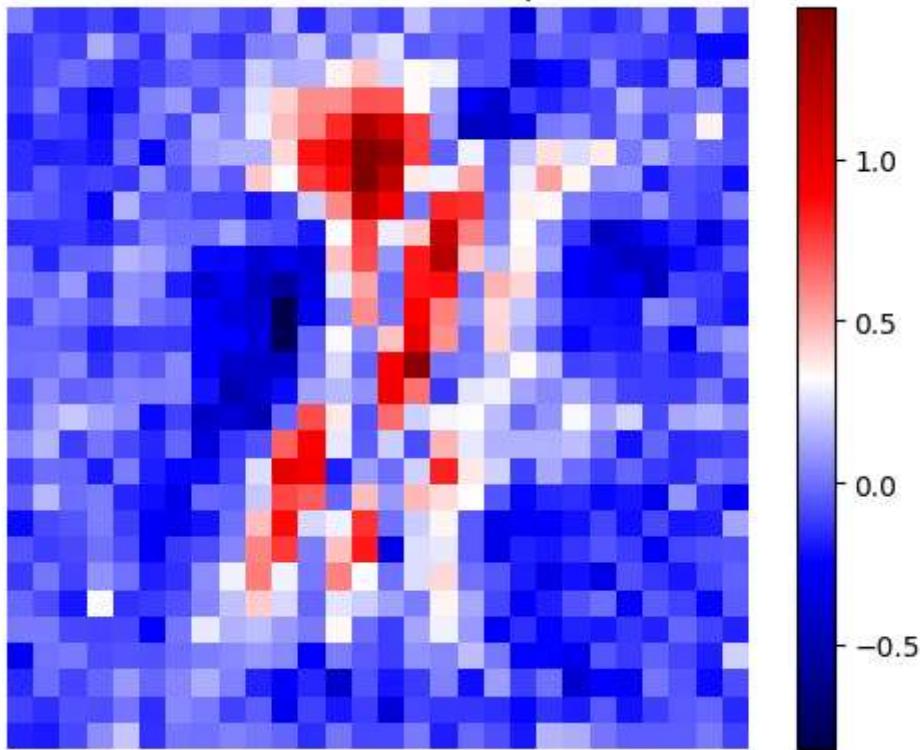
saving image to .../heatmap_4.png
writing data in npy-format to .../heatmap.npy





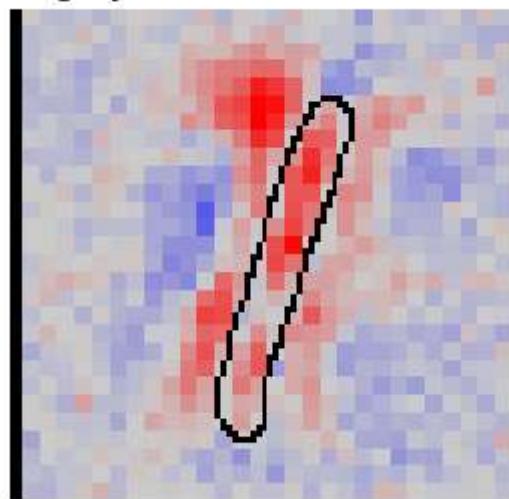
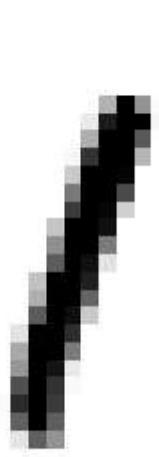
True Class: 1
Predicted Class: 1

Relevance Heatmap



Applying custom heatmap: gray-red
saving image to ../heatmap_5_gray-red.png

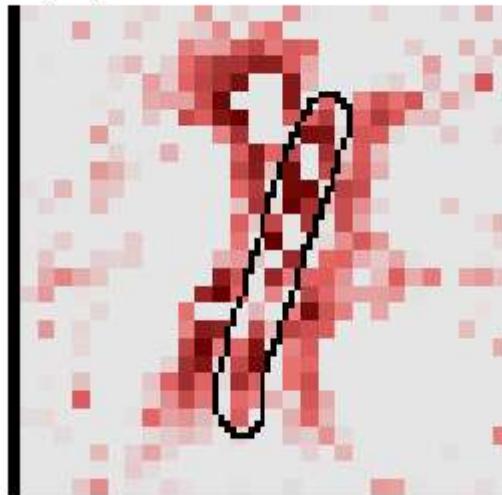
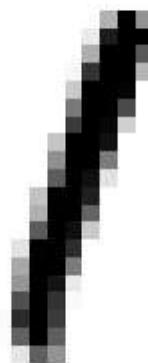
Sample #5 - gray-red



Applying custom heatmap: gray-red2

saving image to ../heatmap_5_gray-red2.png

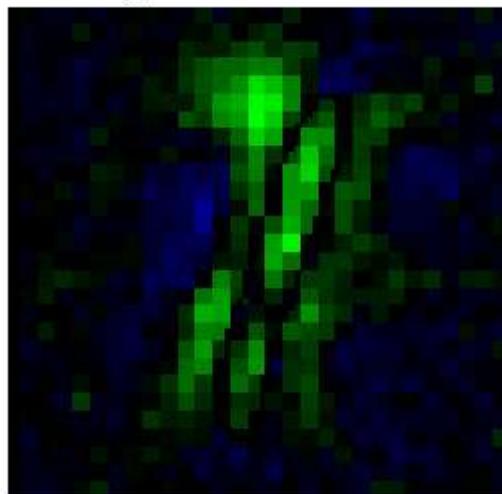
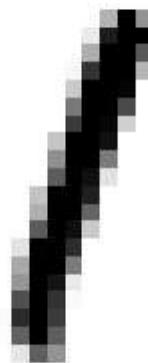
Sample #5 - gray-red2



Applying custom heatmap: black-green

saving image to ../heatmap_5_black-green.png

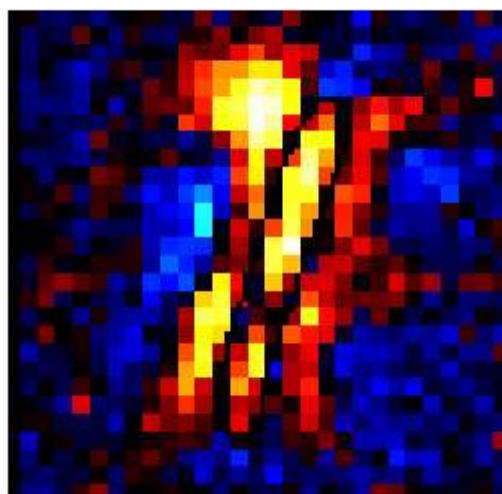
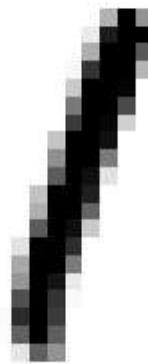
Sample #5 - black-green



Applying custom heatmap: black-firered

saving image to ../heatmap_5_black-firered.png

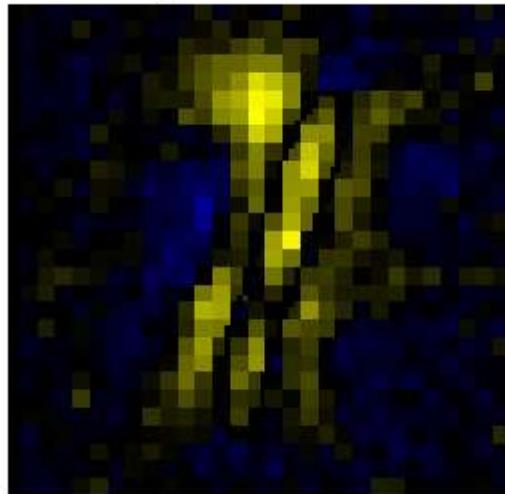
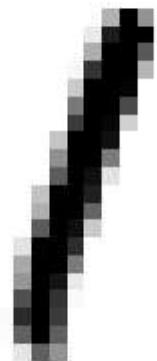
Sample #5 - black-firered



