Assignment 2: Report basic

1) Checking the gradients

In order to check the correctness of the gradients I used the following formula

$$\frac{|g_a - g_n|}{\max(\text{eps}, |g_a| + |g_n|)}$$
 where eps a very small positive number

where g_a is the analytically computed gradient and g_n is the numerically computed gradient and made sure that this relative error is small for each component of the gradient(bias gradient and weights gradient). To compute the numerical gradients I used the function *ComputeGradsNumSlow.py*.

First Layer

Weigh matrix

```
sum of abs differences: 7.833e-03
mean of abs values grad: 1.209e-01 grad_num: 1.209e-01
min of abs values grad: 1.457e-06 grad_num: 1.251e-06
max of abs values grad: 6.873e-01 grad_num: 6.873e-01
```

Bias matrix

```
sum of abs differences: 5.684e-07
mean of abs values grad: 1.316e-01 grad_num: 1.316e-01
min of abs values grad: 4.711e-02 grad_num: 4.711e-02
max of abs values grad: 4.574e-01 grad num: 4.574e-01
```

SecondLayer

Weigh matrix 2

```
- sum of abs differences: 7.442e-05
- mean of abs values grad: 5.806e-01 grad_num: 5.806e-01
- min of abs values grad: 5.705e-04 grad_num: 5.702e-04
- max of abs values grad: 8.951e+00 grad_num: 8.951e+00
```

Bias matrix 2

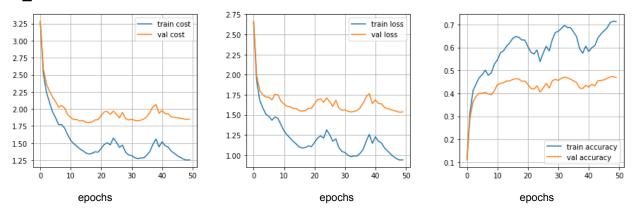
```
- sum of abs differences: 2.324e-07
- mean of abs values grad: 1.697e-01 grad_num: 1.697e-01
- min of abs values grad: 1.754e-02 grad_num: 1.754e-02
- max of abs values grad: 8.484e-01 grad_num: 8.484e-01
```

From the slow relative error we can conclude that the gradients are implemented correctly.

2) Training and validation loss/cost when using cyclical learning rates

In the following the curves for the validation, loss and accuracy of the training and validation set are shown. The network is a 2-layer NN with 50 hidden nodes. The training is done for 50 epochs with batch size 100, lambda=0 and eta_min = 1e-5, eta_max=1e-1 and $n_s = 800$.

The cyclical learning rate is represented on the rhythmic cycles of the curves every 2n_s.



The final results are:

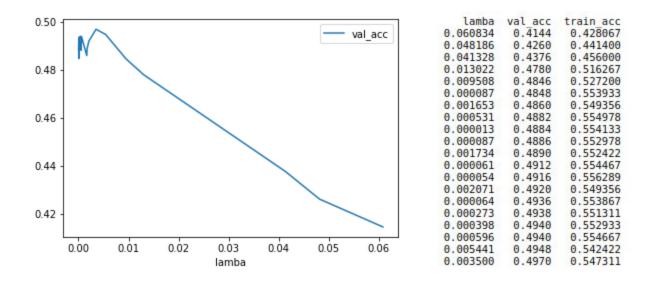
Train set accuracy: 0.6989 Validation set accuracy: 0.4638

Test set accuracy: 0.4619

3) Coarse search for lambda

In order to choose the right hyperparameter for lambda for the neural network mentioned in part 2 we do a coarse to fine search by sampling a lambda between 10^l_min and 10^l_max. Since we have to properly train the network later we search for the best lambda while using **the full data set** leaving out only 5000 samples for the validation set. For each lambda the neuronal network is trained for 4 epochs. Since n_s=900 and batch size is 100 with a dataset of 45000 samples, this corresponds to **2** cycles.

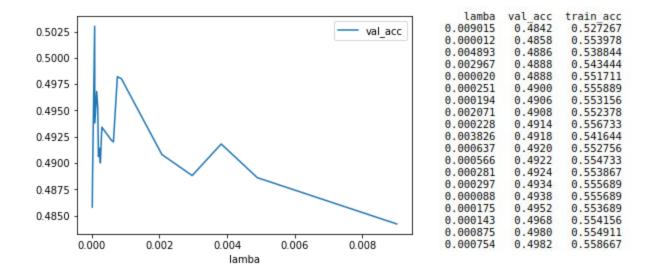
Coarse search: I_min = -5 and I_max=-1.



The three best networks deliver validation accuracies of 0.4970, 0.4948 and 0.4940 respectively.

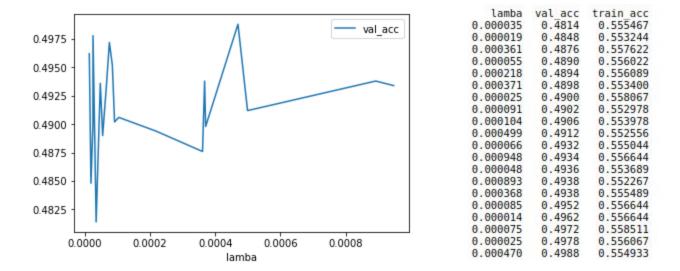
4) Fine search for lambda

Fine search 1: I_min = -5 and I_max=-2.



The three best networks deliver validation accuracies of 0.4982, 0.4980 and 0.4968 respectively.

Fine search I min = -5 and I max=-3

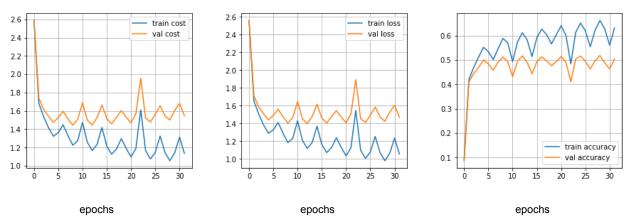


The three best networks deliver validation accuracies of 0.4988, 0.4978 and 0.4972 respectively.

In conclusion the best accuracy in the validation set is achieved by lambda=0.00047.

5) Full training set with previously found lambda

The 2-layer network with 50 hidden nodes is trained for 32 epochs which corresponds to 8 cycles using the following parameters: lambda = 0.000470, eta_min=1e-5, eta_max=1e-1, n_s=900



It delivers the following accuracies:

Train set accuracy: 0.66922222222223

Validation set accuracy: 0.5188

Test set accuracy: 0.5117