

```
In [1]: import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
```

```
In [5]: filter0, filter1, weight0, bias0, loss_training_set ,loss_test_set = CNN()
```

0.0 % proceeding

```
-----[[ 4.94433685  0.04137146]
[ -5.78303531 -25.75520418]] [[ 1.14944205  2.44013671 -4.59932328]
[ 20.24954362 -0.62890586 -13.38398682]
[ 6.79403157 -18.03979232 -0.58239126]]
```

training loss is 8.003639884524588

test loss is 4.977614911653973

0.642 accuracy!

2.0 % proceeding

```
-----[[ 4.93684997 -0.21642669]
[ -6.15443437 -26.96640717]] [[ 1.15607207  2.44929106 -4.62500252]
[ 20.21473097 -0.80585084 -14.02510843]
[ 6.72239544 -18.07591874 -0.46975341]]
```

training loss is 4.193071658468663

test loss is 3.4713092306728717

0.714 accuracy!

4.0 % proceeding

```
-----[[ 4.9336302 -0.36068088]
[ -6.3902654 -27.72466848]] [[ 1.15270744  2.45318333 -4.65398408]
[ 20.1919348 -0.91681943 -14.3898473 ]
[ 6.66505236 -18.11666224 -0.41988831]]
```

training loss is 3.259235034314103

test loss is 2.9390850867588516

0.722 accuracy!

6.0 % proceeding

```
-----[[ 4.9317505 -0.44960596]
[ -6.57794075 -28.29133791]] [[ 1.14629541  2.45502173 -4.68135756]
[ 20.17135488 -0.99909505 -14.62594272]
[ 6.61422686 -18.16696102 -0.40796971]]
```

training loss is 2.64117681184015

test loss is 2.3579486998365446

0.716 accuracy!

8.0 % proceeding

```
-----[[ 4.93028928 -0.50689941]
[ -6.73931053 -28.72524031]] [[ 1.13955834  2.45577622 -4.70437314]
[ 20.15119189 -1.06366314 -14.78655269]
[ 6.57178147 -18.21692109 -0.41262023]]
```

training loss is 2.0838387640238

test loss is 1.8596418309413887

0.719 accuracy!

10.0 % proceeding

```
-----[[ 4.92943522 -0.5489492 ]
[ -6.88308688 -29.08118041]] [[ 1.13325155 2.45605938 -4.72294079]
[ 20.13226668 -1.11588044 -14.89953569]
[ 6.53731321 -18.26421116 -0.42348227]]
```

training loss is 1.6468740984536188

test loss is 1.5102513735304761

0.7010000000000001 accuracy!

12.0 % proceeding

```
-----[[ 4.92884221 -0.58114836]
[ -7.00183854 -29.36588167]] [[ 1.12794265 2.45615451 -4.73658054]
[ 20.11672307 -1.15689666 -14.98497325]
[ 6.51120106 -18.30297667 -0.43269181]]
```

training loss is 1.4266864493695475

test loss is 1.3674599870524156

0.706 accuracy!

14.000000000000002 % proceeding

```
-----[[ 4.92833892 -0.60477631]
[ -7.09803042 -29.58629074]] [[ 1.12410677 2.45627961 -4.7464686 ]
[ 20.10461932 -1.18926877 -15.05392182]
[ 6.49100745 -18.33201077 -0.43783459]]
```

training loss is 1.322407102200045

test loss is 1.2953234076302347

0.7030000000000001 accuracy!

16.0 % proceeding

```
-----[[ 4.92795552 -0.62289366]
[ -7.17466163 -29.75377004]] [[ 1.12148263 2.45645794 -4.7538237 ]
[ 20.09563764 -1.2144128 -15.11087499]
[ 6.47543744 -18.35258975 -0.43942606]]
```

training loss is 1.2519212019936354

test loss is 1.2442857786734063

0.698 accuracy!

18.0 % proceeding

```
-----[[ 4.92771905 -0.63537731]
[ -7.23206892 -29.87518344]] [[ 1.11988992 2.45667964 -4.75939616]
[ 20.08923628 -1.23354604 -15.15814141]
[ 6.46347837 -18.36658737 -0.43868308]]
```

training loss is 1.1907586210894245

test loss is 1.201915254324404

0.696 accuracy!