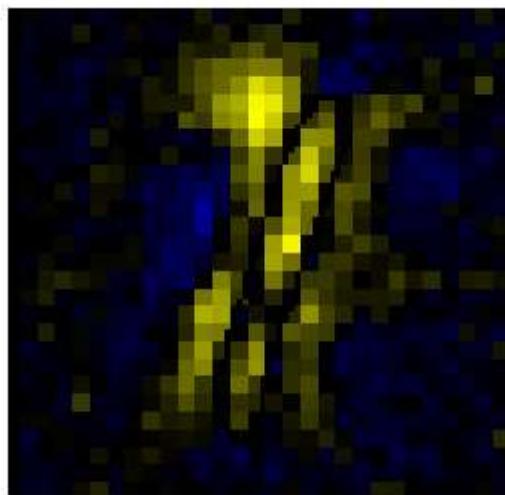
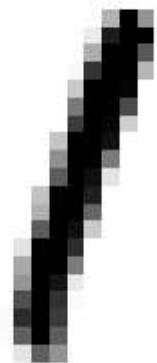
Applying custom heatmap: blue-black-yellow

saving image to ../heatmap_5_blue-black-yellow.png

Sample #5 - blue-black-yellow

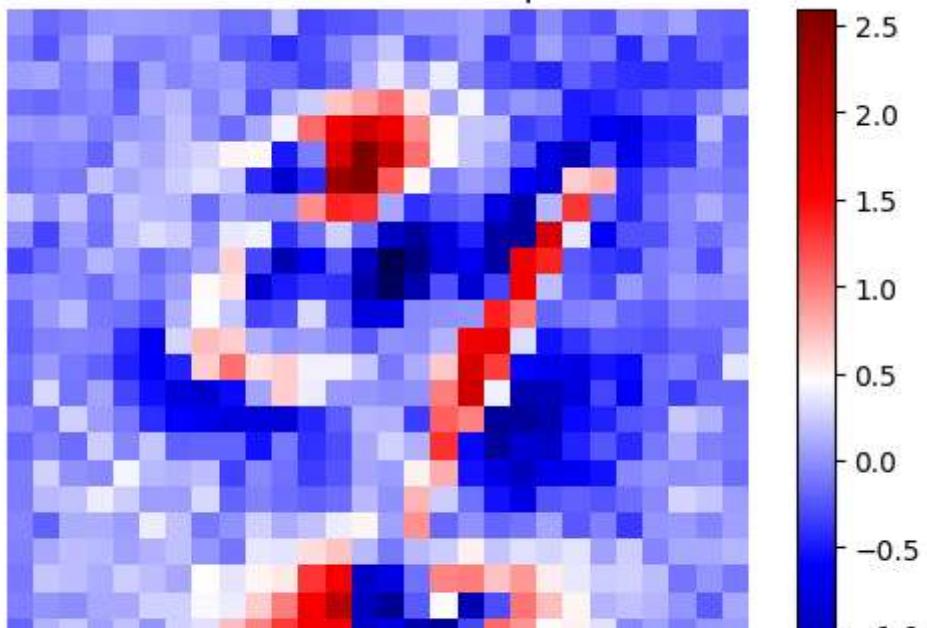


saving image to .../heatmap_5.png
writing data in npy-format to .../heatmap.npy



True Class: 4
Predicted Class: 4

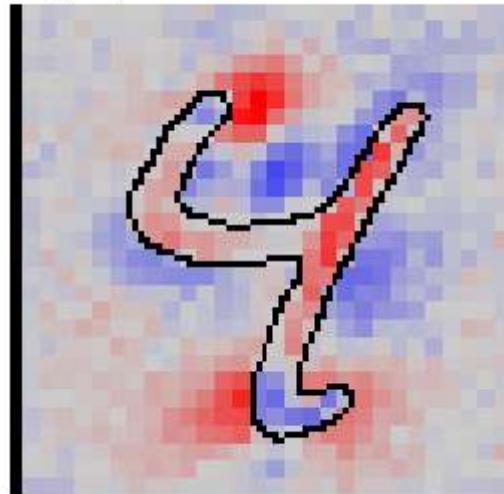
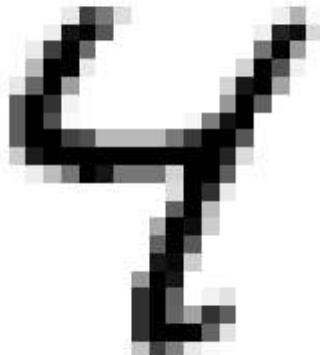
Relevance Heatmap





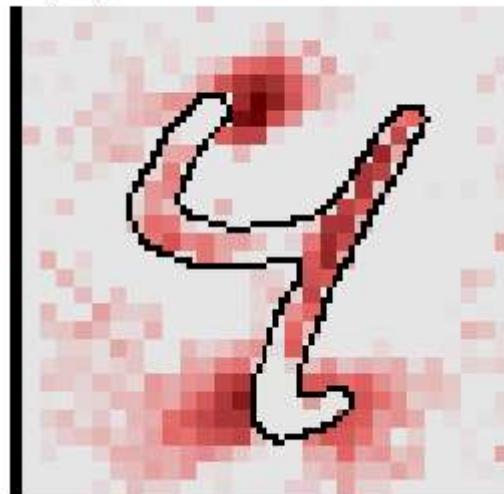
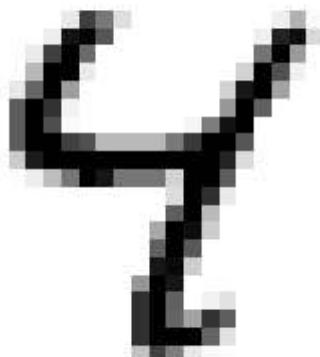
Applying custom heatmap: gray-red
saving image to ../heatmap_6_gray-red.png

Sample #6 - gray-red



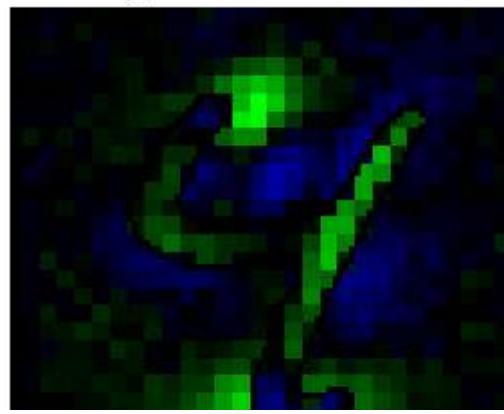
Applying custom heatmap: gray-red2
saving image to ../heatmap_6_gray-red2.png

Sample #6 - gray-red2



Applying custom heatmap: black-green
saving image to ../heatmap_6_black-green.png

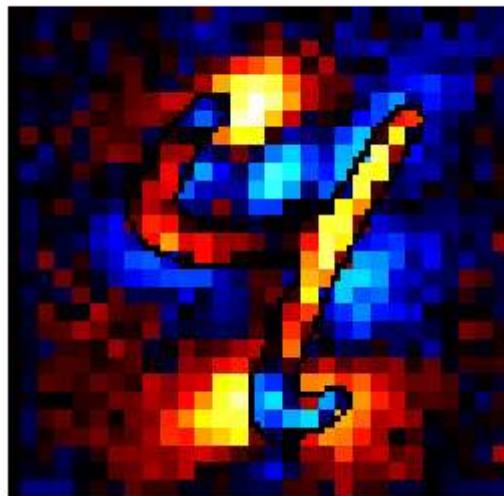
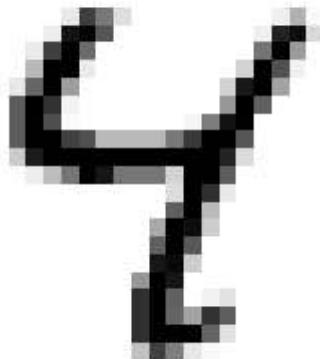
Sample #6 - black-green





