The summary of the 6-layer CNN network:

```
Classifier_1d_6_conv_v2(
   (raw): Sequential(
    (0): SepConv1d v4(
      (layers): Sequential(
         (0): Conv2d(2, 2, kernel\_size=(1, 8), stride=(1, 2), padding=(0, 3), groups=2)
         (1): Conv2d(2, 32, kernel_size=(1, 1), stride=(1, 1))
         (2): BatchNorm2d(32, eps=Te-05, momentum=0.1, affine=True, track_running_stats=True)
         (3): ReLU(inplace=True)
         (4): Dropout(p=0.5, inplace=False)
    (1): SepConv1d v4(
      (layers): Sequential(
         (0): Conv2d(32, 32, kernel_size=(1, 8), stride=(1, 4), padding=(0, 2), groups=32)
         (1): Conv2d(32, 64, kernel_size=(1, 1), stride=(1, 1))
(2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
         (3): ReLU(inplace=True)
         (4): Dropout(p=0.5, inplace=False)
    (2): SepConv1d_v4(
    (layers): Sequential(
        (0): Conv2d(64, 64, kernel_size=(1, 8), stride=(1, 4), padding=(0, 2), groups=64)
(1): Conv2d(64, 128, kernel_size=(1, 1), stride=(1, 1))
         (2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
         (3): ReLU(inplace=True)
         (4): Dropout(p=0.5, inplace=False)
    (3): SepConv1d_v4(
       (layers): Sequential(
         (0): Conv2d(128, 128, kernel size=(1, 8), stride=(1, 4), padding=(0, 2), groups=128)
         (1): Conv2d(128, 256, kernel_size=(1, 1), stride=(1, 1))
         (2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
         (3): ReLU(inplace=True)
         (4): Dropout(p=0.5, inplace=False)
    (4): SepConv1d_v4(
       (layers): Sequential(
         (0): Conv2d(256, 256, kernel_size=(1, 8), stride=(1, 4), padding=(0, 2), groups=256) (1): Conv2d(256, 512, kernel_size=(1, 1), stride=(1, 1))
         (2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
         (3): ReLU(inplace=True)
         (4): Dropout(p=0.5, inplace=False)
    (5): SepConv1d_v4(
       (layers): Sequential(
         (0): Conv2d(512, 512, kernel_size=(1, 8), stride=(1, 4), padding=(0, 2), groups=512)
         (1): Conv2d(512, 1024, kernel_size=(1, 1), stride=(1, 1))
         (2): BatchNorm2d(1024, eps=le-05, momentum=0.1, affine=True, track_running_stats=True)
         (3): ReLU(inplace=True)
         (4): Dropout(p=0.2, inplace=False)
  (FC): Sequential(
    (0): Flatten()
    (1): Linear(in_features=1024, out_features=128, bias=True)
    (2): ReLU(inplace=True)
    (3): Dropout(p=0.5, inplace=False)
    (4): Linear(in_features=128, out_features=128, bias=True)
    (5): ReLU(inplace=True)
    (6): Dropout(p=0.5, inplace=False)
  (out): Sequential(
    (0): Linear(in_features=128, out_features=2, bias=True)
  (quant): QuantStub()
  (dequant): DeQuantStub()
```

## Table of the network parameters:

Layer (type)	Output Shape	Param #
QuantStub-1	[512, 2, 2048]	0
Conv2d-2	[512, 2, 1, 1024]	18
Conv2d-3	[512, 32, 1, 1024]	96
BatchNorm2d-4	[512, 32, 1, 1024]	64
ReLU-5	[512, 32, 1, 1024]	0
Dropout-6	[512, 32, 1, 1024]	0
SepConv1d v4-7	[512, 32, 1, 1024]	0
Conv2d-8	[512, 32, 1, 256]	288
Conv2d-9	[512, 64, 1, 256]	2,112
BatchNorm2d-10	[512, 64, 1, 256]	128
ReLU-11	[512, 64, 1, 256]	0
Dropout-12	[512, 64, 1, 256]	0
SepConv1d v4-13	[512, 64, 1, 256]	0
Conv2d - 14	[512, 64, 1, 64]	576
Conv2d-15	[512, 128, 1, 64]	8,320
BatchNorm2d-16	[512, 128, 1, 64]	256
ReLU-17	[512, 128, 1, 64]	0
Dropout-18	[512, 128, 1, 64]	0
SepConv1d v4-19	[512, 128, 1, 64]	Θ
Conv2d-20	[512, 128, 1, 16]	1,152
Conv2d-21	[512, 256, 1, 16]	33,024
BatchNorm2d-22	[512, 256, 1, 16]	512
ReLU-23	[512, 256, 1, 16]	0
Dropout-24	[512, 256, 1, 16]	0
SepConv1d v4-25	[512, 256, 1, 16]	0
Conv2d-26	[512, 256, 1, 4]	2,304
Conv2d-27	[512, 512, 1, 4]	131,584
BatchNorm2d-28	[512, 512, 1, 4]	1,024
ReLU-29	[512, 512, 1, 4]	0
Dropout-30	[512, 512, 1, 4]	0
SepConv1d v4-31	[512, 512, 1, 4]	0
Conv2d-32	[512, 512, 1, 1]	4,608
Conv2d-33	[512, 1024, 1, 1]	525,312
BatchNorm2d-34	[512, 1024, 1, 1]	2,048
ReLU-35	[512, 1024, 1, 1]	0
Dropout-36	[512, 1024, 1, 1]	0
SepConv1d v4-37	[512, 1024, 1, 1]	ő
Flatten-38	[512, 1024, 1, 1]	Õ
Linear-39	[512, 1024]	131,200
ReLU-40	[512, 128]	0
Dropout-41	[512, 128]	Õ
Linear-42	[512, 128]	16,512
ReLU-43	[512, 128]	0
Dropout-44	[512, 128]	0
Linear-45	[512, 120]	258
DeQuantStub-46	[512, 2]	0
	[J12, 2]	

Total params: 861,396 Trainable params: 861,396 Non-trainable params: 0 -----Input size (MB): 8.00 Forward/backward pass size (MB): 1353.02 Params size (MB): 3.29 Estimated Total Size (MB): 1364.30 \_\_\_\_\_\_

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Detail of the network's per layer computations and parameters:

```
Classifier_1d_6_conv_v2(
 0.861 M, 100.000% Params, 3.254 MMac, 100.000% MACs,
  (raw): Sequential(
   0.713 M, 82.822% Params, 3.106 MMac, 95.453% MACs,
    (0): SepConv1d_v4(
     0.0 M, 0.021% Params, 0.215 MMac, 6.608% MACs,
      (layers): Sequential(
       0.0 M, 0.021% Params, 0.215 MMac, 6.608% MACs,
        (0): Conv2d(0.0 M, 0.002% Params, 0.018 MMac, 0.566% MACs, 2, 2, kernel_size=(1, 8), stride=(1, 2), padding=(0, 3), groups=2) (1): Conv2d(0.0 M, 0.011% Params, 0.098 MMac, 3.021% MACs, 2, 32, kernel_size=(1, 1), stride=(1, 1))
        (2): BatchNorm2d(0.0 M, 0.007% Params, 0.066 MMac, 2.014% MACs, 32, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
        (3): ReLU(0.0 M, 0.000% Params, 0.033 MMac, 1.007% MACs, inplace=True)
        (4): Dropout(0.0 M, 0.000% Params, 0.0 MMac, 0.000% MACs, p=0.5, inplace=False)
    (1): SepConv1d v4(
     0.003 M, 0.293% Params, 0.664 MMac, 20.390% MACs,
      (layers): Sequential(
       0.003 M, 0.293% Params, 0.664 MMac, 20.390% MACs,
        (0): Conv2d(0.0 M, 0.033% Params, 0.074 MMac, 2.266% MACs, 32, 32, kernel_size=(1, 8), stride=(1, 4), padding=(0, 2), groups=32)
        (1): Conv2d(0.002 M, 0.245% Params, 0.541 MMac, 16.614% MACs, 32, 64, kernel_size=(1, 1), stride=(1, 1))
        (2): BatchNorm2d(0.0 M, 0.015% Params, 0.033 MMac, 1.007% MACs, 64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
        (3): ReLU(0.0 M, 0.000% Params, 0.016 MMac, 0.503% MACs, inplace=True)
        (4): Dropout(0.0 M, 0.000% Params, 0.0 MMac, 0.000% MACs, p=0.5, inplace=False)
    (2): SepConv1d_v4(
     0.009 \text{ M}, 1.0\overline{6}2\% Params, 0.594 \text{ MMac}, 18.250\% MACs,
      (layers): Sequential(
       0.009 M, 1.062% Params, 0.594 MMac, 18.250% MACs,
        (0): Conv2d(0.001 M, 0.067% Params, 0.037 MMac, 1.133% MACs, 64, 64, kernel_size=(1, 8), stride=(1, 4), padding=(0, 2), groups=64)
        (1): Conv2d(0.008 M, 0.966% Params, 0.532 MMac, 16.362% MACs, 64, 128, kernel_size=(1, 1), stride=(1, 1))
        (2): BatchNorm2d(0.0 M, 0.030% Params, 0.016 MMac, 0.503% MACs, 128, eps=1e-0\overline{5}, momentum=0.1, affine=True, track running stats=True)
        (3): ReLU(0.0 M, 0.000% Params, 0.008 MMac, 0.252% MACs, inplace=True)
        (4): Dropout(0.0 M, 0.000% Params, 0.0 MMac, 0.000% MACs, p=0.5, inplace=False)
    (3): SepConv1d_v4(
     0.035 M, 4.027% Params, 0.559 MMac, 17.181% MACs,
      (layers): Sequential(
       0.035 M, 4.027% Params, 0.559 MMac, 17.181% MACs,
        (0): Conv2d(0.001 M, 0.134% Params, 0.018 MMac, 0.566% MACs, 128, 128, kernel_size=(1, 8), stride=(1, 4), padding=(0, 2), groups=128)
        (1): Conv2d(0.033 M, 3.834% Params, 0.528 MMac, 16.237% MACs, 128, 256, kernel size=(1, 1), stride=(1, 1))
        (2): BatchNorm2d(0.001 M, 0.059% Params, 0.008 MMac, 0.252% MACs, 256, eps=1e-\overline{0}5, momentum=0.1, affine=True, track_running_stats=True)
        (3): ReLU(0.0 M, 0.000% Params, 0.004 MMac, 0.126% MACs, inplace=True)
        (4): Dropout(0.0 M, 0.000% Params, 0.0 MMac, 0.000% MACs, p=0.5, inplace=False)
    (4): SepConv1d_v4(
     0.135 M, 15.662% Params, 0.542 MMac, 16.646% MACs,
      (layers): Sequential(
       0.135 M, 15.662% Params, 0.542 MMac, 16.646% MACs,
        (0): Conv2d(0.002 M, 0.267% Params, 0.009 MMac, 0.283% MACs, 256, 256, kernel_size=(1, 8), stride=(1, 4), padding=(0, 2), groups=256)
        (1): Conv2d(0.132 M, 15.276% Params, 0.526 MMac, 16.174% MACs, 256, 512, kernel_size=(1, 1), stride=(1, 1))
        (2): BatchNorm2d(0.001 M, 0.119% Params, 0.004 MMac, 0.126% MACs, 512, eps=1e-0\overline{5}, momentum=0.1, affine=True, track_running_stats=True)
        (3): ReLU(0.0 M, 0.000% Params, 0.002 MMac, 0.063% MACs, inplace=True)
        (4): Dropout(0.0 M, 0.000% Params, 0.0 MMac, 0.000% MACs, p=0.5, inplace=False)
    (5): SepConv1d_v4(
      0.532 M, 61.756% Params, 0.533 MMac, 16.378% MACs,
      (layers): Sequential(
       0.532 M, 61.756% Params, 0.533 MMac, 16.378% MACs,
        (0): Conv2d(0.005 M, 0.535% Params, 0.005 MMac, 0.142% MACs, 512, 512, kernel_size=(1, 8), stride=(1, 4), padding=(0, 2), groups=512)
        (1): Conv2d(0.525 M, 60.984% Params, 0.525 MMac, 16.142% MACs, 512, 1024, kernel_size=(1, 1), stride=(1, 1))
        (2): BatchNorm2d(0.002 M, 0.238% Params, 0.002 MMac, 0.063% MACs, 1024, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
        (3): ReLU(0.0 M, 0.000% Params, 0.001 MMac, 0.031% MACs, inplace=True)
        (4): Dropout(0.0 M, 0.000% Params, 0.0 MMac, 0.000% MACs, p=0.2, inplace=False)
  (FC): Sequential(
   0.148 M, 17.148% Params, 0.148 MMac, 4.539% MACs,
    (0): Flatten(0.0 M, 0.000% Params, 0.0 MMac, 0.000% MACs, )
    (1): Linear(0.131 M, 15.231% Params, 0.131 MMac, 4.028% MACs, in_features=1024, out_features=128, bias=True)
    (2): ReLU(0.0 M, 0.000% Params, 0.0 MMac, 0.004% MACs, inplace=True)
    (3): Dropout(0.0 M, 0.000% Params, 0.0 MMac, 0.000% MACs, p=0.5, inplace=False)
    (4): Linear(0.017 M, 1.917% Params, 0.016 MMac, 0.503% MACs, in_features=128, out_features=128, bias=True)
    (5): ReLU(0.0 M, 0.000% Params, 0.0 MMac, 0.004% MACs, inplace=True)
    (6): Dropout(0.0 M, 0.000% Params, 0.0 MMac, 0.000% MACs, p=0.5, inplace=False)
  (out): Sequential(
   0.0 M, 0.030% Params, 0.0 MMac, 0.008% MACs,
    (0): Linear(0.0 M, 0.030% Params, 0.0 MMac, 0.008% MACs, in features=128, out features=2, bias=True)
  (quant): QuantStub(0.0 M, 0.000% Params, 0.0 MMac, 0.000% MACs, )
  (dequant): DeQuantStub(0.0 M, 0.000% Params, 0.0 MMac, 0.000% MACs, )
______
______
```