20.0 % proceeding

```
-----[[ 4.92759511 -0.64352256]
[ -7.27454142 -29.96133808]] [[ 1.11890038 2.45693139 -4.76385701]
```

[20.08496302 -1.24777013 -15.19766777]
[6.45438257 -18.37560738 -0.43646998]]

training loss is 1.1421427446682928
test loss is 1.138595704890959
0.696 accuracy!
22.0 % proceeding

-----[[4.92753173 -0.64844036]
[-7.30584047 -30.02190477]] [[1.11831927 2.45720628 -4.7674215]
[20.08235754 -1.25810725 -15.23096281]
[6.44771015 -18.38117549 -0.43344535]]

training loss is 1.094222806014637
test loss is 1.0696523843354897
0.6950000000000001 accuracy!
24.0 % proceeding

-----[[4.92746724 -0.65137162]
[-7.33148833 -30.06875864]] [[1.11801516 2.45749674 -4.77017661]
[20.08084829 -1.26584741 -15.25974321]
[6.44256005 -18.38510421 -0.4303051]]

training loss is 1.0599526016115133
test loss is 1.0299392908220248
0.696 accuracy!
26.0 % proceeding

-----[[4.92740098 -0.65329187]
[-7.3543042 -30.10871716]] [[1.11788885 2.45778303 -4.77243328]
[20.08017223 -1.27183347 -15.28515997]
[6.43825001 -18.38829971 -0.42731999]]

training loss is 1.034899674288894
test loss is 1.0035842504328654
0.6930000000000001 accuracy!
28.000000000000004 % proceeding

-----[[4.92730709 -0.65448013]
[-7.37493166 -30.14351266]] [[1.11789546 2.45807316 -4.77439301]
[20.08024187 -1.27627005 -15.30771382]
[6.43464993 -18.39095417 -0.42449837]]

training loss is 1.0153014903815953
test loss is 0.9839442360806112
0.692 accuracy!
30.0 % proceeding

-----[[4.92721836 -0.65509132]
[-7.39338491 -30.17381712]] [[1.11800245 2.45837161 -4.77612456]
[20.08097342 -1.27934198 -15.32788026]
[6.43166415 -18.39321686 -0.42183117]]

training loss is 0.9995644016097224
test loss is 0.96787691095104
0.6930000000000001 accuracy!
32.0 % proceeding

-----[[4.92714124 -0.65523647]
[-7.4094518 -30.19980622]] [[1.118202 2.4586812 -4.77761257]
[20.08228804 -1.28123195 -15.34602835]
[6.42916196 -18.39516137 -0.4193051]]

training loss is 0.9861675937462557
test loss is 0.9548874697059625
0.694 accuracy!
34.0 % proceeding

-----[[4.92708465 -0.65524522]
[-7.42357827 -30.22256489]] [[1.11847021 2.45899599 -4.77887335]
[20.08406413 -1.28222708 -15.36254322]
[6.42700253 -18.39680355 -0.41682647]]

training loss is 0.9758015912864914
test loss is 0.9445409064520675
0.696 accuracy!
36.0 % proceeding

-----[[4.92704532 -0.65521529]
[-7.43626242 -30.24290204]] [[1.11880529 2.45931395 -4.77991722]
[20.08617067 -1.28254195 -15.37778595]
[6.42508462 -18.39817975 -0.41428568]]

training loss is 0.9683697058295507
test loss is 0.9372986468638074
0.698 accuracy!
38.0 % proceeding

-----[[4.9270116 -0.65524545]
[-7.44773007 -30.26145761]] [[1.11918047 2.4596341 -4.7807745]
[20.08844397 -1.28244145 -15.39211979]
[6.42332073 -18.39927492 -0.4115963]]

training loss is 0.9634715107767409
test loss is 0.9332774227010151
0.7 accuracy!
40.0 % proceeding

-----[[4.92697539 -0.65530813]
[-7.45828086 -30.27856401]] [[1.11959776 2.45995597 -4.78147221]
[20.09078629 -1.2821072 -15.40579307]
[6.42164694 -18.40006743 -0.40872252]]

training loss is 0.9598591283259228
test loss is 0.9305213644658691
0.7 accuracy!
42.0 % proceeding

-----[[4.92694133 -0.65539777]
[-7.46776148 -30.2940585]] [[1.12003999 2.4602781 -4.78202588]
[20.09316633 -1.28161556 -15.41891991]

[6.42004229 -18.40053208 -0.40569546]]

training loss is 0.9566553516970187
test loss is 0.9278036063094377
0.698 accuracy!
44.0 % proceeding