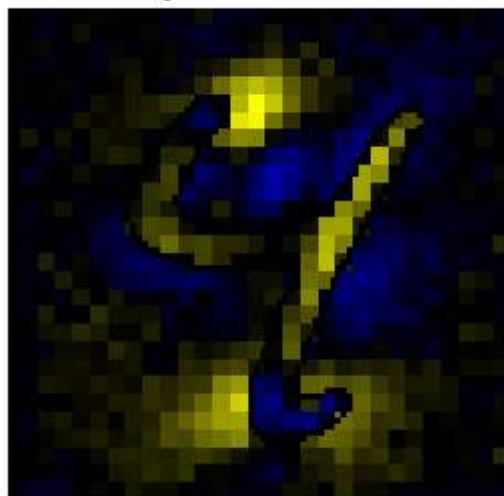
Applying custom heatmap: black-firered
saving image to .../heatmap_6_black-firered.png

Sample #6 - black-firered

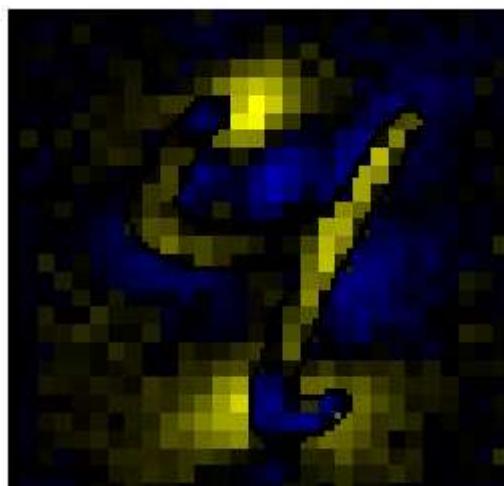
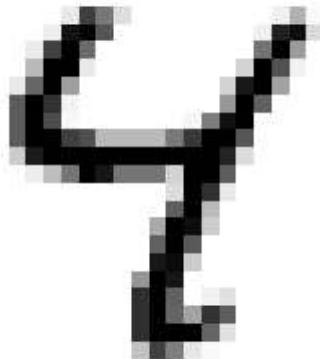


Applying custom heatmap: blue-black-yellow
saving image to .../heatmap_6_blue-black-yellow.png

Sample #6 - blue-black-yellow

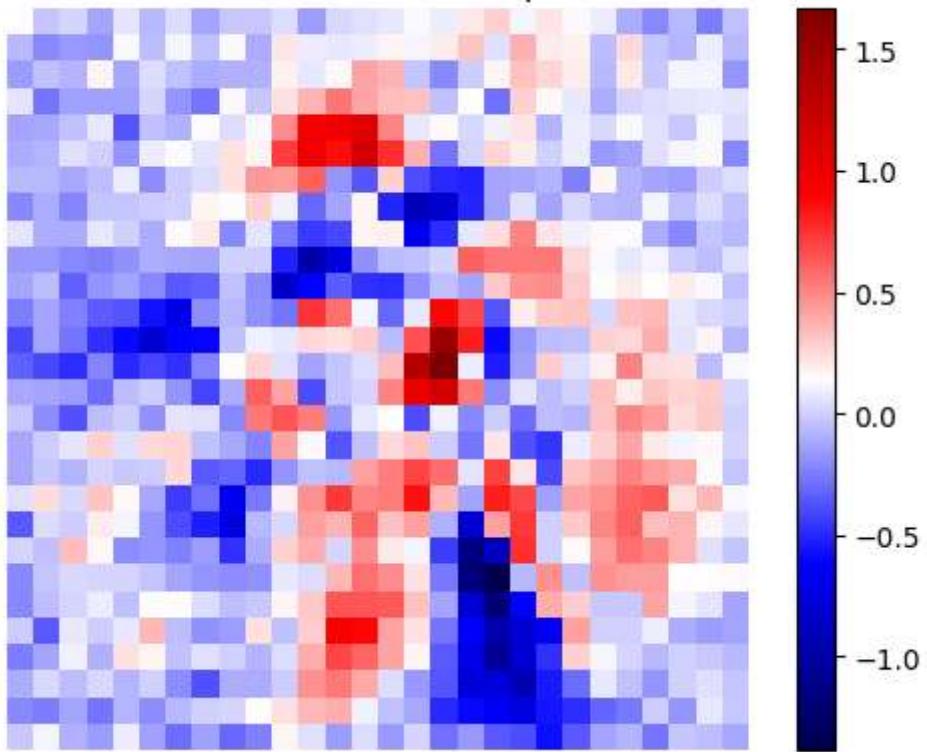


saving image to .../heatmap_6.png
writing data in npy-format to .../heatmap.npy



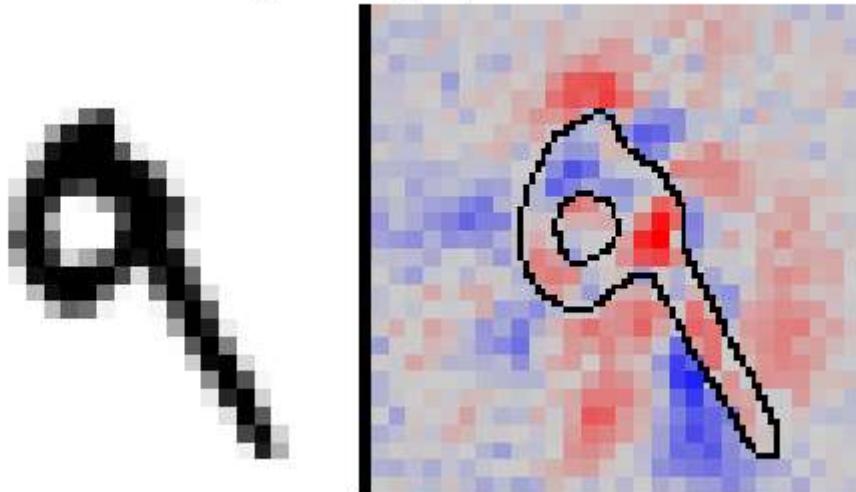
True Class: 9
Predicted Class: 9

Relevance Heatmap



Applying custom heatmap: gray-red
saving image to ../heatmap_7_gray-red.png

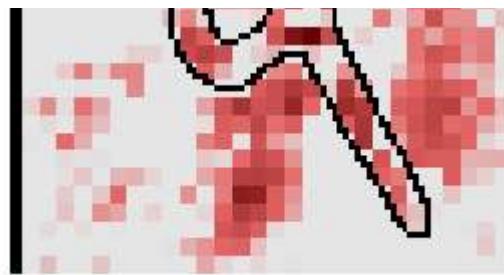
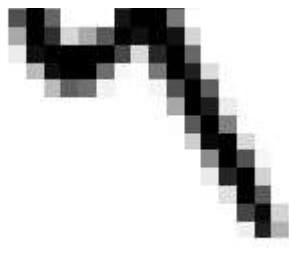
Sample #7 - gray-red



Applying custom heatmap: gray-red2
saving image to ../heatmap_7_gray-red2.png

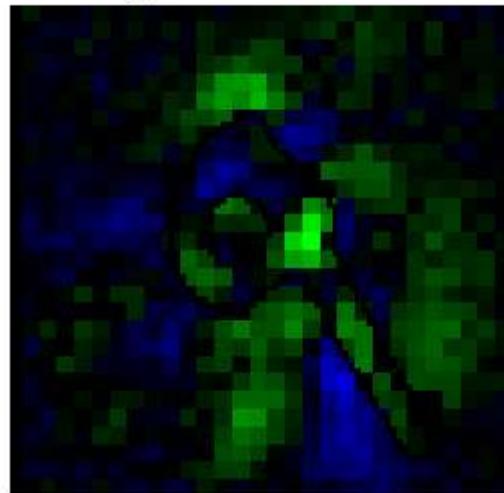
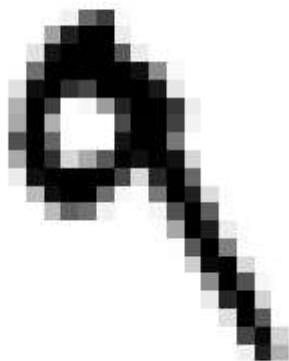
Sample #7 - gray-red2





