In [1]:

```
import torch
import torch.nn as nn
from torch.utils.data import DataLoader
import numpy as np
import logging

from utils.metrics import Metrics
from utils.utils import convertModel
from models.model import Model
from fer2013_dataset import FER2013Dataset

from utils.backprop import Backprop

from utils.viz import visualize_kernels, visualize_feature_maps
import os
os.environ['CUDA_VISIBLE_DEVICES'] = '-1'
logging.disable(logging.CRITICAL)
```

Load Untrained Model and visualize saliency and feature maps

In [2]:

```
# Number of samples to display for image related visualizations!
view_samples = 10

device = torch.device("cpu")

test_dataset = FER2013Dataset("data/fer2013/fer2013/fer2013.csv", "PrivateTest")

test_loader = DataLoader(test_dataset, batch_size=1, shuffle=True, pin_memory=True)

model = Model("results/bestModel.json")

model.eval()
```

['Training' 'PublicTest' 'PrivateTest']

```
Out[2]:
```

```
Model(
  (network): ModuleList(
    (0): ConvBlock(
      (conv): Conv2d(1, 32, kernel size=(3, 3), stride=(1, 1), paddi
ng=(2, 2)
      (normalization): BatchNorm2d(32, eps=1e-05, momentum=0.1, affi
ne=True, track running stats=True)
      (activation): ReLU(inplace=True)
    (1): ConvBlock(
      (conv): Conv2d(32, 32, kernel size=(3, 3), stride=(1, 1), padd
ing=(2, 2)
      (normalization): BatchNorm2d(32, eps=1e-05, momentum=0.1, affi
ne=True, track running stats=True)
      (activation): ReLU(inplace=True)
      (pooling): MaxPool2d(kernel size=2, stride=2, padding=0, dilat
ion=1, ceil mode=False)
    (2): ConvBlock(
      (conv): Conv2d(32, 64, kernel size=(3, 3), stride=(1, 1), padd
ing=(2, 2)
      (normalization): BatchNorm2d(64, eps=1e-05, momentum=0.1, affi
ne=True, track_running stats=True)
      (activation): ReLU(inplace=True)
    (3): ConvBlock(
      (conv): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1), padd
ing=(1, 1)
      (normalization): BatchNorm2d(64, eps=1e-05, momentum=0.1, affi
ne=True, track running stats=True)
      (activation): ReLU(inplace=True)
      (pooling): MaxPool2d(kernel size=2, stride=2, padding=0, dilat
ion=1, ceil mode=False)
    (4): ConvBlock(
      (conv): Conv2d(64, 128, kernel size=(3, 3), stride=(1, 1), pad
ding=(1, 1)
      (normalization): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
      (activation): ReLU(inplace=True)
    (5): ConvBlock(
      (conv): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1), pa
dding=(1, 1)
      (normalization): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
      (activation): ReLU(inplace=True)
    (6): ConvBlock(
      (conv): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1), pa
dding=(1, 1)
      (normalization): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track_running_stats=True)
      (activation): ReLU(inplace=True)
    (7): AdaptiveAvgPool2d(output size=(5, 5))
    (8): NetworkHead(
      (dropout1): Dropout(p=0.5, inplace=False)
      (fc1): Linear(in features=6400, out features=64, bias=True)
```

1), padding=(2, 2))

first conv layer

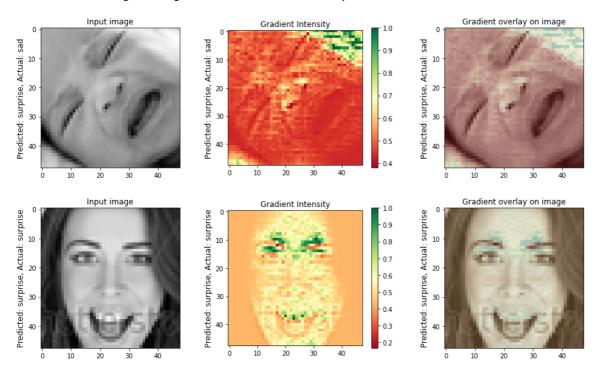
```
Saliency Map visualize by backpropagating target class to the
```