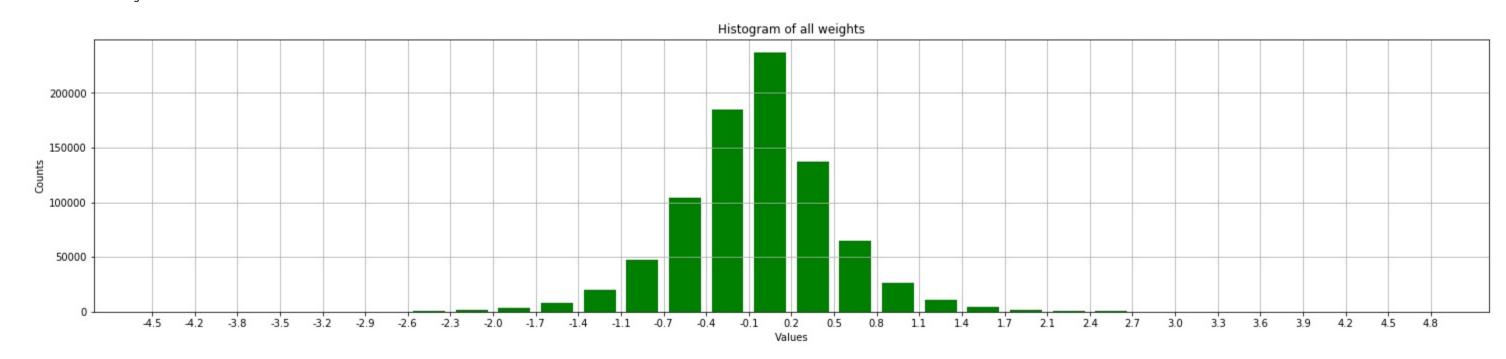
The network's accuracy:

The network has the following accuracy: TP: 99.10 , FP: 3.88 AF\_threshold = 3 (observing at least 3 AF signs to classify ECG as AF) TP: 98.77 , FP: 1.5 AF threshold = 7

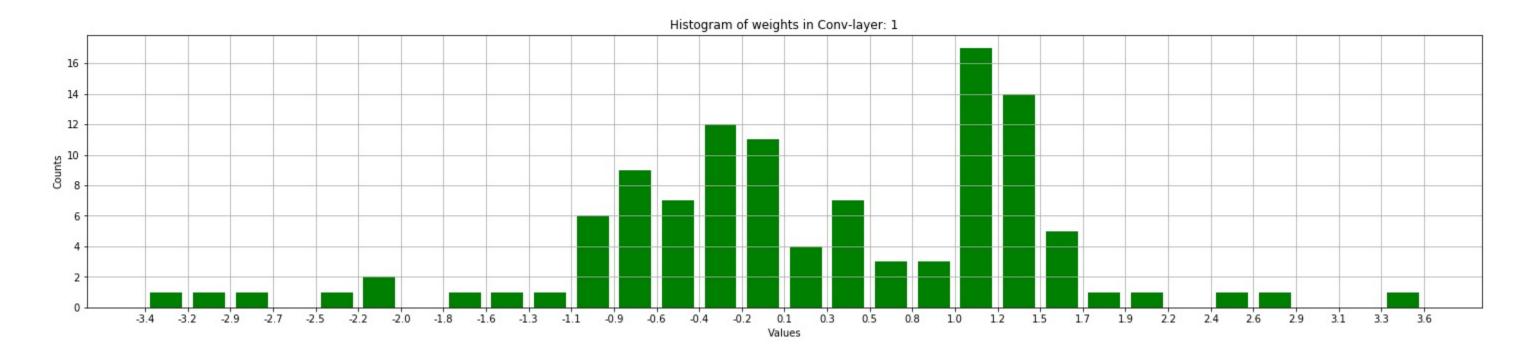
Statistical analysis of the weights:

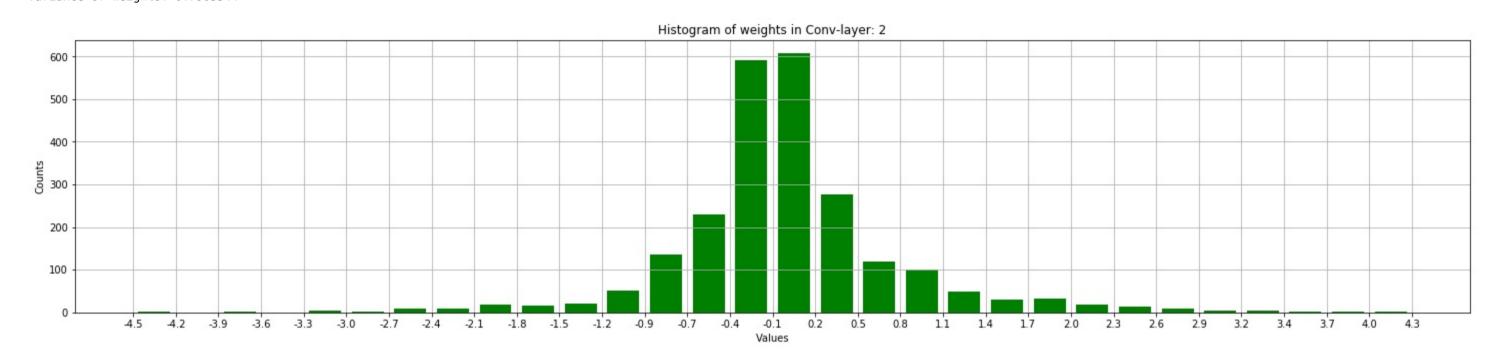
======== All weights of the network

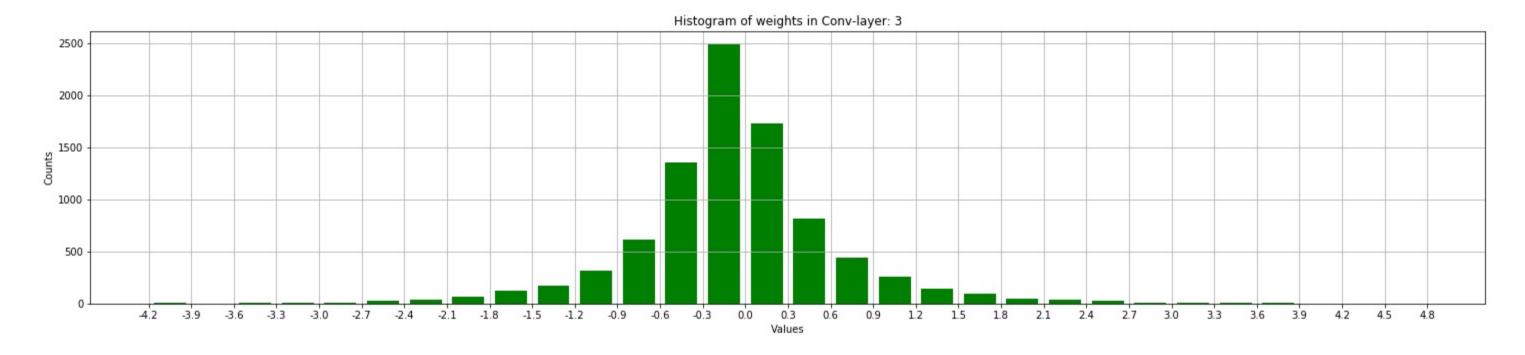
Minimum of weights: -4.4108548 Maximum of weights: 4.8913217 Average value of weights: -0.0088230 Variance of weights: 0.5572891