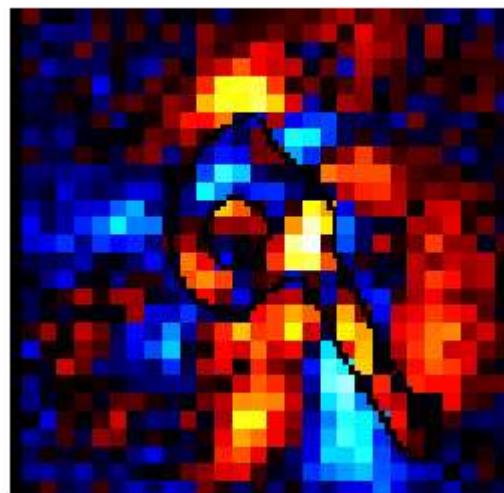
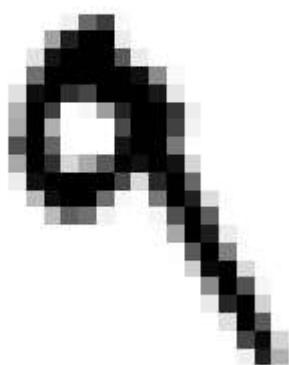
Applying custom heatmap: black-green
saving image to ../heatmap_7_black-green.png

Sample #7 - black-green



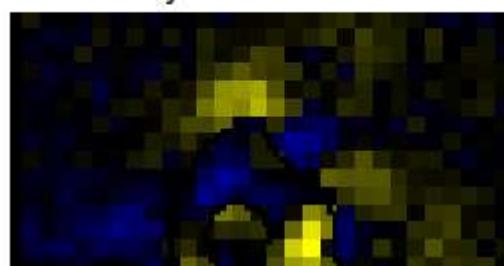
Applying custom heatmap: black-firered
saving image to ../heatmap_7_black-firered.png

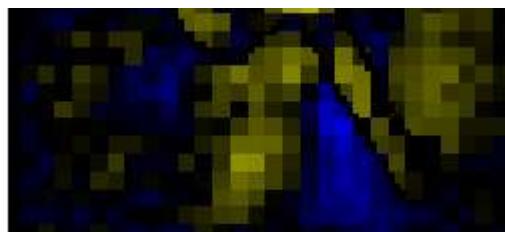
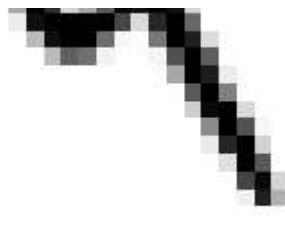
Sample #7 - black-firered



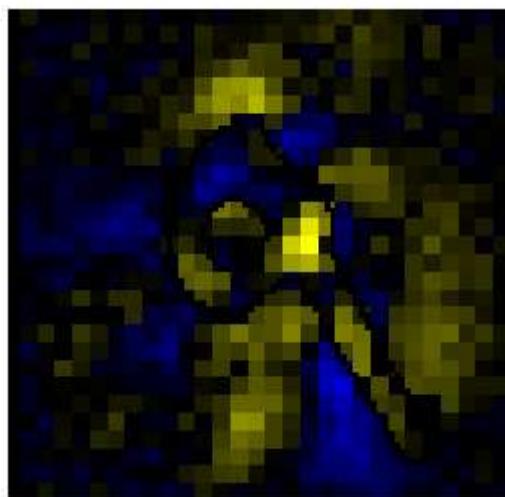
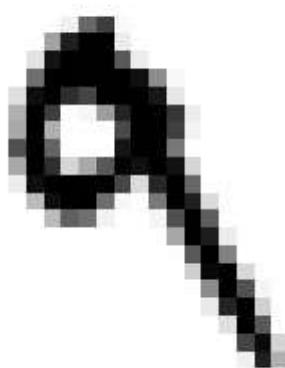
Applying custom heatmap: blue-black-yellow
saving image to ../heatmap_7_blue-black-yellow.png

Sample #7 - blue-black-yellow



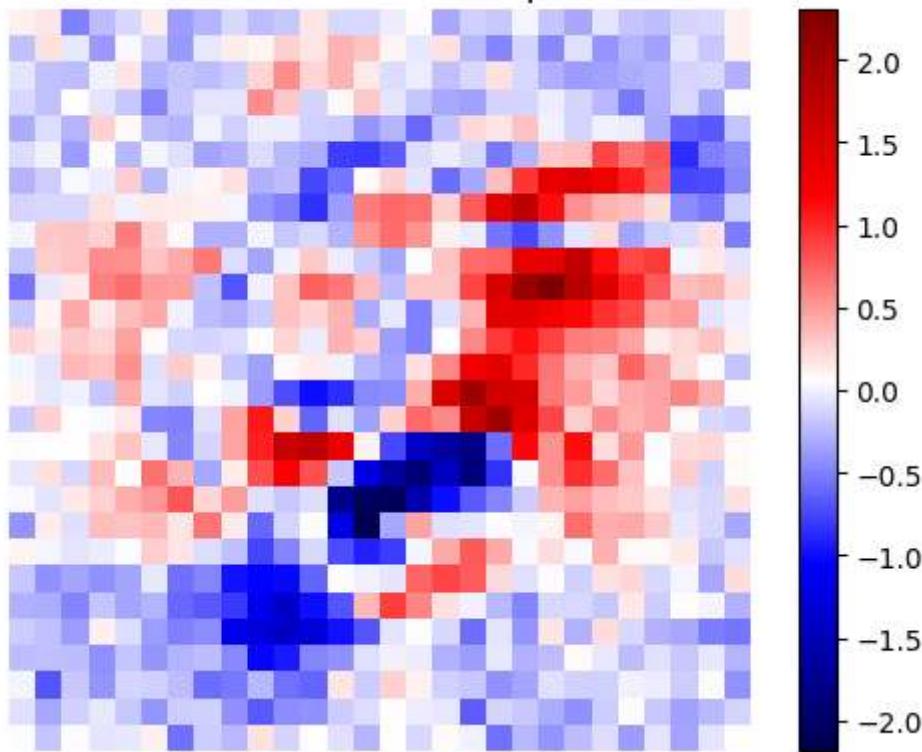


saving image to ../heatmap_7.png
writing data in npy-format to ../heatmap.npy



True Class: 5
Predicted Class: 5

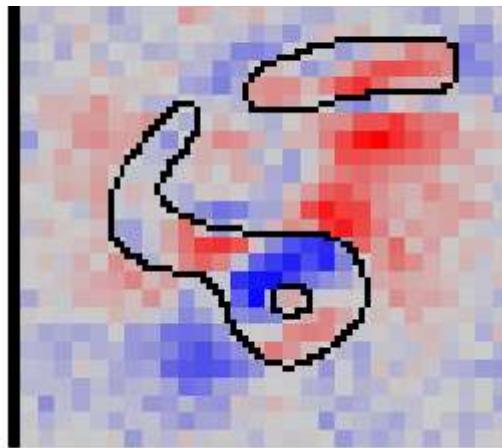
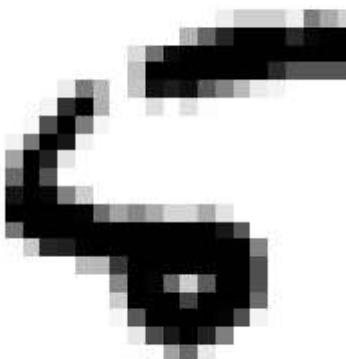
Relevance Heatmap