In [4]:

```
for idx, batch in enumerate(test_loader):
    image, target = batch["image"].to(device), batch["emotion"].to(device)
    image.requires_grad = True
    backprop.visualize(image, target.item(), cmap='RdYlGn', guided=True, alpha=
0.3, class_mapping=test_dataset.get_class_mapping())
    if idx == view_samples:
        break
```

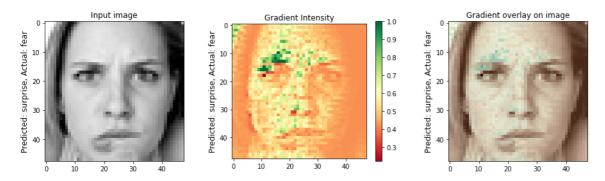
/repos/MSAI/ComputerVision-KEN4255/Emotion-Recognition/utils/backpro p.py:100: UserWarning: The predicted class index 5 does notequal the target class index 4. Calculating the gradient w.r.t. the predicted class.

'Calculating the gradient w.r.t. the predicted class.'



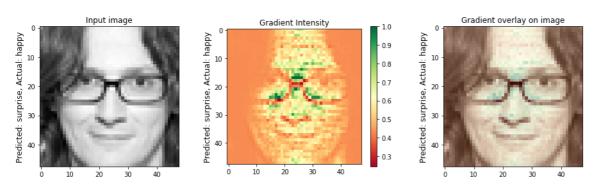
/repos/MSAI/ComputerVision-KEN4255/Emotion-Recognition/utils/backprop.py:100: UserWarning: The predicted class index 5 does notequal the target class index 2. Calculating the gradient w.r.t. the predicted class.

'Calculating the gradient w.r.t. the predicted class.'



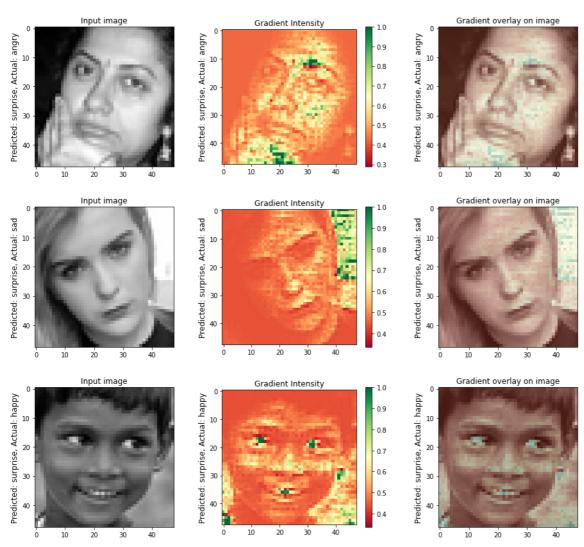
/repos/MSAI/ComputerVision-KEN4255/Emotion-Recognition/utils/backpro p.py:100: UserWarning: The predicted class index 5 does notequal the target class index 3. Calculating the gradient w.r.t. the predicted class.

'Calculating the gradient w.r.t. the predicted class.'



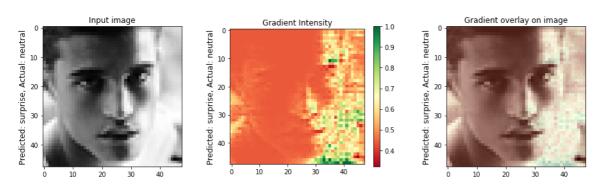
/repos/MSAI/ComputerVision-KEN4255/Emotion-Recognition/utils/backpro p.py:100: UserWarning: The predicted class index 5 does notequal the target class index 0. Calculating the gradient w.r.t. the predicted class.