-----[[4.92691073 -0.65543507]
[-7.47614734 -30.30778539]] [[1.1205205 2.46060131 -4.78244715]
[20.09556103 -1.28100277 -15.43156555]
[6.41849683 -18.40066848 -0.40255059]]

training loss is 0.9536371499691187
test loss is 0.9246432829244149
0.698 accuracy!
46.0 % proceeding

-----[[4.92688443 -0.65540488]
[-7.48354289 -30.31990461]] [[1.12102013 2.46092378 -4.7827542]
[20.09797515 -1.28027848 -15.44375598]
[6.41700182 -18.40050153 -0.39929011]]

training loss is 0.9507608462865061
test loss is 0.9214789877739745
0.698 accuracy!
48.0 % proceeding

-----[[4.92686692 -0.65522483]
[-7.48997641 -30.33028651]] [[1.12154297 2.46124793 -4.78296943]
[20.10039025 -1.27946976 -15.45552432]
[6.41555862 -18.40003579 -0.39592429]]

training loss is 0.947963954178294
test loss is 0.9185885552602256
0.698 accuracy!
50.0 % proceeding

-----[[4.92685551 -0.65486026]
[-7.49553239 -30.33899852]] [[1.12209162 2.46157225 -4.78309946]
[20.10282361 -1.27857958 -15.46688731]
[6.41414886 -18.39928851 -0.39244939]]

training loss is 0.9451830208238162
test loss is 0.9159398742210301
0.698 accuracy!
52.0 % proceeding

-----[[4.92684746 -0.65443559]
[-7.50037801 -30.34640034]] [[1.1226573 2.46189708 -4.78314471]
[20.10527637 -1.27760789 -15.47786968]
[6.41277391 -18.39828047 -0.38886598]]

training loss is 0.9424001941921551
test loss is 0.9131980065899373
0.6990000000000001 accuracy!
54.0 % proceeding

-----[[4.92683983 -0.65395753]
[-7.50455563 -30.35256773]] [[1.12323691 2.46222262 -4.78310881]
[20.10774449 -1.27656301 -15.48850185]
[6.41143233 -18.39702976 -0.38519048]]

training loss is 0.9396220442041564
test loss is 0.9100343070626784
0.698 accuracy!
56.00000000000001 % proceeding

-----[[4.92683394 -0.65337671]
[-7.50797032 -30.35739234]] [[1.12383158 2.46254887 -4.78300215]
[20.11021765 -1.27545881 -15.49882036]
[6.41011953 -18.3955435 -0.38144003]]

training loss is 0.9369684938149693
test loss is 0.9069711970736034
0.698 accuracy!
57.99999999999999 % proceeding

-----[[4.92682935 -0.65276623]
[-7.51067996 -30.36105614]] [[1.12443465 2.46287597 -4.78282169]
[20.1126955 -1.2742957 -15.50885616]
[6.40883184 -18.39384757 -0.37762795]]

training loss is 0.9345568179854897
test loss is 0.9044848982659157
0.698 accuracy!
60.0 % proceeding

-----[[4.92682494 -0.65211192]
[-7.51268182 -30.36374039]] [[1.12502827 2.46320122 -4.78257702]
[20.11518734 -1.27308332 -15.5186323]
[6.40757106 -18.39197568 -0.37375981]]

training loss is 0.9324194894843837
test loss is 0.9026605446228219
0.6990000000000001 accuracy!
62.0 % proceeding

-----[[4.92681953 -0.65144241]
[-7.51406821 -30.3654983]] [[1.12562803 2.46352658 -4.78227349]
[20.11768861 -1.27181433 -15.52818171]
[6.40633316 -18.38992933 -0.3698334]]

training loss is 0.9304513499158469
test loss is 0.9013235268462263
0.698 accuracy!
64.0 % proceeding

-----[[4.92681266 -0.65077447]
[-7.51491667 -30.36653615]] [[1.12623366 2.46385108 -4.78193536]
[20.12019814 -1.2705039 -15.53752393]
[6.40510781 -18.38774443 -0.36584528]]

training loss is 0.92843516252572
test loss is 0.8997219831284374
0.698 accuracy!
66.0 % proceeding