Applying custom heatmap: gray-red
saving image to ../heatmap_8_gray-red.png

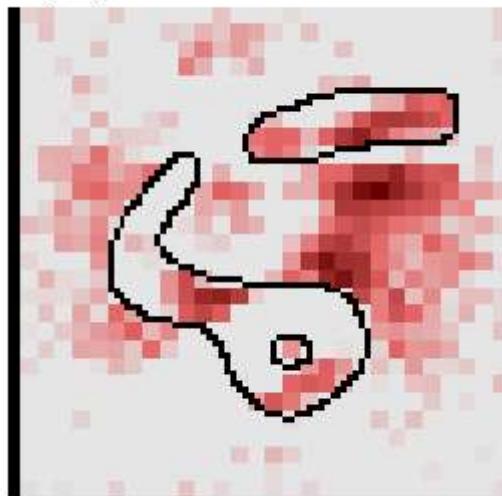
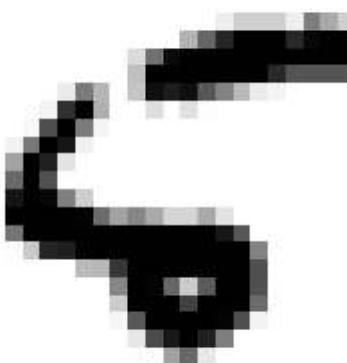
Sample #8 - gray-red





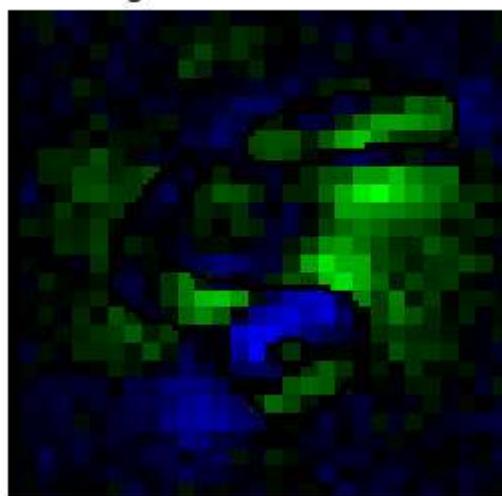
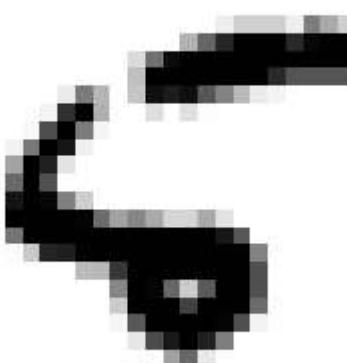
Applying custom heatmap: gray-red2
saving image to ../heatmap_8_gray-red2.png

Sample #8 - gray-red2



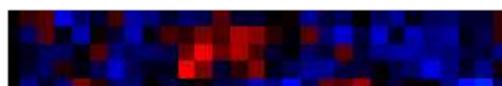
Applying custom heatmap: black-green
saving image to ../heatmap_8_black-green.png

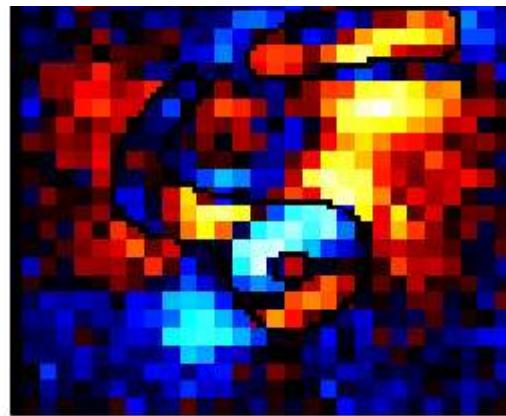
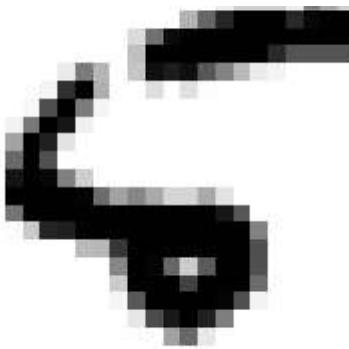
Sample #8 - black-green



Applying custom heatmap: black-firered
saving image to ../heatmap_8_black-firered.png

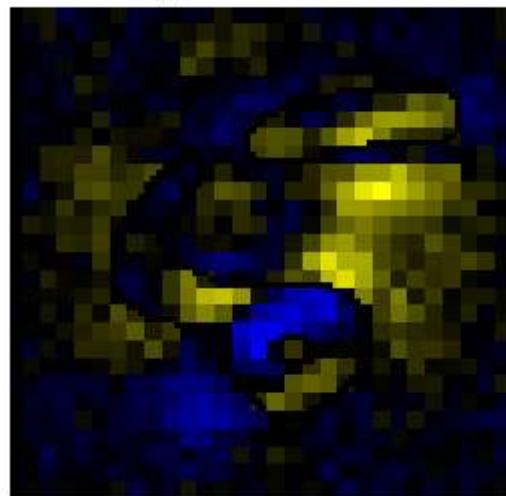
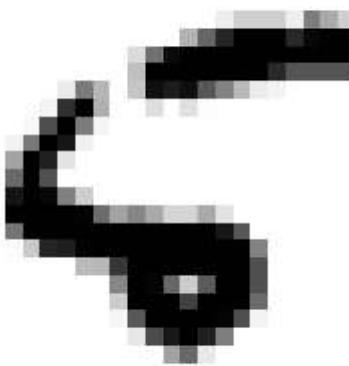
Sample #8 - black-firered



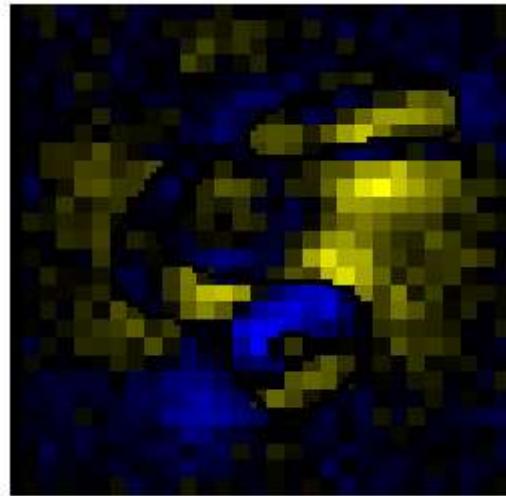


Applying custom heatmap: blue-black-yellow
saving image to .../heatmap_8_blue-black-yellow.png

Sample #8 - blue-black-yellow



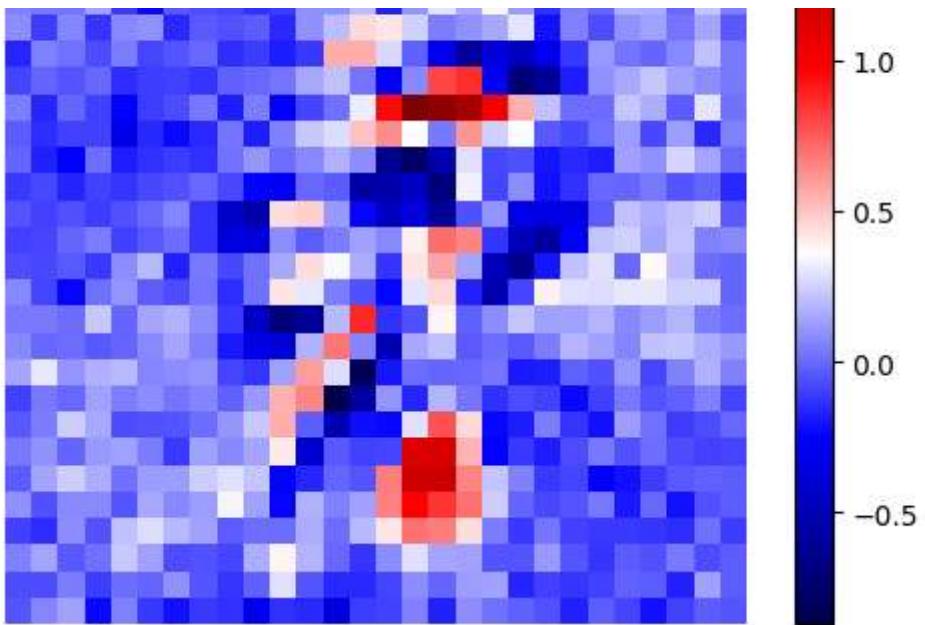
saving image to .../heatmap_8.png
writing data in npy-format to .../heatmap.npy