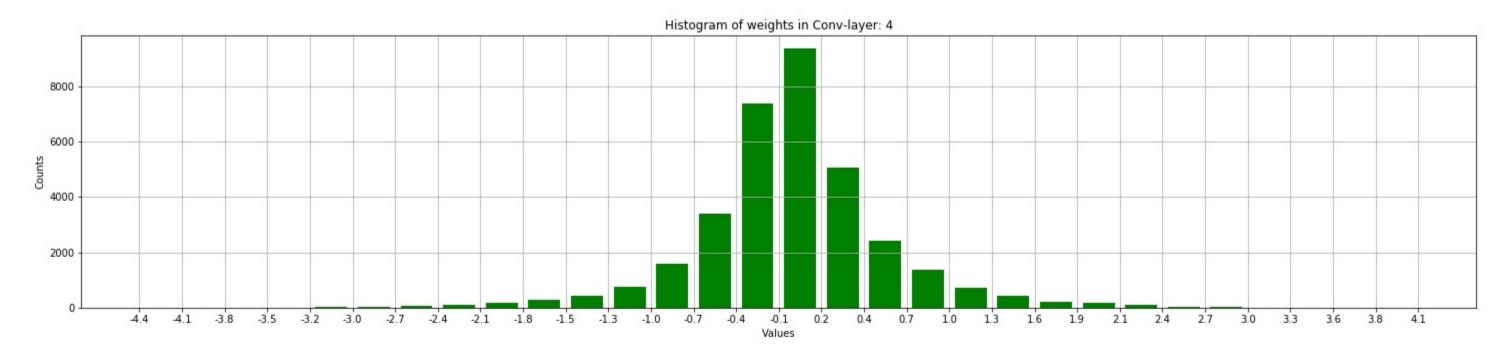
====== Conv. Layer: 1 Minimum of weights: -3.3641193 Maximum of weights: 3.6236877 Average value of weights: 0.2965188 Variance of weights: 1.1729762

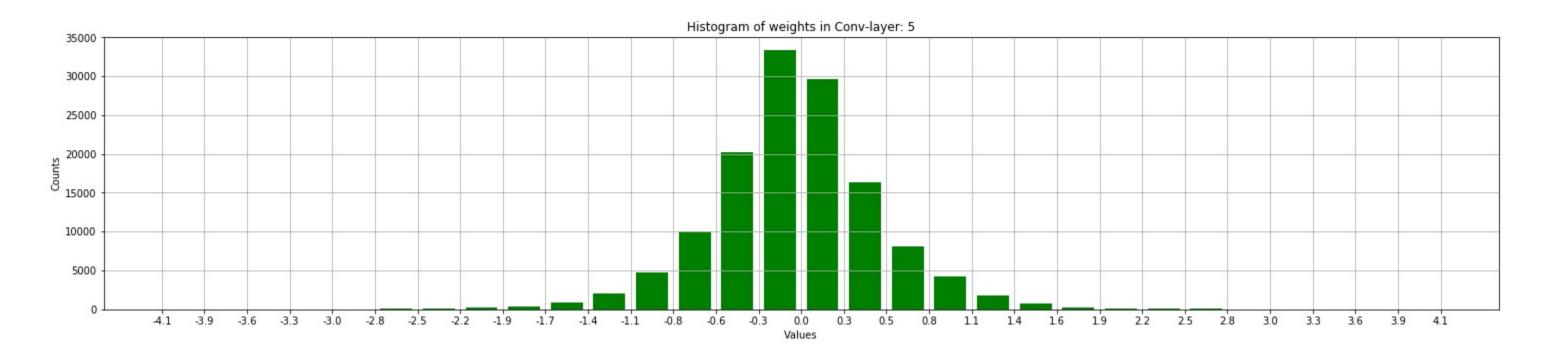




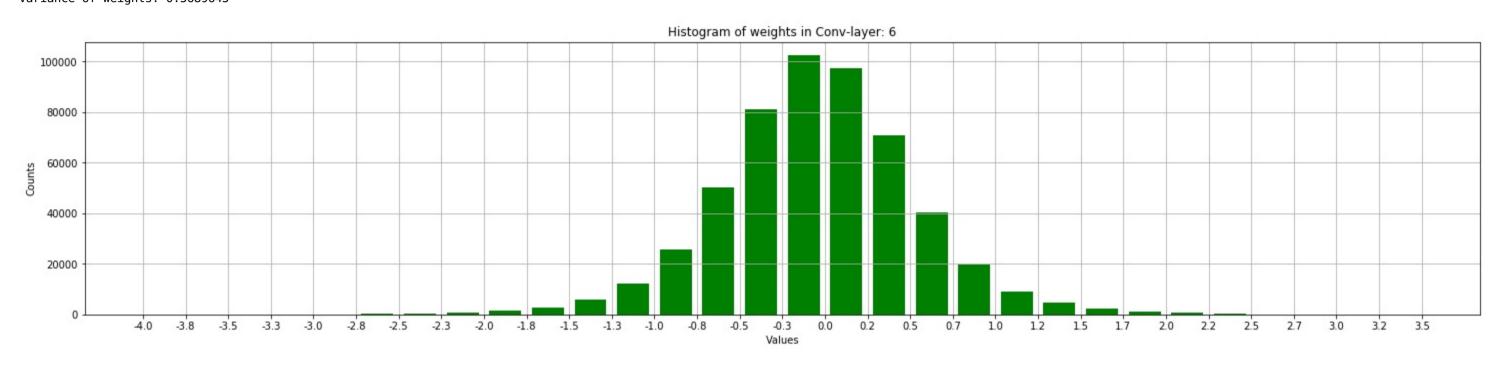


======== Conv. Layer: 4
Minimum of weights: -4.3188796
Maximum of weights: 4.1786413
Average value of weights: 0.0214825
Variance of weights: 0.6290445

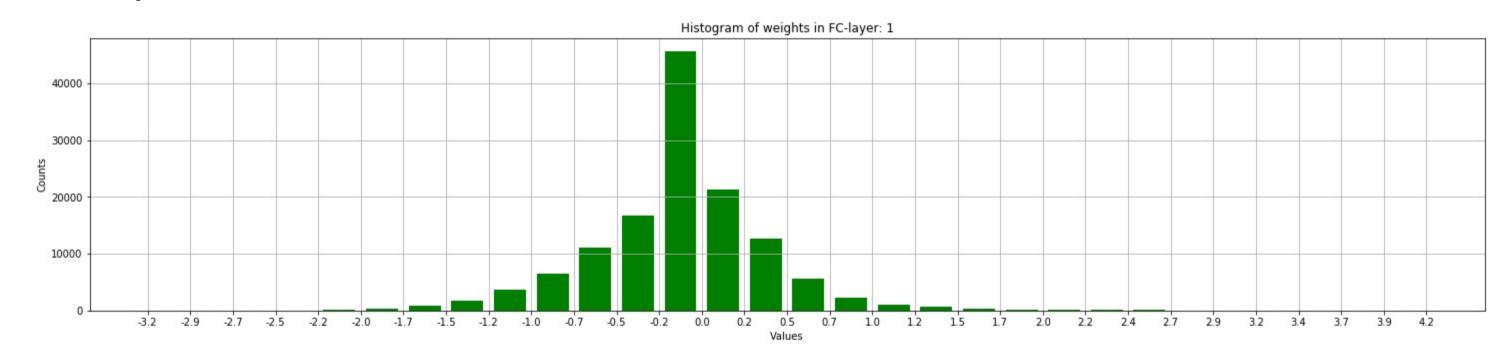




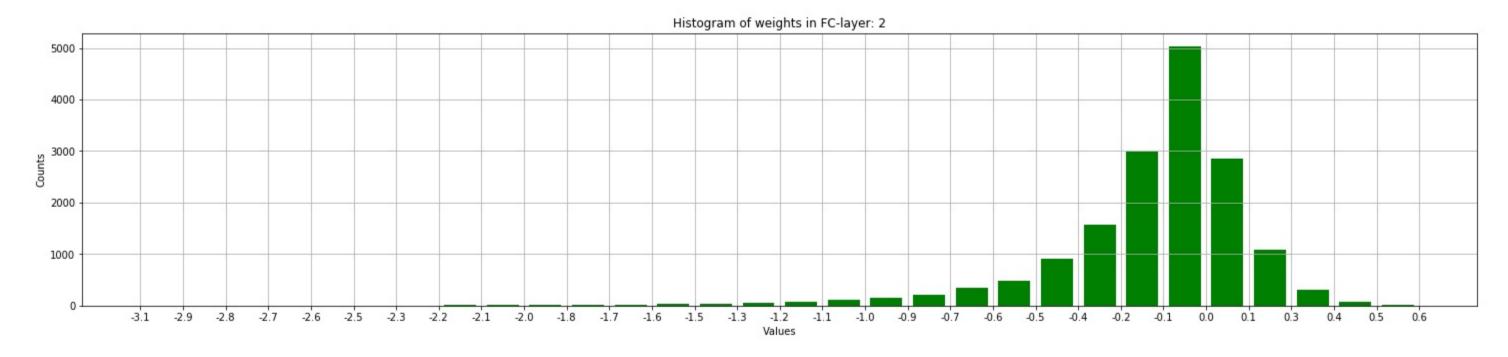
====== Conv. Layer: 6 Minimum of weights: -3.9567604 Maximum of weights: 3.5482860 Average value of weights: 0.0016828 Variance of weights: 0.5689645



======= Fully connected Layer: 1 Minimum of weights: -3.1398053 Maximum of weights: 4.2100883 Average value of weights: -0.0663840 Variance of weights: 0.5053805



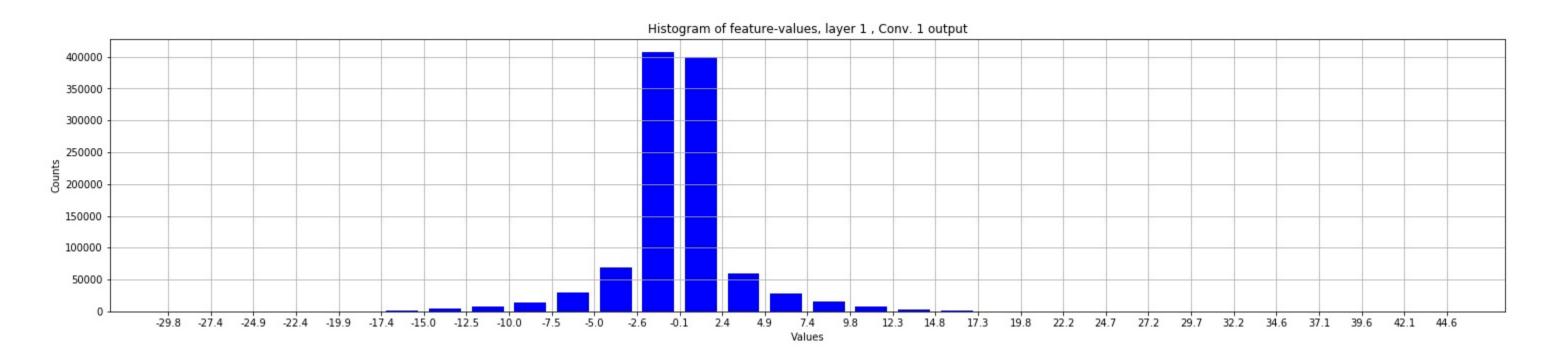
======= Fully connected Layer: 2 Minimum of weights: -3.0179267 Maximum of weights: 0.6770992 Average value of weights: -0.0958171 Variance of weights: 0.2994871