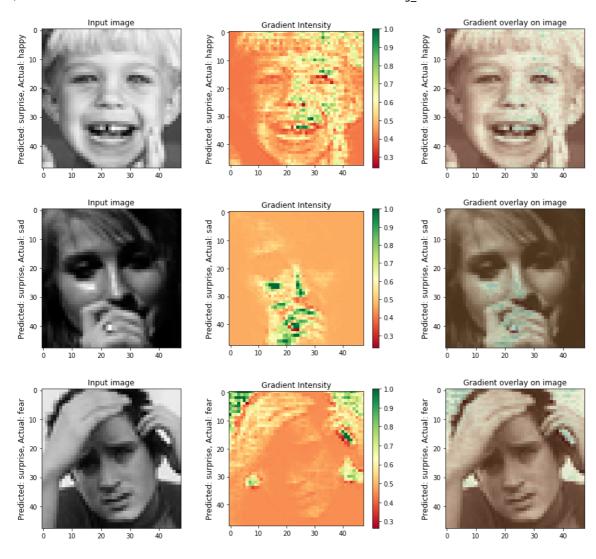
'Calculating the gradient w.r.t. the predicted class.'



/repos/MSAI/ComputerVision-KEN4255/Emotion-Recognition/utils/backpro p.py:100: UserWarning: The predicted class index 5 does notequal the target class index 6. Calculating the gradient w.r.t. the predicted class.

'Calculating the gradient w.r.t. the predicted class.'





Visualize Feature Maps

In [5]:

```
import matplotlib.pyplot as plt

activation = visualize_feature_maps(model.modules())

batch = next(iter(test_loader))
image, target = batch["image"].to(device), batch["emotion"].to(device)
out = model(image)
fig, axarr = plt.subplots(len(activation), 5, figsize=(100, 100))

for convidx, act in enumerate(activation):
    act = activation[act].squeeze()
    for idx in range(5):
        axarr[convidx, idx].imshow(act[idx], cmap='jet')
```

```
Conv2d(1, 32, kernel_size=(3, 3), stride=(1, 1), padding=(2, 2))
Conv2d(32, 32, kernel_size=(3, 3), stride=(1, 1), padding=(2, 2))
Conv2d(32, 64, kernel_size=(3, 3), stride=(1, 1), padding=(2, 2))
conv3
Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1), padding=(1, 1))
conv4
Conv2d(64, 128, kernel size=(3, 3), stride=(1, 1), padding=(1, 1))
conv5
Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1), padding=(1, 1))
Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1))
conv7
```

Load Trained Model

```
In [6]:
```

```
model = convertModel("results/bestModel.pth.tar", model).to(device)
model.eval()
```

dict_keys(['state', 'param_groups'])
Loaded Model: results/bestModel.pth.tar successfully

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Out[6]:

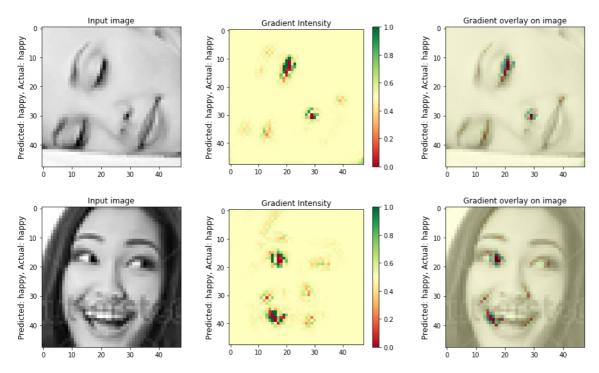
```
Model(
  (network): ModuleList(
    (0): ConvBlock(
      (conv): Conv2d(1, 32, kernel size=(3, 3), stride=(1, 1), paddi
ng=(2, 2)
      (normalization): BatchNorm2d(32, eps=1e-05, momentum=0.1, affi
ne=True, track running stats=True)
      (activation): ReLU(inplace=True)
    (1): ConvBlock(
      (conv): Conv2d(32, 32, kernel size=(3, 3), stride=(1, 1), padd
ing=(2, 2)
      (normalization): BatchNorm2d(32, eps=1e-05, momentum=0.1, affi
ne=True, track running stats=True)
      (activation): ReLU(inplace=True)
      (pooling): MaxPool2d(kernel size=2, stride=2, padding=0, dilat
ion=1, ceil mode=False)
    (2): ConvBlock(
      (conv): Conv2d(32, 64, kernel size=(3, 3), stride=(1, 1), padd
ing=(2, 2)
      (normalization): BatchNorm2d(64, eps=1e-05, momentum=0.1, affi
ne=True, track_running stats=True)
      (activation): ReLU(inplace=True)
    (3): ConvBlock(
      (conv): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1), padd
ing=(1, 1)
      (normalization): BatchNorm2d(64, eps=1e-05, momentum=0.1, affi
ne=True, track running stats=True)
      (activation): ReLU(inplace=True)
      (pooling): MaxPool2d(kernel size=2, stride=2, padding=0, dilat
ion=1, ceil mode=False)
    (4): ConvBlock(
      (conv): Conv2d(64, 128, kernel size=(3, 3), stride=(1, 1), pad
ding=(1, 1)
      (normalization): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
      (activation): ReLU(inplace=True)
    (5): ConvBlock(
      (conv): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1), pa
dding=(1, 1)
      (normalization): BatchNorm2d(128, eps=1e-05, momentum=0.1, aff
ine=True, track running stats=True)
      (activation): ReLU(inplace=True)
    (6): ConvBlock(
      (conv): Conv2d(128, 256, kernel size=(3, 3), stride=(1, 1), pa
dding=(1, 1)
      (normalization): BatchNorm2d(256, eps=1e-05, momentum=0.1, aff
ine=True, track_running_stats=True)
      (activation): ReLU(inplace=True)
    (7): AdaptiveAvgPool2d(output size=(5, 5))
    (8): NetworkHead(
      (dropout1): Dropout(p=0.5, inplace=False)
      (fc1): Linear(in features=6400, out features=64, bias=True)
```

```
(relu): ReLU(inplace=True)
        (final_layer): Linear(in_features=64, out_features=7, bias=Tru
e)
    )
    )
)
```

In [7]:

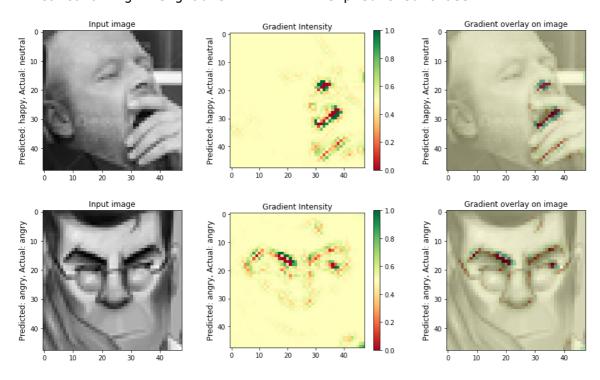
```
backprop = Backprop(model)
for idx, batch in enumerate(test_loader):
    image, target = batch["image"].to(device), batch["emotion"].to(device)
    image.requires_grad = True
    backprop.visualize(image, target.item(), cmap='RdYlGn', guided=True, alpha=
0.5, class_mapping=test_dataset.get_class_mapping())
    if idx == view_samples:
        break
```

Registered Hook at: {} Conv2d(1, 32, kernel_size=(3, 3), stride=(1, 1), padding=(2, 2))