-----[[4.92680298 -0.65013057]
[-7.51534196 -30.36704989]] [[1.12685474 2.46417567 -4.78155735]
[20.12271585 -1.26915233 -15.54667643]
[6.40388797 -18.38545242 -0.36180589]]

training loss is 0.9261793869489717
test loss is 0.8969607648276419
0.698 accuracy!
68.0 % proceeding

-----[[4.92679028 -0.64956034]
[-7.51544559 -30.3673123]] [[1.1274812 2.46449986 -4.78114822]
[20.12523653 -1.26776844 -15.55565455]
[6.40267229 -18.38308812 -0.35772371]]

training loss is 0.9238745223602366
test loss is 0.8940413634363094
0.698 accuracy!
70.0 % proceeding

-----[[4.92677329 -0.64905396]
[-7.51530201 -30.36742495]] [[1.12810524 2.46482643 -4.78072213]
[20.12776601 -1.26635175 -15.56446592]
[6.40145546 -18.38067719 -0.35361181]]

training loss is 0.9217560603579604
test loss is 0.8914994411696999
0.6990000000000001 accuracy!
72.0 % proceeding

-----[[4.92675478 -0.64862811]
[-7.51496879 -30.3674962]] [[1.12871865 2.46515231 -4.78028377]
[20.13030659 -1.26490818 -15.5731381]
[6.40024015 -18.37823304 -0.34947492]]

training loss is 0.9200946032241998
test loss is 0.889507535178336
0.698 accuracy!
74.0 % proceeding

-----[[4.92673615 -0.64823156]
[-7.51446659 -30.36746793]] [[1.1293422 2.46547825 -4.77984056]
[20.13284779 -1.26344707 -15.58172172]
[6.3990291 -18.37575452 -0.34531119]]

training loss is 0.9191120575721087
test loss is 0.8881745827496368
0.6990000000000001 accuracy!
76.0 % proceeding

-----[[4.92671583 -0.64785052]
[-7.51386834 -30.36729741]] [[1.12996499 2.46580405 -4.77939958]
[20.13539133 -1.26198234 -15.59025496]
[6.39782658 -18.37324502 -0.34112845]]

training loss is 0.9186100023744963
test loss is 0.8874170930956741
0.7010000000000001 accuracy!
78.0 % proceeding

-----[[4.92669404 -0.64747883]
[-7.51325854 -30.36706785]] [[1.13058624 2.46612904 -4.7789559]
[20.13793171 -1.26051773 -15.59876531]
[6.39662956 -18.37071841 -0.3369374]]

training loss is 0.918296921520483
test loss is 0.8869937655933281
0.7 accuracy!
80.0 % proceeding

-----[[4.92667106 -0.64709982]
[-7.51261366 -30.36679497]] [[1.13120538 2.46645472 -4.77851332]
[20.14046906 -1.25904579 -15.60725707]
[6.39544178 -18.36818358 -0.33274619]]

training loss is 0.9180477543228991
test loss is 0.8867354459624197
0.7 accuracy!
82.0 % proceeding

-----[[4.92664559 -0.64674209]
[-7.51201094 -30.36658117]] [[1.13182395 2.46678106 -4.77807312]
[20.14300609 -1.25758112 -15.61573363]
[6.39425713 -18.36565121 -0.32855625]]

training loss is 0.9178260087382377
test loss is 0.8865611339067304
0.7 accuracy!
84.0 % proceeding

-----[[4.92662043 -0.64638643]
[-7.51143287 -30.3663536]] [[1.13243504 2.46710632 -4.77763885]
[20.14554041 -1.25612231 -15.62420102]
[6.39307425 -18.3631177 -0.32436834]]

training loss is 0.917619768425272
test loss is 0.8864343844705276
0.7 accuracy!
86.0 % proceeding

-----[[4.92659505 -0.64602769]
[-7.51085786 -30.36616161]] [[1.13304712 2.46743112 -4.77720347]
[20.14807402 -1.25466312 -15.63265953]
[6.39189352 -18.36059251 -0.32018455]]

training loss is 0.9174244834089438
test loss is 0.8863364301097091
0.7 accuracy!
88.0 % proceeding

-----[[4.92656825 -0.64567331]
[-7.51030819 -30.36604972]] [[1.13365936 2.46775591 -4.7767669]
[20.15060694 -1.2532023 -15.64111057]
[6.39071106 -18.35807849 -0.31600926]]

training loss is 0.9172370833697676
test loss is 0.8862558242807257
0.7010000000000001 accuracy!
90.0 % proceeding