True Class: 9
Predicted Class: 9

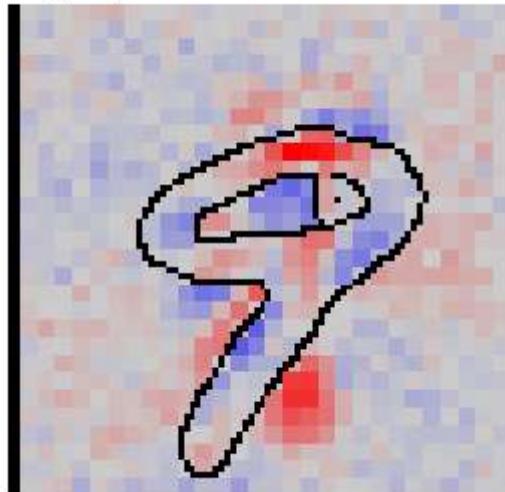
Relevance Heatmap





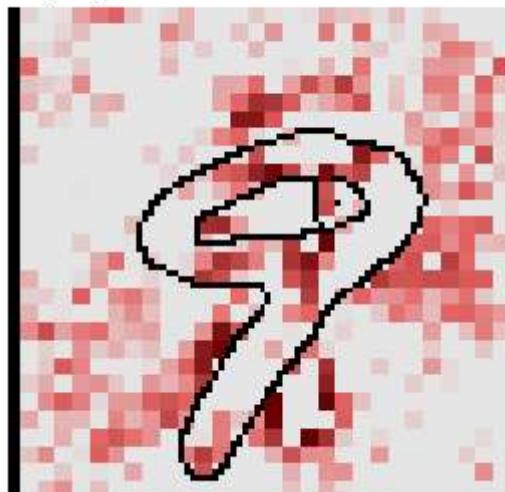
Applying custom heatmap: gray-red
saving image to ../heatmap_9_gray-red.png

Sample #9 - gray-red



Applying custom heatmap: gray-red2
saving image to ../heatmap_9_gray-red2.png

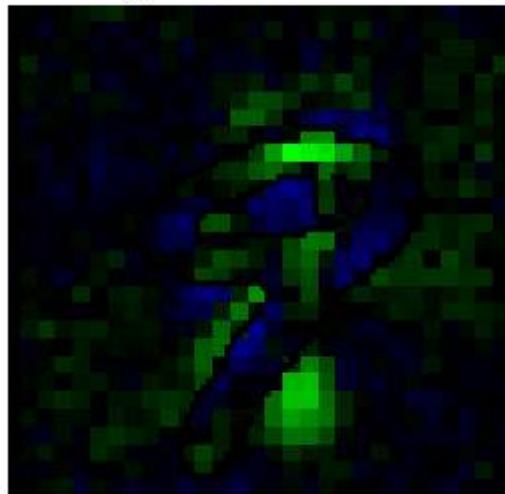
Sample #9 - gray-red2



Applying custom heatmap: black-green

saving image to ../heatmap_9_black-green.png

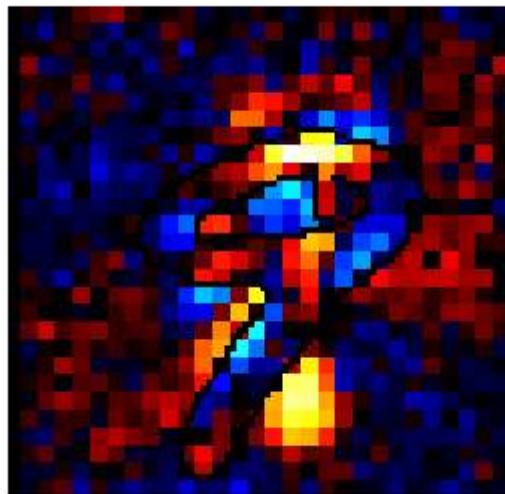
Sample #9 - black-green



Applying custom heatmap: black-firered

saving image to ../heatmap_9_black-firered.png

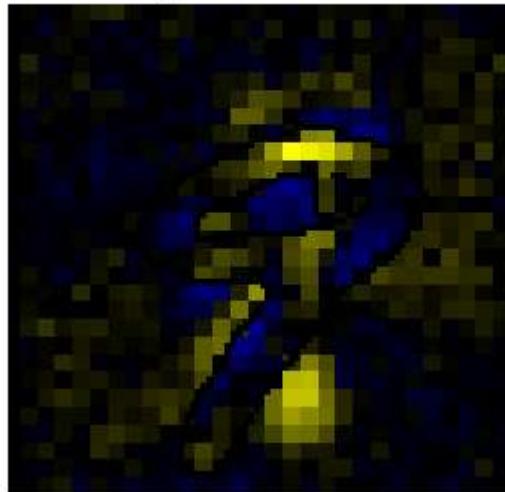
Sample #9 - black-firered



Applying custom heatmap: blue-black-yellow

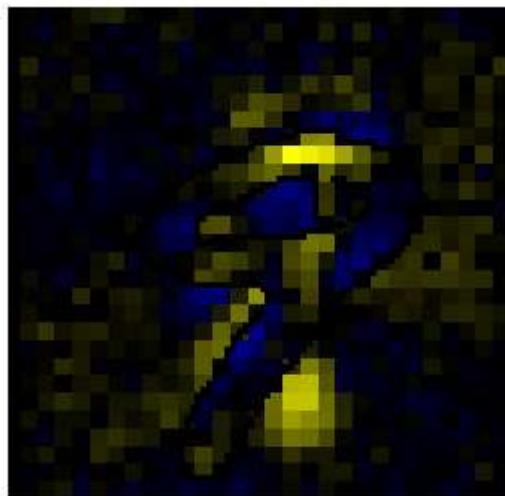
saving image to ../heatmap_9_blue-black-yellow.png

Sample #9 - blue-black-yellow



saving image to ../heatmap_9.png

writing data in npy-format to ../heatmap.npy



writing data in npy-format to ../Rbatch.npy
Computation of 256 heatmaps using cupy in 0.261s

```
import matplotlib.pyplot as plt
import matplotlib.image as mpimg
import os

# Path to the folder where heatmaps were saved
base_path = '../' # adjust if needed

num_samples = 10
cmap_names = [
    'gray-red', 'gray-red2', 'black-green', 'black-firered', 'blue-black-yellow'
] # Same as keys in render.custom_maps

# Set up a figure grid: rows = samples, columns = colormaps
fig, axes = plt.subplots(num_samples, len(cmap_names), figsize=(4 * len(cmap_names), 3 * num_samples))

for i in range(num_samples):
    for j, cmap in enumerate(cmap_names):
        img_path = os.path.join(base_path, f'heatmap_{i}_{cmap}.png')
        img = mpimg.imread(img_path)

        ax = axes[i, j] if num_samples > 1 else axes[j]
        ax.imshow(img)
        ax.axis('off')
        if i == 0:
            ax.set_title(cmap)

plt.tight_layout()
plt.show()
```