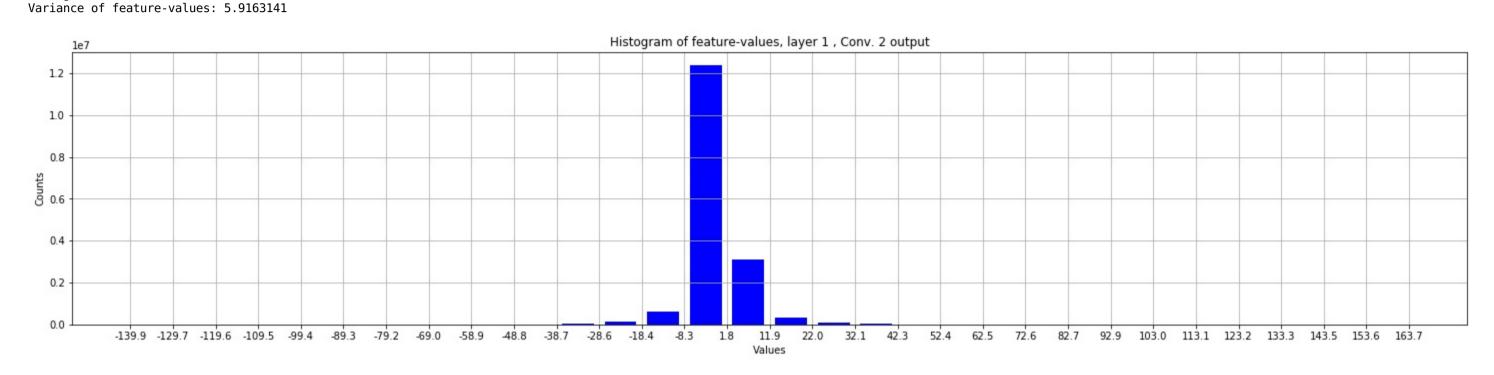
\_\_\_\_\_\_ Statistical analysis of the intermediate signals:

======= Feature-values, layer 1 ,Conv. 1 output:

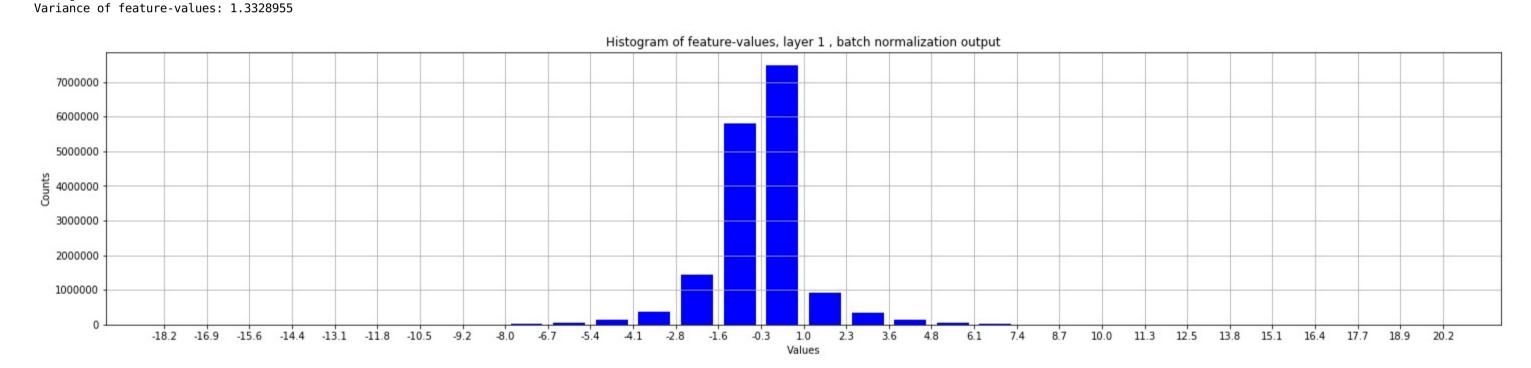
Minimum of feature-values: -29.7912560 Maximum of feature-values: 44.6093559 Average value of feature-values: -0.1026003 Variance of feature-values: 3.1774480

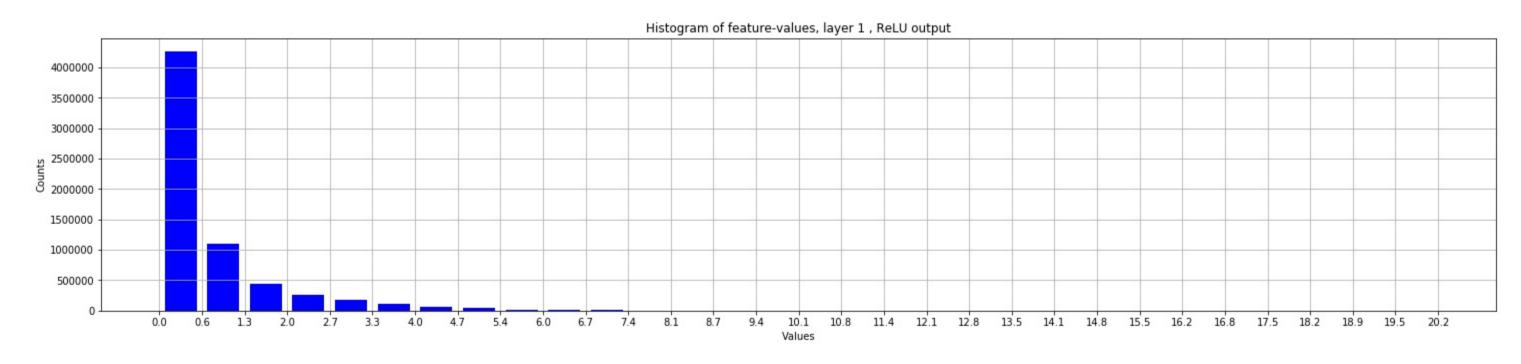


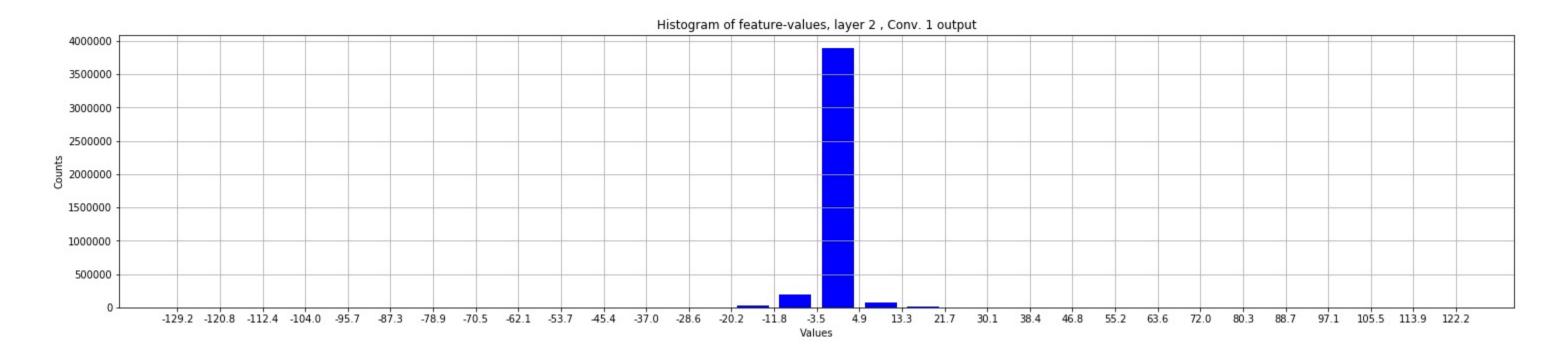
========= Feature-values, layer 1 ,Conv. 2 output: Minimum of feature-values: -139.8166656 Maximum of feature-values: 163.7385101 Average value of feature-values: -0.0175379



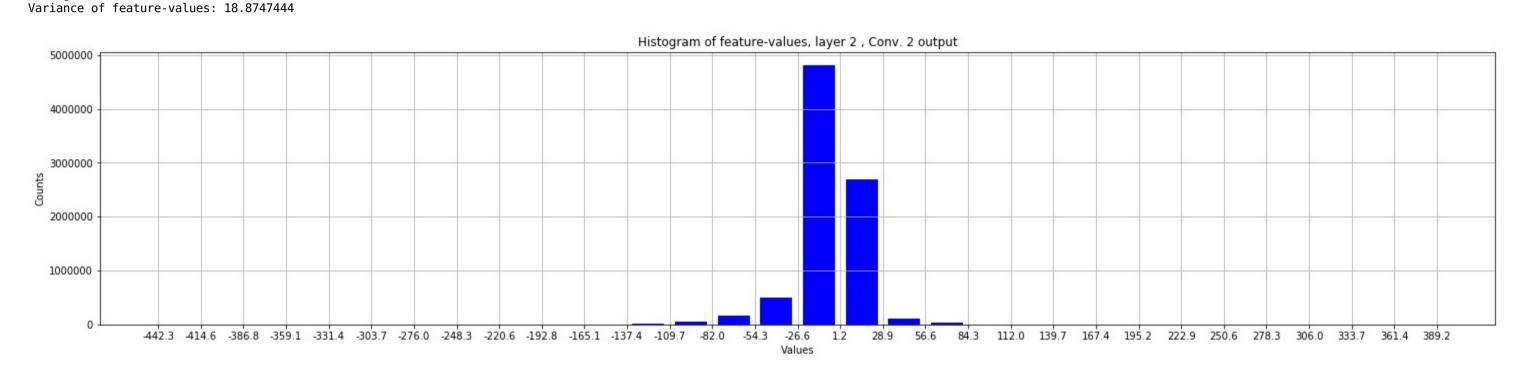
======== Feature-values, layer 1 ,batch normalization output: Minimum of feature-values: -18.1598396
Maximum of feature-values: 20.2694416
Average value of feature-values: -0.2499660