-----[[4.92654249 -0.64532773]
[-7.50976083 -30.36593555]] [[1.13426967 2.4680799 -4.77633448]
[20.1531406 -1.25174154 -15.64955381]
[6.38953409 -18.35556117 -0.31183544]]

training loss is 0.9170548194329042
test loss is 0.8861850791968385
0.7010000000000001 accuracy!
92.0 % proceeding

-----[[4.92651965 -0.64497026]
[-7.50921593 -30.36579618]] [[1.13487259 2.46840309 -4.77590746]
[20.15566748 -1.25028113 -15.65799443]
[6.38836915 -18.35304183 -0.30765924]]

training loss is 0.9168779036867174
test loss is 0.8861201429954124
0.7010000000000001 accuracy!
94.0 % proceeding

-----[[4.92649617 -0.64461248]
[-7.50867721 -30.36567335]] [[1.13548439 2.46872592 -4.7754878]
[20.15819096 -1.24882274 -15.66643049]
[6.38721123 -18.35052048 -0.30348423]]

training loss is 0.9167046341711819
test loss is 0.8860580237362455
0.7010000000000001 accuracy!
96.0 % proceeding

-----[[4.92647196 -0.64425364]
[-7.50815519 -30.36557149]] [[1.13608658 2.46904721 -4.77507085]
[20.16071376 -1.24736358 -15.67486324]
[6.38605642 -18.34800279 -0.29931453]]

training loss is 0.9165338581471204
test loss is 0.8859967370932528
0.7010000000000001 accuracy!
98.0 % proceeding

```
-----[[ 4.92644686 -0.64388612]
[ -7.50762122 -30.36547113]] [[ 1.13669327  2.46936908 -4.77465317]
[ 20.16323359 -1.24590789 -15.68329472]
[ 6.38490621 -18.34548464 -0.29514632]]
```

training loss is 0.9163660519809752
test loss is 0.8859356175800358
0.7010000000000001 accuracy!

```
In [10]: confusionmatrix, top3matrix, top3matrix_index = ConfusionMatrix_n_top3(filter0, filter1, weight0, bias0, test_data)
```

Loader

```
In [2]: import gzip
import numpy as np
from pathlib import Path
import math
import random

class Dataloader():
    def __init__(self, path, is_train=True, shuffle=True, batch_size=8):
        path = Path(path)
        imagePath = Path(path/'train-images-idx3-ubyte.gz') if is_train else Path(path/'t10k-images-idx3-ubyte.gz')
        labelPath = Path(path/'train-labels-idx1-ubyte.gz') if is_train else Path(path/'t10k-labels-idx1-ubyte.gz')

        self.batch_size = batch_size
        self.images = self.loadImages(imagePath)
        self.labels = self.loadLabels(labelPath)
        self.index = 0
        self.idx = np.arange(0, self.images.shape[0])
        if shuffle: np.random.shuffle(self.idx) # shuffle images

    def __len__(self):
        n_images, _, _ = self.images.shape
        n_images = math.ceil(n_images / self.batch_size)
        return n_images

    def __iter__(self):
        return datasetIterator(self)

    def __getitem__(self, index):
        image = self.images[self.idx[index * self.batch_size:(index + 1) * self.batch_size]]
        label = self.labels[self.idx[index * self.batch_size:(index + 1) * self.batch_size]]
        image = image/255.0
        return image, label

    def loadImages(self, path):
        with gzip.open(path) as f:
            images = np.frombuffer(f.read(), 'B', offset=16)
            images = images.reshape(-1, 1, 28, 28).astype(np.float32)
            return images

    def loadLabels(self, path):
        with gzip.open(path) as f:
            labels = np.frombuffer(f.read(), 'B', offset=8)
            rows = len(labels)
            cols = labels.max() + 1
            one_hot = np.zeros((rows, cols)).astype(np.uint8)
            one_hot[np.arange(rows), labels] = 1
            one_hot = one_hot.astype(np.float64)
            return one_hot

# for enumerate magic python function returns Iterator
class datasetIterator():
    def __init__(self, dataloader):
        self.index = 0
        self.dataloader = dataloader

    def __next__(self):
        if self.index < len(self.dataloader):
            item = self.dataloader[self.index]
            self.index += 1
            return item
        # end of iteration
        raise StopIteration
```