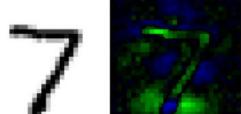
gray-red



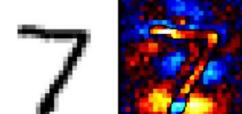
gray-red2



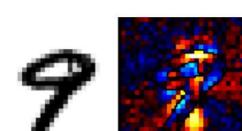
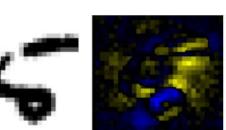
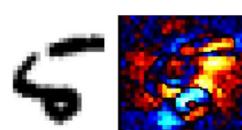
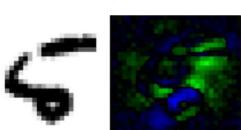
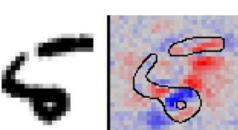
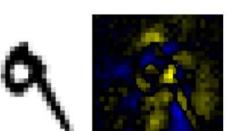
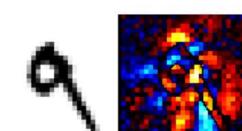
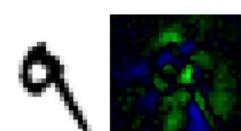
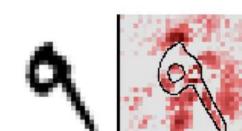
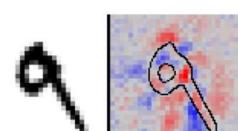
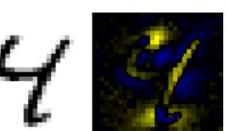
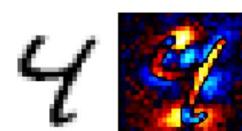
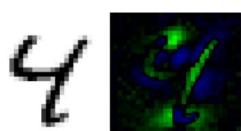
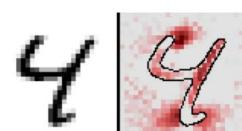
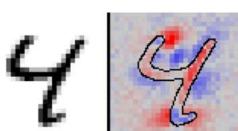
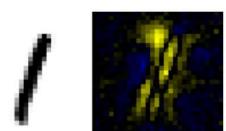
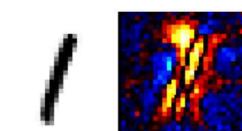
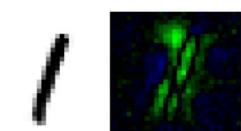
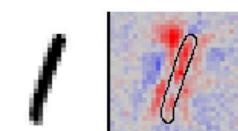
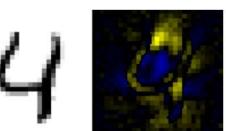
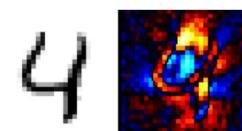
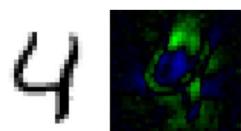
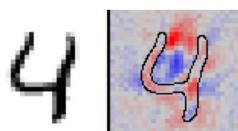
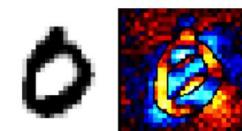
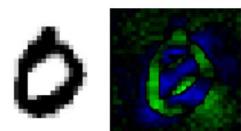
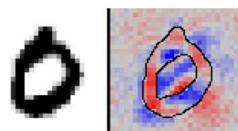
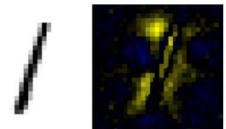
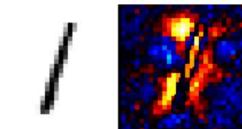
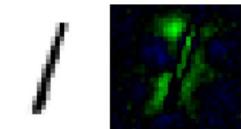
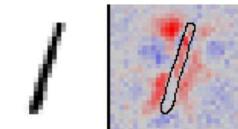
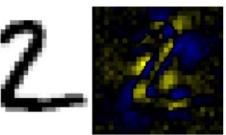
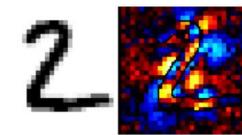
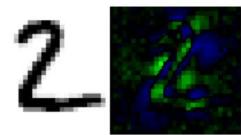
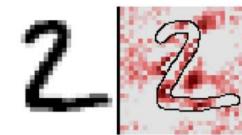
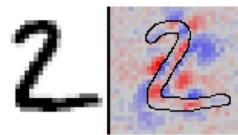
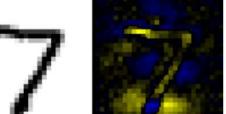
black-green



black-firered



blue-black-yellow



❖ Timing MNIST

```
...  
@author: Sebastian Lapuschkin  
@maintainer: Sebastian Lapuschkin  
@contact: sebastian.lapuschkin@hhi.fraunhofer.de, wojciech.samek@hhi.fraunhofer.de  
@date: 08.11.2017  
@version: 1.0  
@copyright: Copyright (c) 2015-2017, Sebastian Lapuschkin, Alexander Binder, Gregoire Montel  
@license : BSD-2-Clause  
  
compute execution times for different lrp computation variants, from naive to optimized.  
...  
  
import time  
import matplotlib.pyplot as plt  
import importlib.util as imp  
import numpy  
import numpy as np  
if imp.find_spec("cupy"): #use cupy for GPU support if available  
    import cupy  
    import cupy as np  
na = np.newaxis  
  
  
  
  
#load a neural network, as well as the MNIST test data and some labels  
nn = model_io.read('../models/MNIST/long-rect.nn') # 99.17% prediction accuracy  
X = data_io.read('../data/MNIST/test_images.npy')  
Y = data_io.read('../data/MNIST/test_labels.npy')  
  
# transfer pixel values from [0 255] to [-1 1] to satisfy the expected input / training pair  
X = X / 127.5 - 1  
  
# transform numeric class labels to vector indicator for uniformity. assume presence of all classes  
I = Y[:,0].astype(int)  
Y = np.zeros([X.shape[0],np.unique(Y).size])  
Y[np.arange(Y.shape[0]),I] = 1  
  
#permute data order for demonstration. or not. your choice.  
I = np.arange(X.shape[0])  
  
  
# do some benchmarking.  
def benchmark():  
    for B in [1, 16, 64]: #batch sizes for work laptop with only 8GB RAM  
        #for B in [1, 16, 64, 256]:
```

```

forward_times_old = []
forward_times_new = []
forward_times_aware = []

lrp_times_old = [] #simple
lrp_times_old2 = [] #epsilon
lrp_times_old3 = [] #alpha 2, beta -1
lrp_times_old4 = [] #alpha 1, beta 0
lrp_times_old5 = [] #alpha 0, beta 1

lrp_times_new = [] #s
lrp_times_new2 = [] #e
lrp_times_new3 = [] #a2b-1
lrp_times_new4 = [] #a1b0
lrp_times_new5 = [] #a0b1

lrp_times_aware = [] #s
lrp_times_aware2 = [] #e
lrp_times_aware3 = [] #a2b-1
lrp_times_aware4 = [] #a1b0
lrp_times_aware5 = [] #a0b1

print('#####')
print('Measuring Speed Gain for batch of {} on FCNN using {}'.format(B, np.__name__))
print('#####')
for i in range(10):
    x = X[:B,:]
    # old code
    t_start = time.time(); yold = nn.forward(x);
    t_start = time.time(); Rold = nn.lrp(yold, 'simple_slow');
    t_start = time.time(); REold = nn.lrp(yold, 'epsilon_slow', 0.01);
    t_start = time.time(); RAold2 = nn.lrp(yold, 'alphabeta_slow', 2.);
    t_start = time.time(); RAold1 = nn.lrp(yold, 'alphabeta_slow', 1.);
    t_start = time.time(); RAold0 = nn.lrp(yold, 'alphabeta_slow', 0.);

    # newer lrp code
    t_start = time.time(); ynew = nn.forward(x);
    t_start = time.time(); Rnew = nn.lrp(ynew, 'simple');
    t_start = time.time(); REnew = nn.lrp(ynew, 'epsilon', 0.01);
    t_start = time.time(); RAnew2 = nn.lrp(ynew, 'alphabeta', 2.);
    t_start = time.time(); RAnew1 = nn.lrp(ynew, 'alphabeta', 1.);
    t_start = time.time(); RAnew0 = nn.lrp(ynew, 'alphabeta', 0.);

    # lrp aware code and forward pass
    t_start = time.time(); yaw = nn.forward(x, lrp_aware=True);
    t_start = time.time(); Raw = nn.lrp(yaw, 'simple');
    t_start = time.time(); REaw= nn.lrp(yaw, 'epsilon', 0.01);
    t_start = time.time(); RAaw2= nn.lrp(yaw, 'alphabeta', 2.);
    t_start = time.time(); RAaw1= nn.lrp(yaw, 'alphabeta', 1.);
    t_start = time.time(); RAaw0= nn.lrp(yaw, 'alphabeta', 0.);