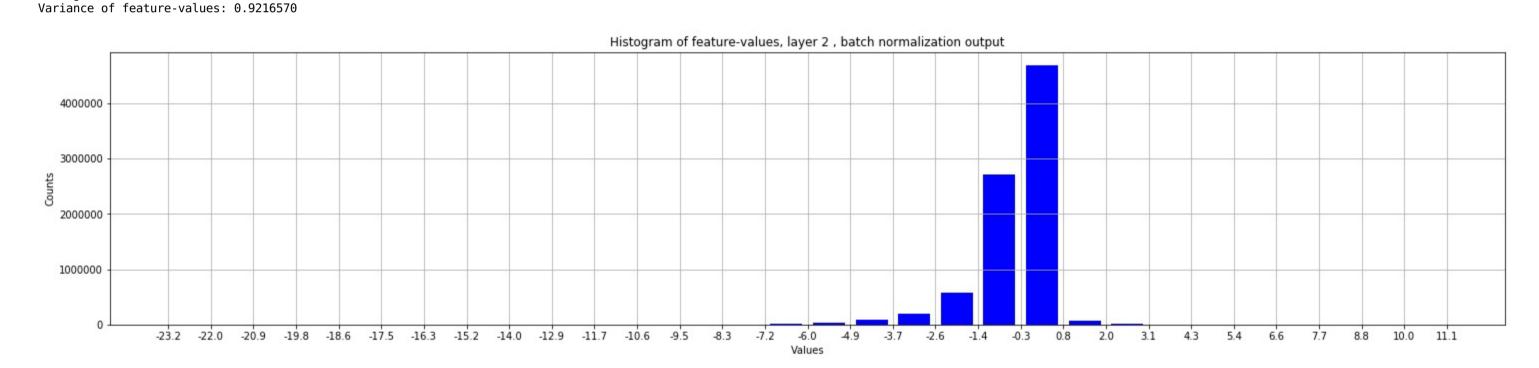


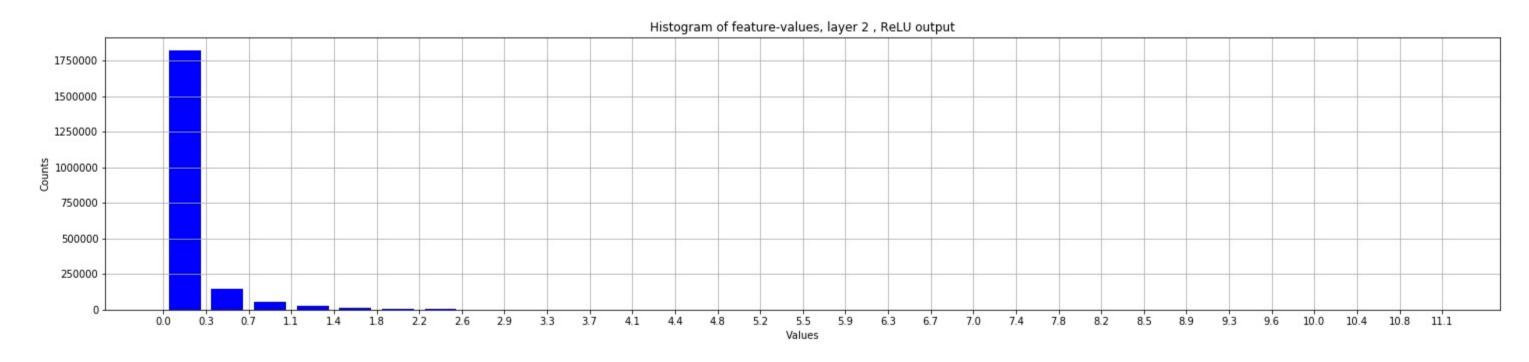


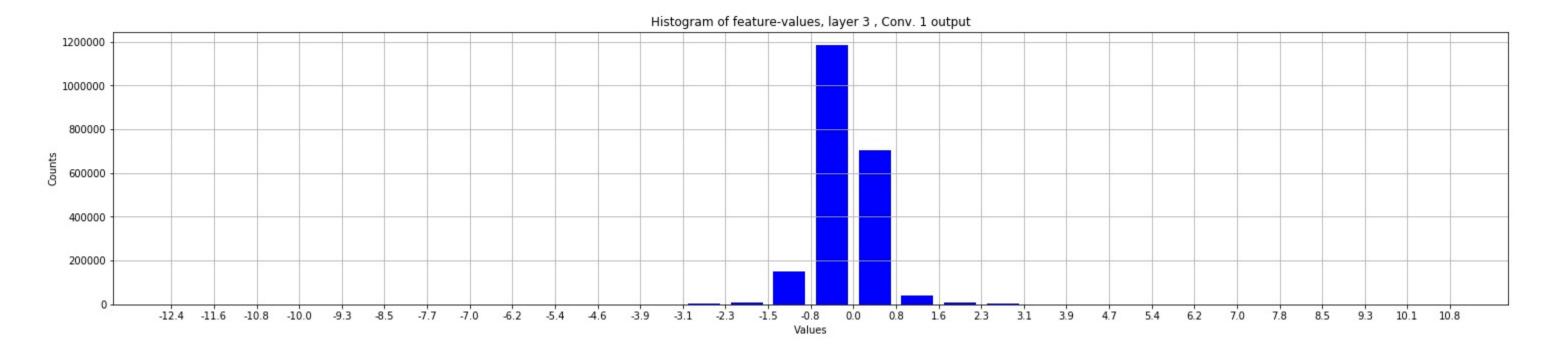
========== Feature-values, layer 2 ,Conv. 2 output: Minimum of feature-values: -442.2140808
Maximum of feature-values: 389.2077942
Average value of feature-values: -5.1846013



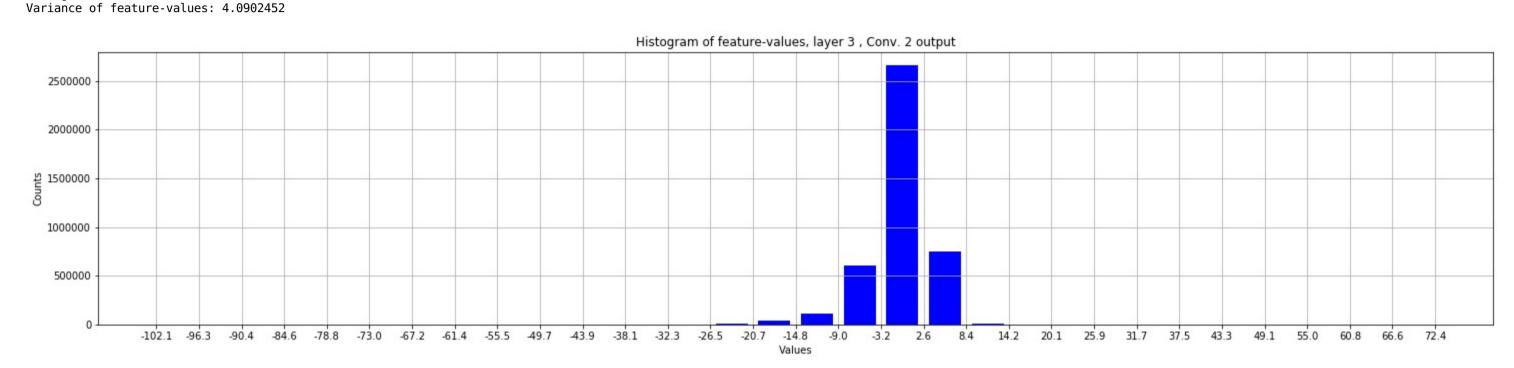
========= Feature-values, layer 2 ,batch normalization output: Minimum of feature-values: -23.1320820 Maximum of feature-values: 11.1865282 Average value of feature-values: -0.4765485



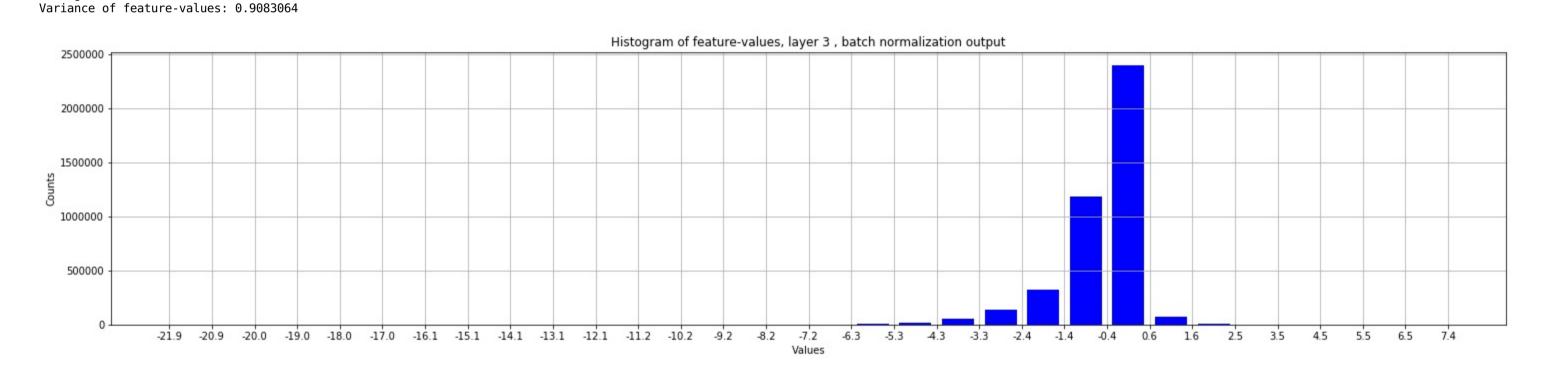




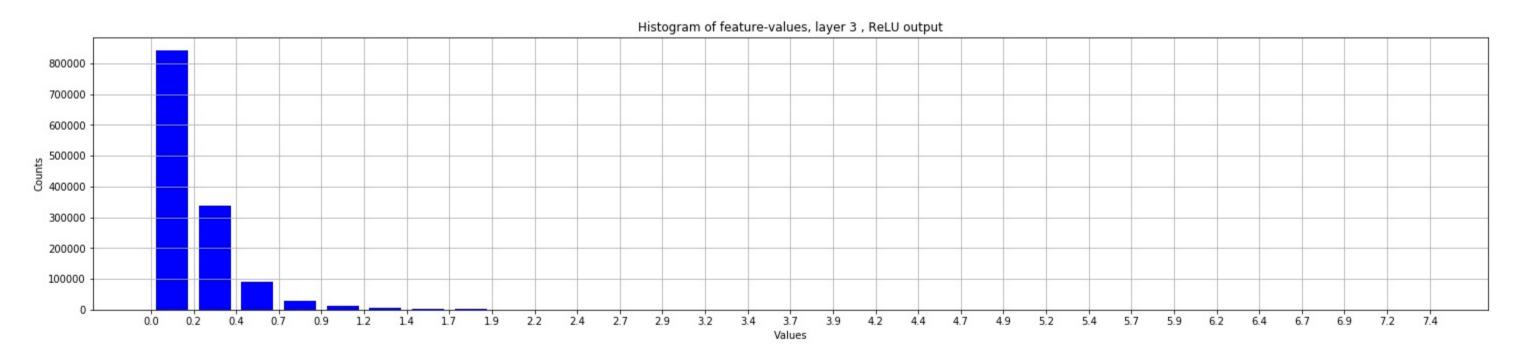
========== Feature-values, layer 3 ,Conv. 2 output: Minimum of feature-values: -102.0276871
Maximum of feature-values: 72.4552383
Average value of feature-values: -0.4983881