function

```

In [21]: def ReLU(value):
          return max(0, value)

def converter_ReLU(array):
    return np.array([ReLU(x) for x in array])

def converter_ReLU_2D(array):
    return np.array([converter_ReLU(x) for x in array])

def zeroorone(value):
    if value > 0:
        return 1
    else:
        return 0

def converter_zeroorone_10(array):
    return np.array([zeroorone(x) for x in array.reshape(10)])

def converter_zeroorone_784(array):
    return np.array([zeroorone(x) for x in array.reshape(784)])

def converter_zeroorone(array):
    k = len(array.shape)
    n = 1
    for i in range(0,k):
        n = n*array.shape[i]
        #print(n)
    return np.array([zeroorone(x) for x in array.reshape(n)]).reshape(math.ceil(math.sqrt(n)),math.ceil(math.sqrt(n)))

def SoftMax(z):
    c = np.max(z)
    exp_z = np.exp(z-c)
    sum_exp_z = np.sum(exp_z)
    y = exp_z / sum_exp_z
    return y

def Conv(array, filter):
    filter = np.flipud(np.fliplr(filter))
    sub_matrices = np.lib.stride_tricks.as_strided(array,shape = tuple(np.subtract(array.shape, filter.shape))+filter.shape)
    return np.einsum('ij,klj->kl', filter, sub_matrices)

def Cross_entropy_loss(y_label, y_prediction):
    return -np.sum(y_label*np.log(y_prediction+1e-7))

def Maxpooling(array):
    n = array.shape[0]
    m = math.ceil(n/2)
    MAX_array = np.zeros(m*m).reshape(m,m)
    MAX_array_indexed = np.zeros(n*n).reshape(n,n)
    for i in range(0,m):
        for j in range(0,m):
            new_array = array[2*i:2*i+2,2*j:2*j+2]
            MAX_array[i][j] = new_array.max()
            Index = (np.where(new_array==new_array.max())[0][0],np.where(new_array==new_array.max())[1][0])
            MAX_array_indexed[2*i+Index[0]][2*j+Index[1]] = 1
    return MAX_array,MAX_array_indexed

def rotate_180(m):
    N = len(m)
    ret = [[0] * N for _ in range(N)]
    for r in range(N):
        for c in range(N):
            ret[N-1-r][N-1-c] = m[r][c]
    return ret

```

```

In [4]: def CNN(batchsize=100, epoch=50, testing = 1000):
        #ready for dataset
        learning_rate = batchSize/60000
        iteration = math.ceil(60000/batchsize)
        loss_training_set = []
        loss_test_set = []

        training_data = DataLoader(
            path=".",
            shuffle=True,
            batch_size=batchsize
        )
        test_data = DataLoader(
            path=".",
            shuffle=True,
            is_train = False,
            batch_size = 1
        )