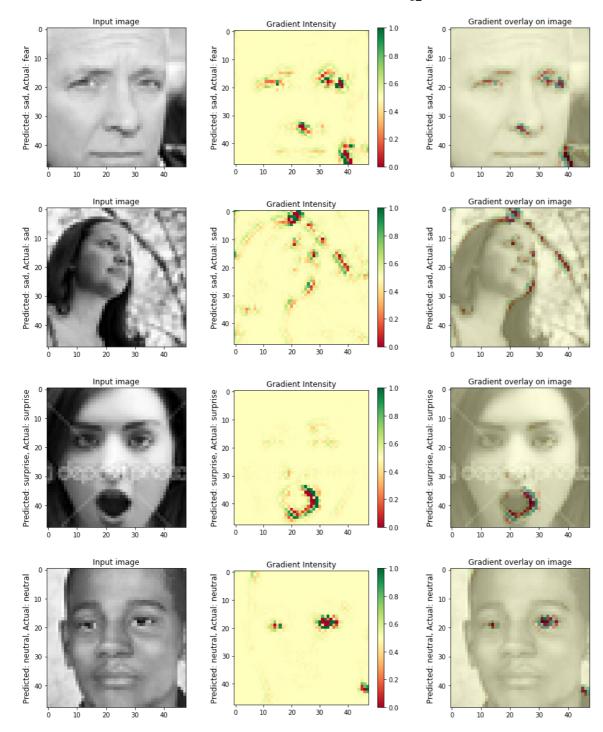
/repos/MSAI/ComputerVision-KEN4255/Emotion-Recognition/utils/backprop.py:100: UserWarning: The predicted class index 3 does notequal the target class index 6. Calculating the gradient w.r.t. the predicted class.

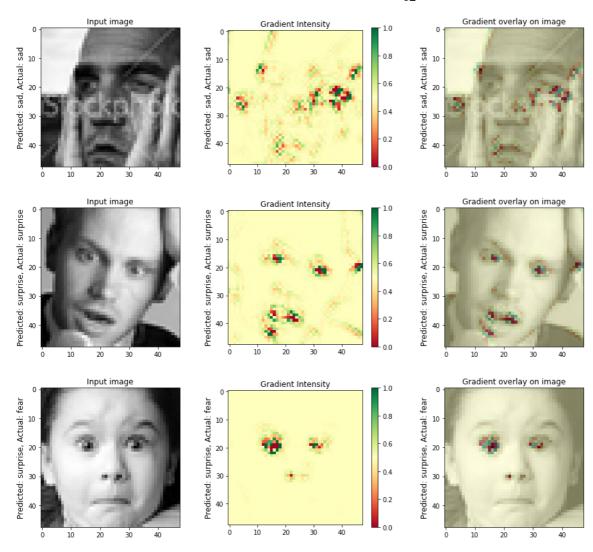
'Calculating the gradient w.r.t. the predicted class.'



/repos/MSAI/ComputerVision-KEN4255/Emotion-Recognition/utils/backpro p.py:100: UserWarning: The predicted class index 4 does notequal the target class index 2. Calculating the gradient w.r.t. the predicted class.

'Calculating the gradient w.r.t. the predicted class.'





In [8]:

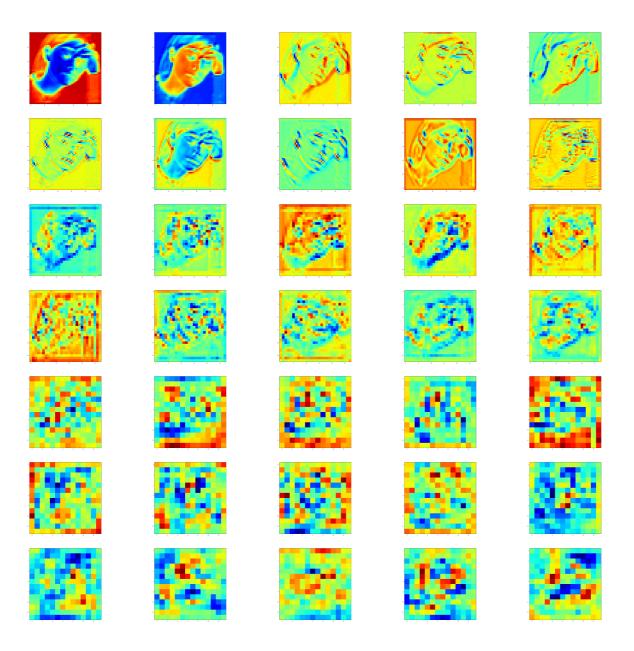
```
import matplotlib.pyplot as plt

activation = visualize_feature_maps(model.modules())

batch = next(iter(test_loader))
image, target = batch["image"].to(device), batch["emotion"].to(device)
out = model(image)
fig, axarr = plt.subplots(len(activation), 5, figsize=(100, 100))

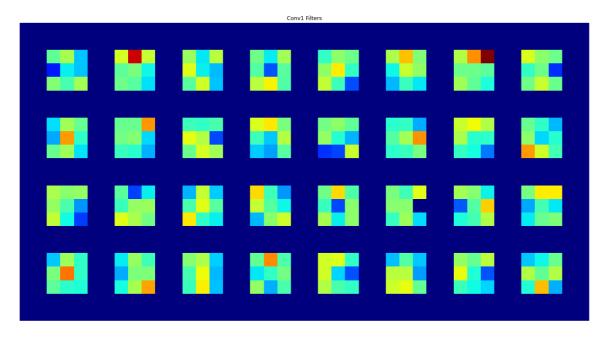
for convidx, act in enumerate(activation):
    act = activation[act].squeeze()
    for idx in range(5):
        axarr[convidx, idx].imshow(act[idx], cmap='jet')
```

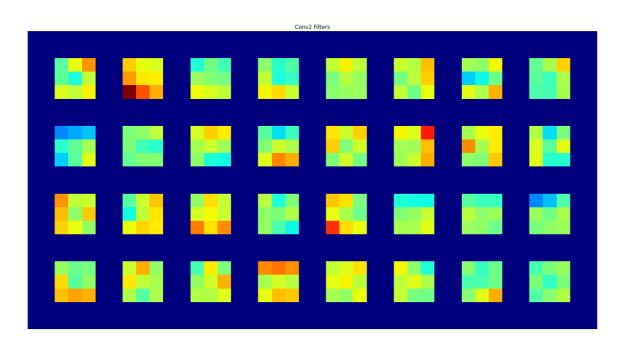
```
Conv2d(1, 32, kernel_size=(3, 3), stride=(1, 1), padding=(2, 2))
conv1
Conv2d(32, 32, kernel_size=(3, 3), stride=(1, 1), padding=(2, 2))
conv2
Conv2d(32, 64, kernel_size=(3, 3), stride=(1, 1), padding=(2, 2))
conv3
Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
conv4
Conv2d(64, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
conv5
Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
conv6
Conv2d(128, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
conv7
```

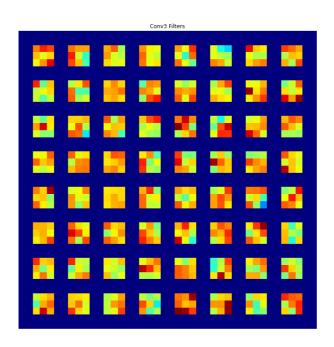


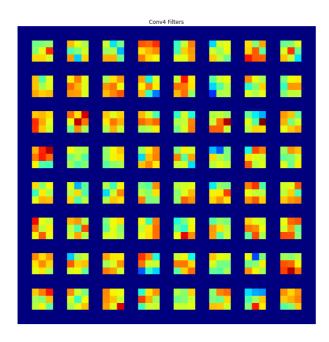
In [9]:

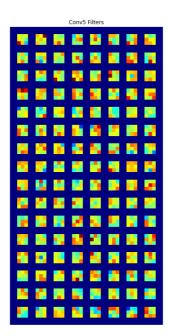
visualize_kernels(model.modules())

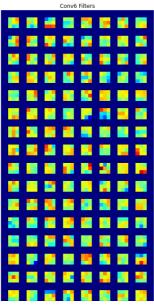


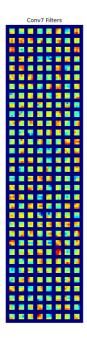












In []:			