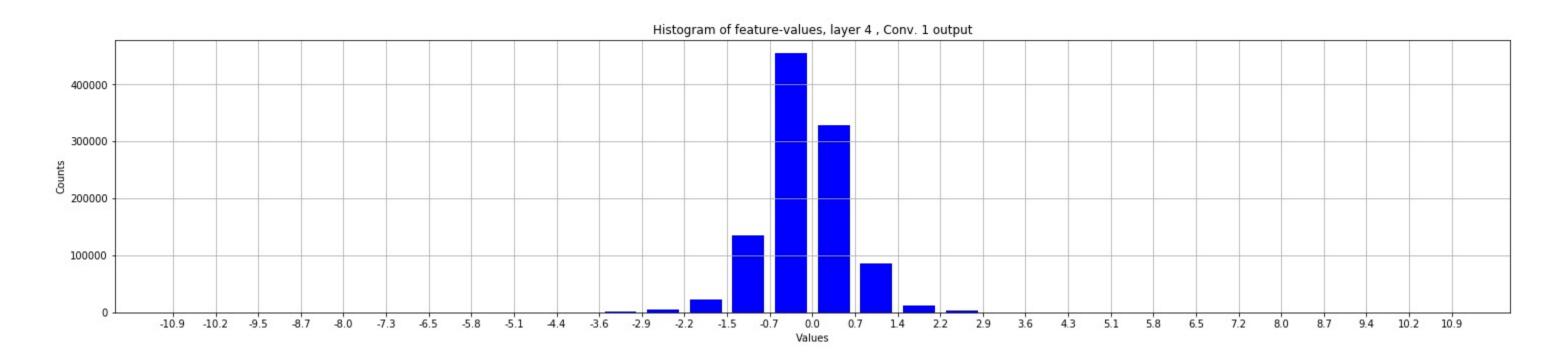


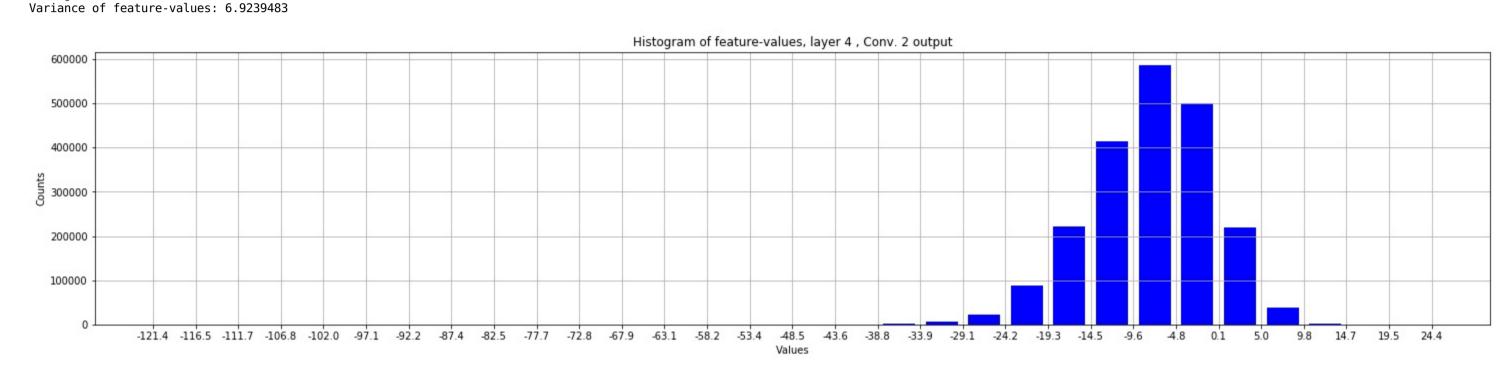
========= Feature-values, layer 3 ,batch normalization output: Minimum of feature-values: -21.8711357
Maximum of feature-values: 7.4836397
Average value of feature-values: -0.4672673



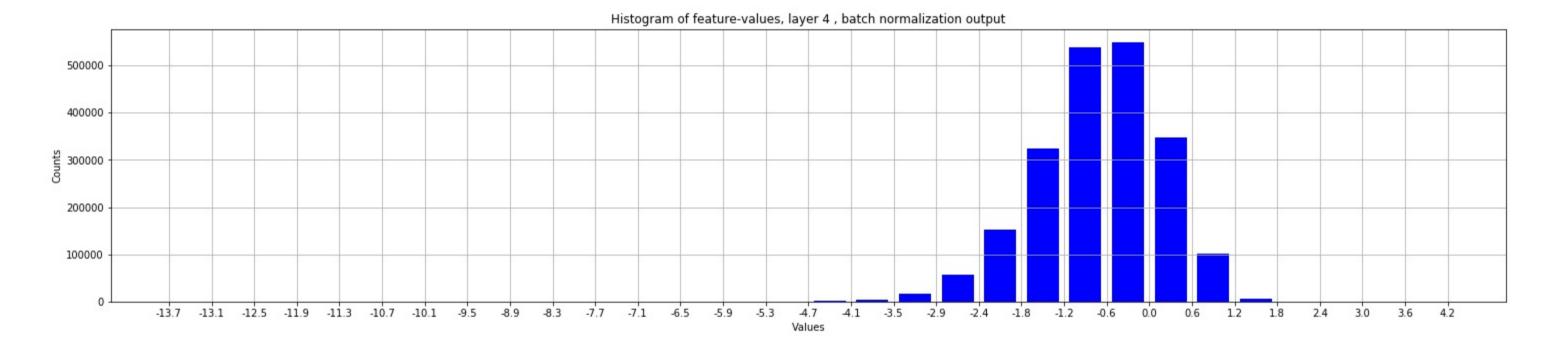
========= Feature-values, layer 3 ,ReLU output: Minimum of feature-values: 0.0000005 Maximum of feature-values: 7.4836397 Average value of feature-values: 0.2437660 Variance of feature-values: 0.2365261

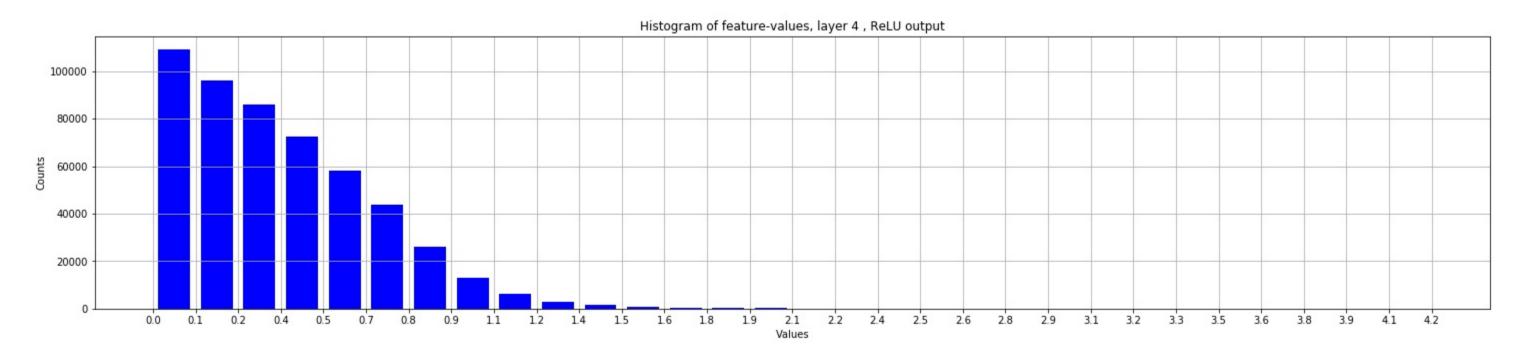




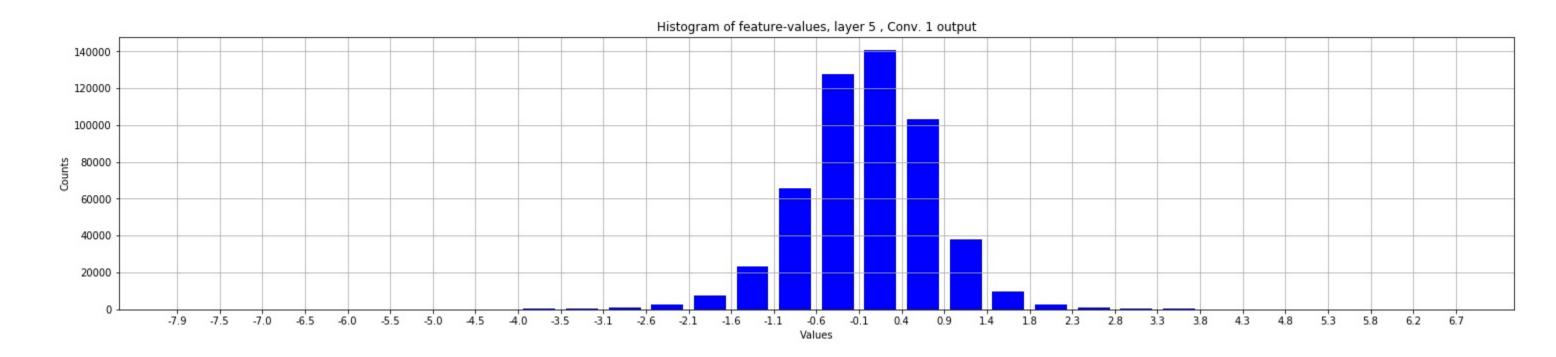


============ Feature-values, layer 4 ,batch normalization output:
Minimum of feature-values: -13.6144218
Maximum of feature-values: 4.2477922
Average value of feature-values: -0.6210883
Variance of feature-values: 0.8689694