```

```

#initialize function
filter0 = np.random.randn(2,2)*10
filter1 = np.random.randn(3,3)*10
weight0 = np.random.randn(25,10) #수정해
bias0 = np.random.randn(10)
for k in range(0,epoch):
    print(100*(k/epoch), "% proceeding")
    training_loss = 0
    error = 0
    test_loss = 0
    for i in range(0,iteration):
        #forward propagation
        print("-",end='')
        delta_3 = 0
        chain_delta_3 = np.zeros(250).reshape(25,10) #수정해
        chain_delta_2 = np.zeros(9).reshape(3,3)
        chain_delta_1 = np.zeros(4).reshape(2,2)

        for j in range(0, batchsize):
            delta_1_b = 0
            delta_2_b = 0
            delta_3_b = 0
            chain_delta_3_b = np.zeros(250).reshape(25,10) #수정해
            chain_delta_2_b = np.zeros(9).reshape(3,3)
            chain_delta_1_b = np.zeros(4).reshape(2,2)

            y_label = training_data.__getitem__(i)[1][j]
            #layer 0 forward propagation
            layer_0 = training_data.__getitem__(i)[0][j].reshape(28,28) #28,28
            layer_0_conv = Conv(layer_0,filter0) #26*26
            #print(layer_0_conv.shape)
            layer_0_relu = converter_ReLU_2D(layer_0_conv) #26*26
            layer_0_pooling,layer_0_pooling_index = Maxpooling(layer_0_relu) #13*13,26*26
            #layer 1 forward propagation
            layer_1_conv = Conv(layer_0_pooling,filter1) #10*10
            layer_1_relu = converter_ReLU_2D(layer_1_conv) #10*10
            layer_1_pooling,layer_1_pooling_index = Maxpooling(layer_1_relu) #5*5,10*10
            #layer 2 forward propagation
            layer_2 = layer_1_pooling.reshape(25)
            layer_2_weight = np.dot(layer_2, weight0) + bias0
            layer_2_softmax = SoftMax(layer_2_weight)
            training_loss += Cross_entropy_loss(y_label,layer_2_softmax)

            #layer 2 backward propagation
            delta_3_b = ((layer_2_softmax - y_label))/batchsize #batchsize
            chain_delta_3_b = np.dot(layer_2.reshape(25,1), delta_3_b.reshape(1,10))

            #layer 1 backward propagation
            delta_2_b = (((np.dot(weight0,delta_3_b).reshape(5,5)).repeat(2, axis=0)).repeat(2, axis=1))*layer_1_pooling,layer_1_pooling_index)
            chain_delta_2_b = Conv(layer_0_pooling,rotate_180(delta_2_b))
            #print(chain_delta_2_b.shape)

            #layer 0 backward propagation
            delta_1_b = Conv(np.pad(delta_2_b, ((3,3),(3,3)), 'constant', constant_values=0),filter1).reshape(2,2)
            chain_delta_1_b = Conv(layer_0,rotate_180(delta_1_b))

            #update value
            delta_3 += delta_3_b
            chain_delta_3 += chain_delta_3_b
            chain_delta_2 += chain_delta_2_b
            chain_delta_1 += chain_delta_1_b
            weight0 -= (learning_rate * chain_delta_3)
            filter1 -= (learning_rate * chain_delta_2)
            filter0 -= (learning_rate * chain_delta_1)
            bias0 -= delta_3.reshape(10)*learning_rate
            #print(filter1,filter0)
        print(filter0,filter1)
        print("\n")
        print("training loss is",training_loss/60000)
        loss_training_set.append(training_loss/60000)

    for i in range(0,testing):
        y_label = test_data.__getitem__(i)[1]
        #layer 0 forward propagation
        layer_0 = test_data.__getitem__(i)[0].reshape(28,28) #28,28
        layer_0_conv = Conv(layer_0,filter0) #26*26
        #print(layer_0_conv.shape)
        layer_0_relu = converter_ReLU_2D(layer_0_conv) #26*26
        layer_0_pooling,layer_0_pooling_index = Maxpooling(layer_0_relu) #13*13,26*26
        #layer 1 forward propagation
        layer_1_conv = Conv(layer_0_pooling,filter1) #10*10
        layer_1_relu = converter_ReLU_2D(layer_1_conv) #10*10
        layer_1_pooling,layer_1_pooling_index = Maxpooling(layer_1_relu) #5*5,10*10
        #layer 2 forward propagation
        layer_2 = layer_1_pooling.reshape(25)
        layer_2_weight = np.dot(layer_2, weight0) + bias0
        layer_2_softmax = SoftMax(layer_2_weight)

```

```

        test_loss += Cross_entropy_loss(y_label, layer_2_softmax)
        if np.argmax(layer_2_softmax) != np.argmax(test_data.__getitem__(i)[1]):
            error += 1
    print("test loss is", test_loss/testing)
    print(1-(error/testing), "accuracy!")
    loss_test_set.append(test_loss/testing)
    return filter0, filter1, weight0, bias0, loss_training_set, loss_test_set

```

```

In [6]: test_data = Dataloader(
    path="./",
    shuffle=True,
    is_train = False,
    batch_size = 1
)