```

```

tolerance = 1e-8
np.testing.assert_allclose(yold, ynew, rtol=tolerance) # predictions
np.testing.assert_allclose(yold, yaw, rtol=tolerance)

np.testing.assert_allclose(Rold, Rnew, rtol=tolerance) # simple lrp maps
np.testing.assert_allclose(Rold, Raw, rtol=tolerance)

np.testing.assert_allclose(Reold, REnew, rtol=tolerance) # eps lrp maps
np.testing.assert_allclose(Reold, REaw, rtol=tolerance)

np.testing.assert_allclose(RAold2, RAnew2, rtol=tolerance) # alpha2 lrp maps
np.testing.assert_allclose(RAold2, RAaw2, rtol=tolerance)

np.testing.assert_allclose(RAold1, RAnew1, rtol=tolerance) # alpha1 lrp maps
np.testing.assert_allclose(RAold1, RAaw1, rtol=tolerance)

np.testing.assert_allclose(RAold0, RAnew0, rtol=tolerance) # alpha0 lrp maps
np.testing.assert_allclose(RAold0, RAaw0, rtol=tolerance)

print('..',end='')

print()
print('    Mean Forward pass times:')
print('        old: ', numpy.mean(forward_times_old), '({}% speedup vs old)'.format(int((forward_times_new - forward_times_old) / forward_times_old * 100)))
print('        new: ', numpy.mean(forward_times_new), '({}% speedup vs old)'.format(int((forward_times_new - forward_times_old) / forward_times_old * 100)))
print('        aware: ', numpy.mean(forward_times_aware), '({}% speedup vs old)'.format(int((forward_times_aware - forward_times_old) / forward_times_old * 100)))
print('    Mean LRP times 1 (simple lrp):')
print('        old: ', numpy.mean(lrp_times_old), '({}% speedup vs old)'.format(int((lrp_times_new - lrp_times_old) / lrp_times_old * 100)))
print('        new: ', numpy.mean(lrp_times_new), '({}% speedup vs old)'.format(int((lrp_times_new - lrp_times_old) / lrp_times_old * 100)))
print('        aware: ', numpy.mean(lrp_times_aware), '({}% speedup vs old)'.format(int((lrp_times_aware - lrp_times_old) / lrp_times_old * 100)))
print('    Mean LRP times 2 (epsilon lrp):')
print('        old: ', numpy.mean(lrp_times_old2), '({}% speedup vs old)'.format(int((lrp_times_new2 - lrp_times_old2) / lrp_times_old2 * 100)))
print('        new: ', numpy.mean(lrp_times_new2), '({}% speedup vs old)'.format(int((lrp_times_new2 - lrp_times_old2) / lrp_times_old2 * 100)))
print('        aware: ', numpy.mean(lrp_times_aware2), '({}% speedup vs old)'.format(int((lrp_times_aware2 - lrp_times_old2) / lrp_times_old2 * 100)))
print('    Mean LRP times 3 (alpha=2 lrp):')
print('        old: ', numpy.mean(lrp_times_old3), '({}% speedup vs old)'.format(int((lrp_times_new3 - lrp_times_old3) / lrp_times_old3 * 100)))
print('        new: ', numpy.mean(lrp_times_new3), '({}% speedup vs old)'.format(int((lrp_times_new3 - lrp_times_old3) / lrp_times_old3 * 100)))
print('        aware: ', numpy.mean(lrp_times_aware3), '({}% speedup vs old)'.format(int((lrp_times_aware3 - lrp_times_old3) / lrp_times_old3 * 100)))
print('    Mean LRP times 4 (alpha=1 lrp):')
print('        old: ', numpy.mean(lrp_times_old4), '({}% speedup vs old)'.format(int((lrp_times_new4 - lrp_times_old4) / lrp_times_old4 * 100)))
print('        new: ', numpy.mean(lrp_times_new4), '({}% speedup vs old)'.format(int((lrp_times_new4 - lrp_times_old4) / lrp_times_old4 * 100)))
print('        aware: ', numpy.mean(lrp_times_aware4), '({}% speedup vs old)'.format(int((lrp_times_aware4 - lrp_times_old4) / lrp_times_old4 * 100)))
print('    Mean LRP times 5 (alpha=0 lrp):')
print('        old: ', numpy.mean(lrp_times_old5), '({}% speedup vs old)'.format(int((lrp_times_new5 - lrp_times_old5) / lrp_times_old5 * 100)))
print('        new: ', numpy.mean(lrp_times_new5), '({}% speedup vs old)'.format(int((lrp_times_new5 - lrp_times_old5) / lrp_times_old5 * 100)))
print('        aware: ', numpy.mean(lrp_times_aware5), '({}% speedup vs old)'.format(int((lrp_times_aware5 - lrp_times_old5) / lrp_times_old5 * 100)))
print('    Mean Total times with LRP once:')
oldtotal = numpy.mean(numpy.array(lrp_times_old) + numpy.array(forward_times_old))
newtotal = numpy.mean(numpy.array(lrp_times_new) + numpy.array(forward_times_new))
awaretotal = numpy.mean(numpy.array(lrp_times_aware) + numpy.array(forward_times_aware))

```

```

print('      old: ', oldtotal, '({}% speedup vs old)'.format(int(100*(1 - oldtotal)))
print('      new: ', newtotal, '({}% speedup vs old)'.format(int(100*(1 - newtotal)))
print('      aware:', awaretotal, '({}% speedup vs old)'.format(int(100*(1 - awaretotal)))
print('      Mean Total times with LRP twice (simple+epsilon):')
oldtotaltwice = numpy.mean(numpy.array(lrp_times_old) + numpy.array(lrp_times_old2))
newtotaltwice = numpy.mean(numpy.array(lrp_times_new) + numpy.array(lrp_times_new2))
arewtotaltwice = numpy.mean(numpy.array(lrp_times_aware) + numpy.array(lrp_times_aware2))
print('      old: ', oldtotaltwice, '({}% speedup vs old)'.format(int(100*(1 - oldtotaltwice))))
print('      new: ', newtotaltwice, '({}% speedup vs old)'.format(int(100*(1 - newtotaltwice))))
print('      aware:', arewtotaltwice, '({}% speedup vs old)'.format(int(100*(1 - arewtotaltwice))))
print('      Mean Total times with LRP five times(simple,epsilon,alpha=2,alpha=1,alpha=0):')
oldtotalfive = oldtotaltwice + numpy.mean(numpy.array(lrp_times_old3)) + numpy.array(lrp_times_old4)
newtotalfive = newtotaltwice + numpy.mean(numpy.array(lrp_times_new3)) + numpy.array(lrp_times_new4)
arewtotalfive = arewtotaltwice + numpy.mean(numpy.array(lrp_times_aware3)) + numpy.array(lrp_times_aware4)
print('      old: ', oldtotalfive, '({}% speedup vs old)'.format(int(100*(1 - oldtotalfive))))
print('      new: ', newtotalfive, '({}% speedup vs old)'.format(int(100*(1 - newtotalfive))))
print('      aware:', arewtotalfive, '({}% speedup vs old)'.format(int(100*(1 - arewtotalfive))))
print('')
print('')

```

#run benchmark

```

→ aware: 0.0016381502151489257 (8% speedup vs old)
Mean Total times with LRP once:
old: 0.0032855749130249025 (0% speedup vs old)
new: 0.001794886589050293 (45% speedup vs old)
aware: 0.0017532587051391601 (46% speedup vs old)
Mean Total times with LRP twice (simple+epsilon):
old: 0.004994511604309082 (0% speedup vs old)
new: 0.003051614761352539 (38% speedup vs old)
aware: 0.003071498870849609 (38% speedup vs old)
Mean Total times with LRP five times(simple,epsilon,alpha=2,alpha=1,alpha=0):
old: 0.011979770660400391 (0% speedup vs old)
new: 0.00930490493774414 (22% speedup vs old)
aware: 0.008894991874694825 (25% speedup vs old)

```

```

#####
#asuring Speed Gain for batch of 64 on FCNN using cupy
#####
.....
Mean Forward pass times:
old: 0.0020311355590820314 (0% speedup vs old)
new: 0.0007555484771728516 (62% speedup vs old)
aware: 0.001950240135192871 (3% speedup vs old)
Mean LRP times 1 (simple lrp):
old: 0.0031230688095092774 (0% speedup vs old)
new: 0.00279292460757666 (10% speedup vs old)

```