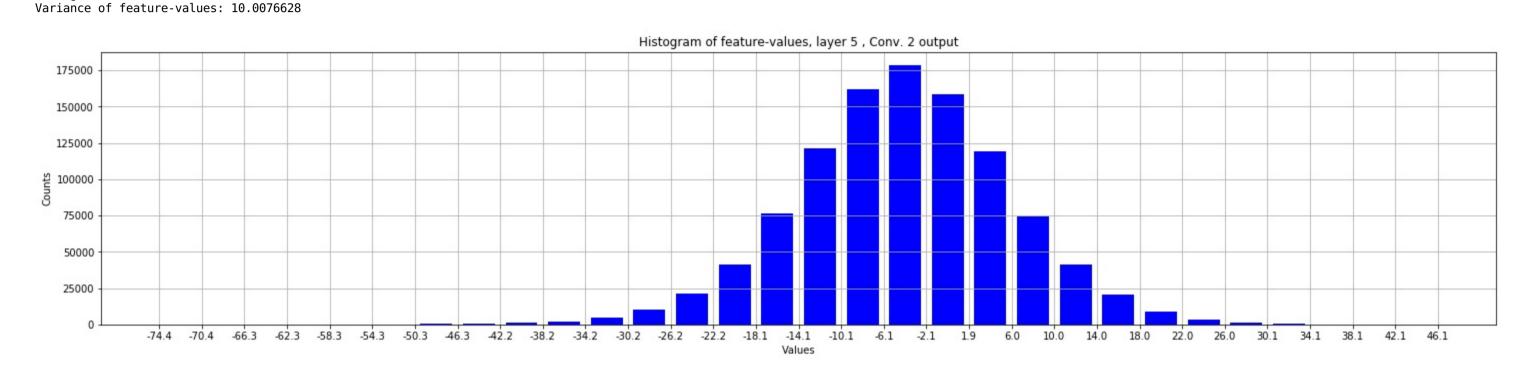




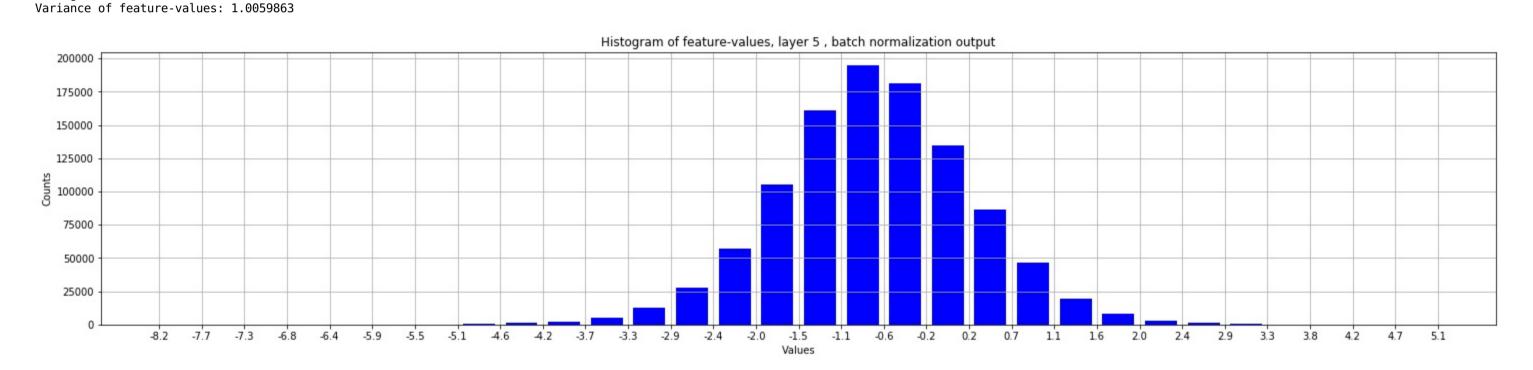
=========== Feature-values, layer 5 ,Conv. 1 output: Minimum of feature-values: -7.8931351 Maximum of feature-values: 6.7803674 Average value of feature-values: 0.0313304 Variance of feature-values: 0.7493501

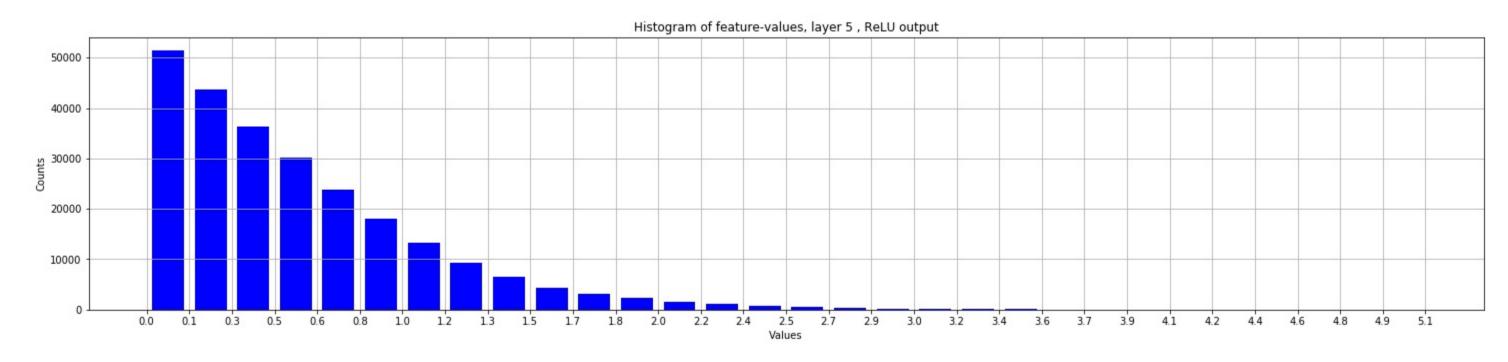


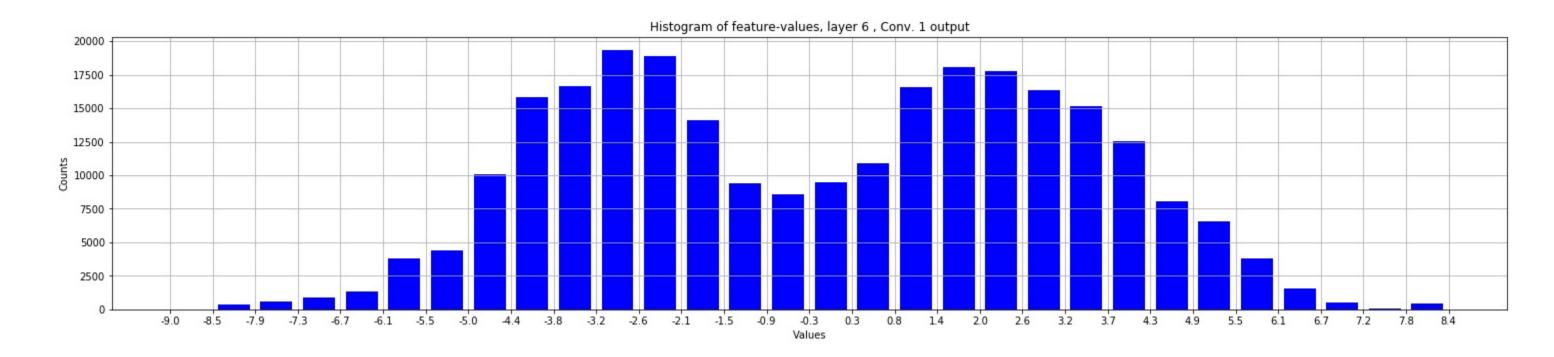
========== Feature-values, layer 5 ,Conv. 2 output: Minimum of feature-values: -74.3217239
Maximum of feature-values: 46.1651535
Average value of feature-values: -4.2845340



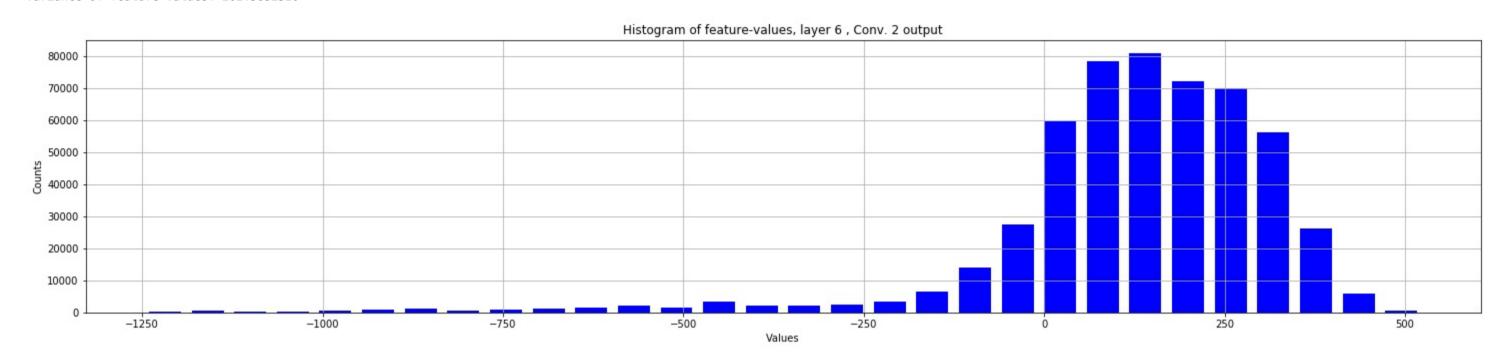
========= Feature-values, layer 5 ,batch normalization output: Minimum of feature-values: -8.1038589
Maximum of feature-values: 5.1482258
Average value of feature-values: -0.6873527