```

```

In [8]: def ConfusionMatrix_n_top3(filter0, filter1, weight0, bias0, test_data):
    confusionmatrix = np.zeros(100).reshape(10,10)
    top3matrix = np.zeros(30).reshape(10,3)
    top3matrix_index = np.zeros(30).reshape(10,3)
    for i in range(0,10000):
        y_label = test_data.__getitem__(i)[1]
        #layer 0 foward propagation
        layer_0 = test_data.__getitem__(i)[0].reshape(28,28) #28,28
        layer_0_conv = Conv(layer_0,filter0) #26*26
        #print(layer_0_conv.shape)
        layer_0_relu = converter_ReLU_2D(layer_0_conv) #26*26
        layer_0_pooling, layer_0_pooling_index = Maxpooling(layer_0_relu) #13*13,26*26
        #layer 1 foward propagation
        layer_1_conv = Conv(layer_0_pooling,filter1) #10*10
        layer_1_relu = converter_ReLU_2D(layer_1_conv) #10*10
        layer_1_pooling, layer_1_pooling_index = Maxpooling(layer_1_relu) #5*5,10*10
        #layer 2 foward propagation
        layer_2 = layer_1_pooling.reshape(25)
        layer_2_weight = np.dot(layer_2, weight0) + bias0
        layer_2_softmax = SoftMax(layer_2_weight)
        y_prediction = layer_2_softmax

        #confusionmatrix
        confusionmatrix[np.argmax(y_label)][np.argmax(y_prediction)] += 1
        index = np.argmin(top3matrix[np.argmax(y_prediction)])
        if top3matrix[np.argmax(y_prediction)][index] <= y_prediction[np.argmax(y_prediction)]:
            top3matrix[np.argmax(y_prediction)][index] = y_prediction[np.argmax(y_prediction)]
            top3matrix_index[np.argmax(y_prediction)][index] = i

    return confusionmatrix, top3matrix, top3matrix_index

```

Loss graph

```

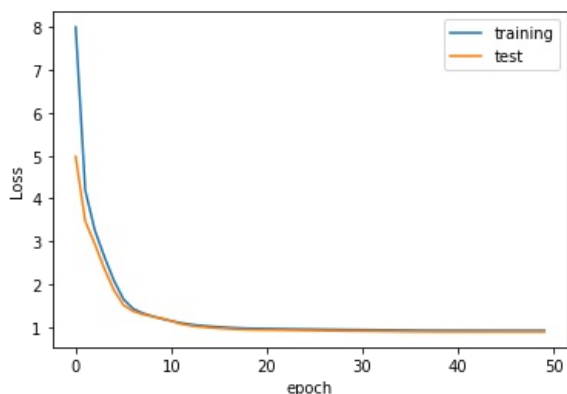
In [13]: index = [x for x in range(50)]

```

```

In [14]: plt.xlabel('epoch')
    plt.ylabel("Loss")
    plt.plot(index, loss_training_set)
    plt.plot(index, loss_test_set)
    plt.legend(['training', 'test'])
    plt.show()

```



Confusion Matrix

```
In [15]: confusionmatrix_visualization = np.array([(100*x)/np.sum(x) for x in confusionmatrix])
```

```
In [16]: confusionmatrix_visualization = np.array([ np.fix(x) for x in confusionmatrix])
```

```
In [17]: confusionmatrix_visualization.astype(int)
```

```
Out[17]: array([[ 749,   0,  16,  13,   7, 100,  26,  43,   4,  22],
 [   1, 1075,  10,   2,   4,  10,   1,   0,  11,  21],
 [  17,  15,  757,  45,   2,  19,  31,  43,  81,  22],
 [  20,  10,   52,  650,   1, 155,  24,  22,  30,  46],
 [  17,  17,  30,   7,  559,  31,  91,  34,  36, 160],
 [  16,   8,  28,  65,   6,  620,  17,  49,  42,  41],
 [  37,   9,  33,   6,  24,  29,  614,  34,  24, 148],
 [  11,  13,  55,   8,  12,   7,   9,  674,  18, 221],
 [   7,  22,  40,  23,   9,  74,  22,  31,  660,  86],
 [  10,   3,   9,   8,   9,  19,  19,  83,  32, 817]])
```

```
In [18]: sns.heatmap(confusionmatrix_visualization.astype(int), annot=True, fmt='d')

plt.title('confusionmatrix', fontsize=1)

plt.show()
```



Top-3 images with probability

```
In [20]: print(top3matrix)
```

```
[[0.99481065 0.9985179 0.99343434]
 [0.99789017 0.99818423 0.99817173]
 [0.99993893 0.99995042 0.99995912]
 [0.98797295 0.98717337 0.98916809]
 [0.9996782  0.99968892 0.99941684]
 [0.99863    0.99947199 0.99863329]
 [0.99994764 0.99979571 0.99979019]
 [0.99682825 0.9966656  0.99590167]
 [0.99964454 0.99972902 0.99952306]
 [0.99807871 0.99715717 0.99810951]]
```

```
In [19]: fig, axes = plt.subplots(10,3, figsize=(3,10))
j=0
for i,ax in enumerate(axes.flat):
    index = math.ceil(top3matrix_index.reshape(30)[j])
    ax.imshow(test_data.__getitem__(index)[0].reshape(28,28))
    j+=1
```