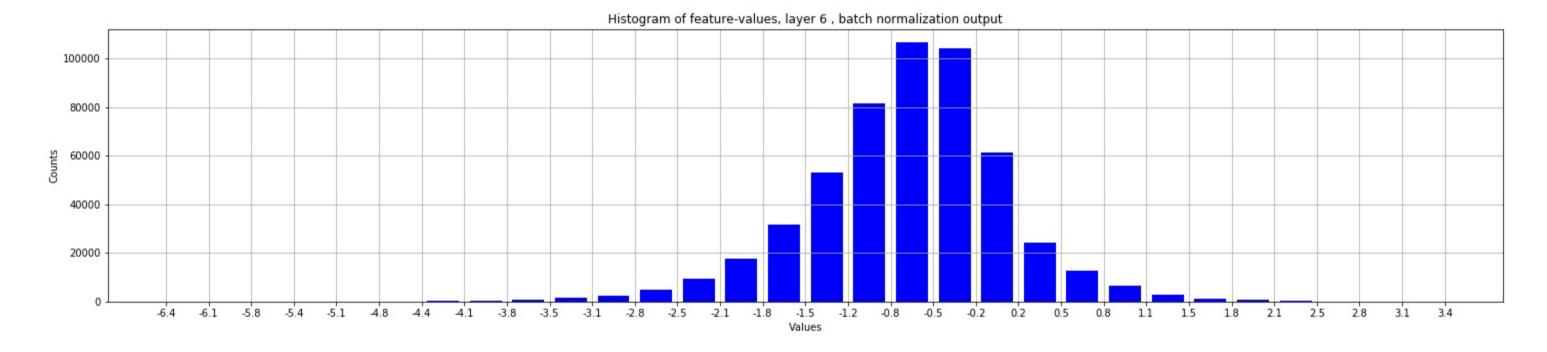


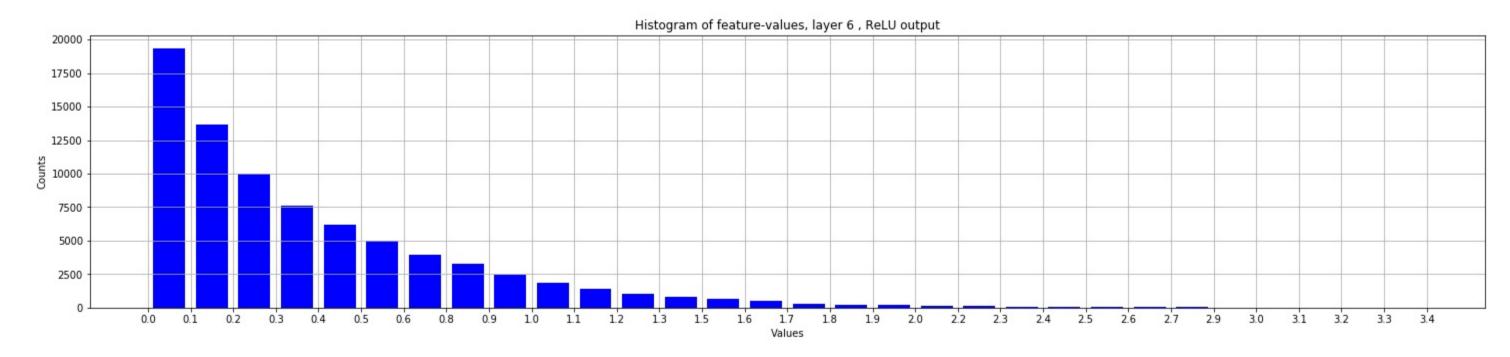


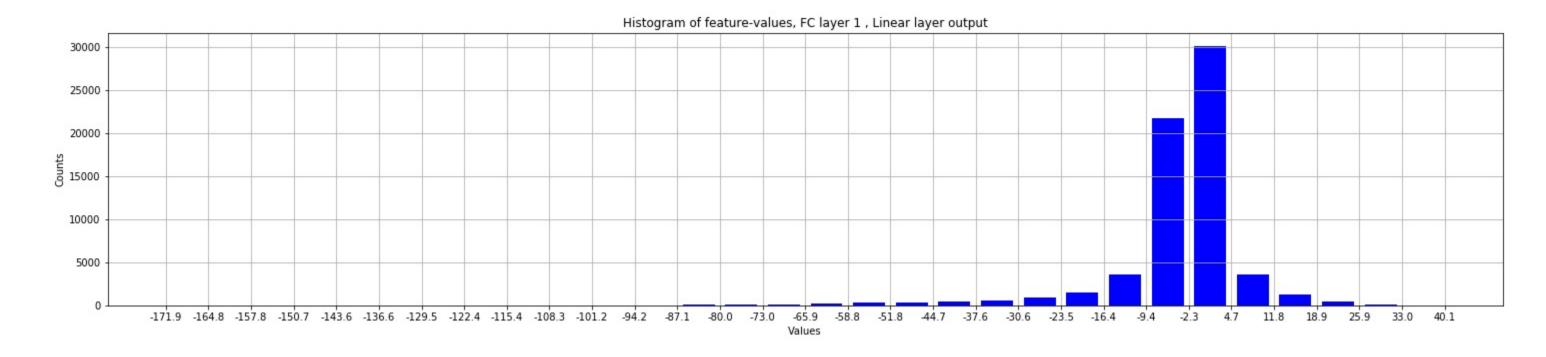


=========== Feature-values, layer 6 ,Conv. 2 output:
Minimum of feature-values: -1248.0252686
Maximum of feature-values: 524.3761597
Average value of feature-values: 123.7501526
Variance of feature-values: 202.5832520

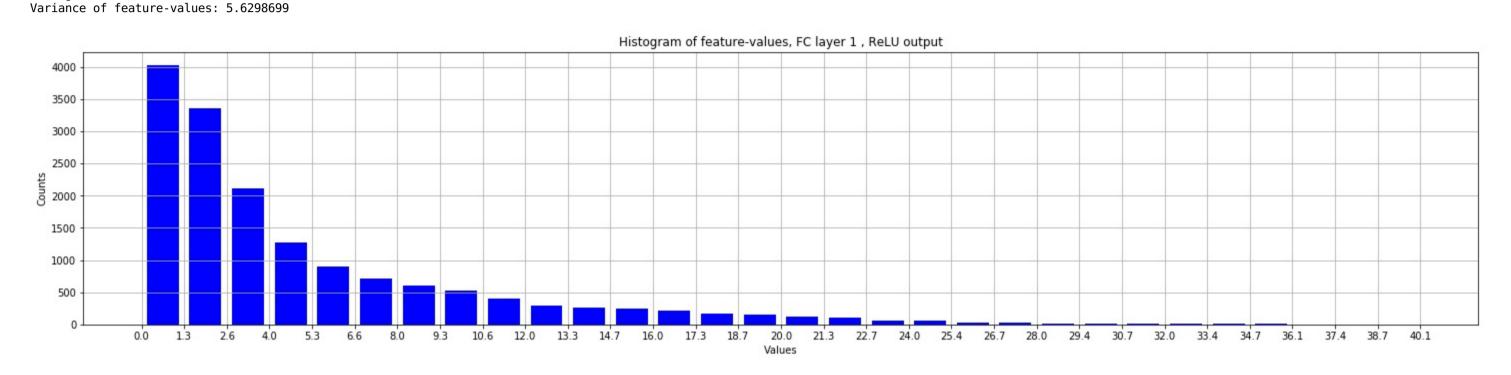




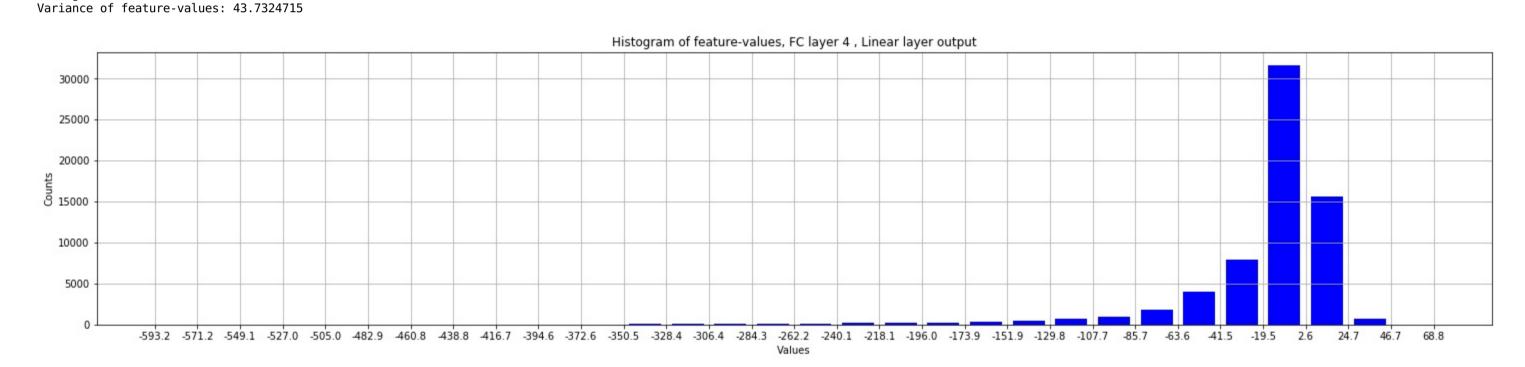




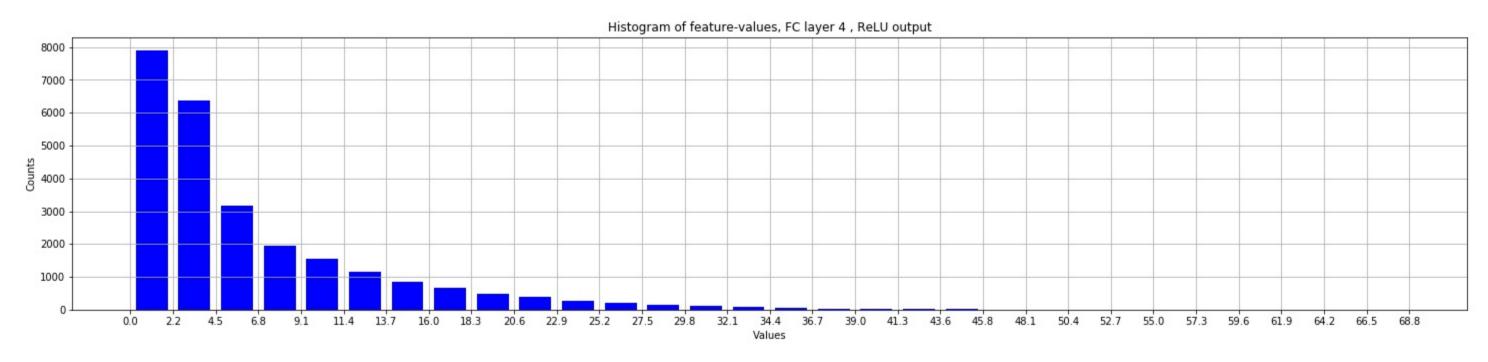
======= Feature-values, FC layer 1 ,ReLU output: Minimum of feature-values: 0.0002006 Maximum of feature-values: 40.1236763 Average value of feature-values: 5.1100726



====== Feature-values, FC layer 4 ,Linear layer output: Minimum of feature-values: -593.1761475 Maximum of feature-values: 68.8441162 Average value of feature-values: -17.6684818



======= Feature-values, FC layer 4 ,ReLU output: Minimum of feature-values: 0.0003807 Maximum of feature-values: 68.8441162 Average value of feature-values: 6.4858184 Variance of feature-values: 7.0906758



## Fix-point quantization to int8:

we performed a fix-point quantization to int8 format.

There is only one de-quantiztion unit in the network in the last layer to get back the output

In each layer a scale factor is determined for each output channel.

For example, a conv-layer with 32 inputs and 64 outputs has 64 different scale-factors.

Accuracy after using training-aware fixpoint quantization:

TP: 99.02 FP: 6.03 AF\_threshold = 3 

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Replacing ReLU activation functions with Sigmoid functions (floating point):

Experiment 1:
All functions in convolution and the fully connected layers are replaced by sigmoids:
TP: 90.84 FP: 25.76 AF\_threshold = 3

Experiment 2:
Only functions in the fully connected layers are replaced by sigmoids:
TP: 99.02 FP: 3.14 AF\_threshold = 3

In [100]: