



TC 5033

Deep Learning

Fully Connected Deep Neural Networks

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Team Members:

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Activity 1b: Implementing a Fully Connected Network for Kaggle ASL Dataset

- Objective

The aim of this part of the activity is to apply your understanding of Fully Connected Networks by implementing a multilayer network for the [Kaggle ASL \(American Sign Language\) dataset](#). While you have been provided with a complete solution for a Fully Connected Network using Numpy for the MNIST dataset, you are encouraged to try to come up with the solution.

- Instructions

This activity requires submission in teams of 3 or 4 members. Submissions from smaller or larger teams will not be accepted unless prior approval has been granted (only due to exceptional circumstances). While teamwork is encouraged, each member is expected to contribute individually to the assignment. The final submission should feature the best arguments and solutions from each team member. Only one person per team needs to submit the completed work, but it is imperative that the names of all team members are listed in a Markdown cell at the very beginning of the notebook (either the first or second cell). Failure to include all team member names will result in the

grade being awarded solely to the individual who submitted the assignment, with zero points given to other team members (no exceptions will be made to this rule).

Load and Preprocess Data: You are provided a starter code to load the data. Be sure to understand the code.

Review MNIST Notebook (Optional): Before diving into this activity, you have the option to revisit the MNIST example to refresh your understanding of how to build a Fully Connected Network using Numpy.

Start Fresh: Although you can refer to the MNIST solution at any point, try to implement the network for the ASL dataset on your own. This will reinforce your learning and understanding of the architecture and mathematics involved.

Implement Forward and Backward Pass: Write the code to perform the forward and backward passes, keeping in mind the specific challenges and characteristics of the ASL dataset.

Design the Network: Create the architecture of the Fully Connected Network tailored for the ASL dataset. Choose the number of hidden layers, neurons, and hyperparameters judiciously.

Train the Model: Execute the training loop, ensuring to track performance metrics such as loss and accuracy.

Analyze and Document: Use Markdown cells to document in detail the choices you made in terms of architecture and hyperparameters, you may use figures, equations, etc to aid in your explanations. Include any metrics that help justify these choices and discuss the model's performance.

- Evaluation Criteria
 - Code Readability and Comments
 - Appropriateness of chosen architecture and hyperparameters for the ASL dataset
 - Performance of the model on the ASL dataset (at least 70% acc)
 - Quality of Markdown documentation
- Submission

Submit this Jupyter Notebook in canvas with your complete solution, ensuring your code is well-commented and includes Markdown cells that explain your design choices, results, and any challenges you encountered.

Import Libraries

	label	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	pixel9	...	pixel775	pixel776	
0	3	107	118	127	134	139	143	146	150	153	...	207	207	
1	6	155	157	156	156	156	157	156	158	158	...	69	149	
2	2	187	188	188	187	187	186	187	188	187	...	202	201	
3	2	211	211	212	212	211	210	211	210	210	...	235	234	
4	12	164	167	170	172	176	179	180	184	185	...	92	105	

5 rows × 785 columns

Load Data

Train and validation data

Inspect shape of the splitted data

```
(27455, 784)
(27455,)
(3586, 784)
(3586, 1)
(3586, 784)
(3586, 1)
```

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Normalise

Lambda function that applies feature scaling to the dataset x_data using the mean x_mean and standard deviation x_std. Feature scaling is a method used to standardize the range of independent variables or features of data.

```
(3.6268384e-06, 0.99999946)
```

Plot Samples

Plot a random sample



The sampled image represents a: w

Equations for Architecture and Loss Function of a ReLU-Activated Neural Network Model

$$z^1 = W^1 X + b^1$$

$$a^1 = \text{ReLU}(z^1)$$

$$z^2 = W^2 a^1 + b^2$$

$$\hat{y} = \frac{e^{z^2_k}}{\sum_j e^{z^2_j}}$$

$$\mathcal{L}(\hat{y}^i, y^i) = -y^i \ln(\hat{y}^i) = -\ln(\hat{y}^i)$$

$$\mathcal{J}(w, b) = \frac{1}{num_samples} \sum_{i=1}^{num_samples} -\ln(\hat{y}^i)$$

Additional Functions

Mini batches

Linear, ReLU and Sequential classes

Linear class

ReLU class

Sequential class

Cost Function

Training Function

Accuracy Function

Create your model and train it

Model Hyperparameters

Neural Network Architecture:

Input -> Linear -> ReLU -> Linear -> ReLU -> Linear -> ReLU -> Linear -> Output

This type of architecture is common in feedforward neural networks, where the goal is to transform the input data through successive layers of computation to make a prediction. ReLU activation functions are used to help combat the vanishing gradient problem, which allows the network to learn faster and perform better on a variety of tasks.

The intention is to test with different combinations of model hyperparameters to determine which performs best.

Training with Neurons: 300, Learning Rate: 0.001

epochs: 1	cost: 0.2982	accuracy: 0.7214
epochs: 2	cost: 0.0145	accuracy: 0.7705
epochs: 3	cost: 0.0050	accuracy: 0.7889
epochs: 4	cost: 0.0041	accuracy: 0.7875
epochs: 5	cost: 0.0022	accuracy: 0.7922
epochs: 6	cost: 0.0015	accuracy: 0.7922
epochs: 7	cost: 0.0013	accuracy: 0.7895
epochs: 8	cost: 0.0013	accuracy: 0.7920
epochs: 9	cost: 0.0011	accuracy: 0.7945
epochs: 10	cost: 0.0012	accuracy: 0.7936
epochs: 11	cost: 0.0008	accuracy: 0.7942
epochs: 12	cost: 0.0006	accuracy: 0.7953
epochs: 13	cost: 0.0006	accuracy: 0.7956
epochs: 14	cost: 0.0006	accuracy: 0.7956
epochs: 15	cost: 0.0005	accuracy: 0.7953
epochs: 16	cost: 0.0005	accuracy: 0.7959
epochs: 17	cost: 0.0004	accuracy: 0.7959
epochs: 18	cost: 0.0005	accuracy: 0.7956
epochs: 19	cost: 0.0004	accuracy: 0.7945
epochs: 20	cost: 0.0003	accuracy: 0.7959

Accuracy: 0.8045

Training with Neurons: 500, Learning Rate: 0.001

epochs: 1	cost: 0.4810	accuracy: 0.6849
epochs: 2	cost: 0.0671	accuracy: 0.7741
epochs: 3	cost: 0.0092	accuracy: 0.7822
epochs: 4	cost: 0.0039	accuracy: 0.7847
epochs: 5	cost: 0.0029	accuracy: 0.7856
epochs: 6	cost: 0.0016	accuracy: 0.7847
epochs: 7	cost: 0.0013	accuracy: 0.7869
epochs: 8	cost: 0.0013	accuracy: 0.7867
epochs: 9	cost: 0.0010	accuracy: 0.7864
epochs: 10	cost: 0.0006	accuracy: 0.7858
epochs: 11	cost: 0.0006	accuracy: 0.7853
epochs: 12	cost: 0.0007	accuracy: 0.7853
epochs: 13	cost: 0.0006	accuracy: 0.7858
epochs: 14	cost: 0.0006	accuracy: 0.7867
epochs: 15	cost: 0.0005	accuracy: 0.7864
epochs: 16	cost: 0.0004	accuracy: 0.7875
epochs: 17	cost: 0.0004	accuracy: 0.7878
epochs: 18	cost: 0.0004	accuracy: 0.7867
epochs: 19	cost: 0.0004	accuracy: 0.7878
epochs: 20	cost: 0.0004	accuracy: 0.7878

Accuracy: 0.8067

Training with Neurons: 700, Learning Rate: 0.001

epochs: 1	cost: 0.7115	accuracy: 0.7094
epochs: 2	cost: 0.0135	accuracy: 0.7727
epochs: 3	cost: 0.0055	accuracy: 0.7727
epochs: 4	cost: 0.0031	accuracy: 0.7780
epochs: 5	cost: 0.0022	accuracy: 0.7780
epochs: 6	cost: 0.0014	accuracy: 0.7783
epochs: 7	cost: 0.0013	accuracy: 0.7794
epochs: 8	cost: 0.0009	accuracy: 0.7800
epochs: 9	cost: 0.0012	accuracy: 0.7803
epochs: 10	cost: 0.0009	accuracy: 0.7811
epochs: 11	cost: 0.0008	accuracy: 0.7814
epochs: 12	cost: 0.0006	accuracy: 0.7814
epochs: 13	cost: 0.0007	accuracy: 0.7814
epochs: 14	cost: 0.0006	accuracy: 0.7814
epochs: 15	cost: 0.0004	accuracy: 0.7817
epochs: 16	cost: 0.0005	accuracy: 0.7814
epochs: 17	cost: 0.0007	accuracy: 0.7817
epochs: 18	cost: 0.0004	accuracy: 0.7817
epochs: 19	cost: 0.0003	accuracy: 0.7819
epochs: 20	cost: 0.0003	accuracy: 0.7814

Accuracy: 0.7962

Training with Neurons: 300, Learning Rate: 0.0005

epochs: 1	cost: 0.2447	accuracy: 0.7724
epochs: 2	cost: 0.0252	accuracy: 0.7998
epochs: 3	cost: 0.0114	accuracy: 0.8026
epochs: 4	cost: 0.0073	accuracy: 0.8093
epochs: 5	cost: 0.0058	accuracy: 0.8109
epochs: 6	cost: 0.0048	accuracy: 0.8132
epochs: 7	cost: 0.0034	accuracy: 0.8146
epochs: 8	cost: 0.0026	accuracy: 0.8143
epochs: 9	cost: 0.0025	accuracy: 0.8143
epochs: 10	cost: 0.0019	accuracy: 0.8176
epochs: 11	cost: 0.0017	accuracy: 0.8171
epochs: 12	cost: 0.0016	accuracy: 0.8176
epochs: 13	cost: 0.0015	accuracy: 0.8171
epochs: 14	cost: 0.0017	accuracy: 0.8187
epochs: 15	cost: 0.0014	accuracy: 0.8193
epochs: 16	cost: 0.0011	accuracy: 0.8187
epochs: 17	cost: 0.0009	accuracy: 0.8185
epochs: 18	cost: 0.0012	accuracy: 0.8173
epochs: 19	cost: 0.0009	accuracy: 0.8196
epochs: 20	cost: 0.0008	accuracy: 0.8199

Accuracy: 0.8224

Training with Neurons: 500, Learning Rate: 0.0005

epochs: 1	cost: 0.1366	accuracy: 0.7504
epochs: 2	cost: 0.0171	accuracy: 0.7649
epochs: 3	cost: 0.0111	accuracy: 0.7638
epochs: 4	cost: 0.0065	accuracy: 0.7733

epochs: 5	cost: 0.0046	accuracy: 0.7683
epochs: 6	cost: 0.0044	accuracy: 0.7786
epochs: 7	cost: 0.0032	accuracy: 0.7775
epochs: 8	cost: 0.0030	accuracy: 0.7794
epochs: 9	cost: 0.0024	accuracy: 0.7811
epochs: 10	cost: 0.0021	accuracy: 0.7836
epochs: 11	cost: 0.0016	accuracy: 0.7805
epochs: 12	cost: 0.0013	accuracy: 0.7825
epochs: 13	cost: 0.0014	accuracy: 0.7858
epochs: 14	cost: 0.0011	accuracy: 0.7875
epochs: 15	cost: 0.0012	accuracy: 0.7872
epochs: 16	cost: 0.0012	accuracy: 0.7878
epochs: 17	cost: 0.0010	accuracy: 0.7883
epochs: 18	cost: 0.0011	accuracy: 0.7897
epochs: 19	cost: 0.0009	accuracy: 0.7878
epochs: 20	cost: 0.0009	accuracy: 0.7895

Accuracy: 0.7975

Training with Neurons: 700, Learning Rate: 0.0005

epochs: 1	cost: 0.1377	accuracy: 0.7323
epochs: 2	cost: 0.0222	accuracy: 0.7699
epochs: 3	cost: 0.0108	accuracy: 0.7833
epochs: 4	cost: 0.0074	accuracy: 0.7867
epochs: 5	cost: 0.0044	accuracy: 0.7917
epochs: 6	cost: 0.0039	accuracy: 0.7889
epochs: 7	cost: 0.0034	accuracy: 0.7903
epochs: 8	cost: 0.0024	accuracy: 0.7869
epochs: 9	cost: 0.0022	accuracy: 0.7934
epochs: 10	cost: 0.0016	accuracy: 0.7925
epochs: 11	cost: 0.0018	accuracy: 0.7925
epochs: 12	cost: 0.0015	accuracy: 0.7928
epochs: 13	cost: 0.0014	accuracy: 0.7934
epochs: 14	cost: 0.0013	accuracy: 0.7934
epochs: 15	cost: 0.0011	accuracy: 0.7962
epochs: 16	cost: 0.0011	accuracy: 0.7948
epochs: 17	cost: 0.0009	accuracy: 0.7942
epochs: 18	cost: 0.0010	accuracy: 0.7948
epochs: 19	cost: 0.0008	accuracy: 0.7934
epochs: 20	cost: 0.0007	accuracy: 0.7962

Accuracy: 0.7987

Training with Neurons: 300, Learning Rate: 0.0001

epochs: 1	cost: 0.8526	accuracy: 0.5945
epochs: 2	cost: 0.3229	accuracy: 0.6949
epochs: 3	cost: 0.1706	accuracy: 0.7170
epochs: 4	cost: 0.0836	accuracy: 0.7379
epochs: 5	cost: 0.0576	accuracy: 0.7443
epochs: 6	cost: 0.0407	accuracy: 0.7485
epochs: 7	cost: 0.0343	accuracy: 0.7532
epochs: 8	cost: 0.0250	accuracy: 0.7540

epochs: 9	cost: 0.0223	accuracy: 0.7538
epochs: 10	cost: 0.0195	accuracy: 0.7593
epochs: 11	cost: 0.0141	accuracy: 0.7607
epochs: 12	cost: 0.0127	accuracy: 0.7610
epochs: 13	cost: 0.0116	accuracy: 0.7635
epochs: 14	cost: 0.0097	accuracy: 0.7635
epochs: 15	cost: 0.0110	accuracy: 0.7644
epochs: 16	cost: 0.0082	accuracy: 0.7672
epochs: 17	cost: 0.0084	accuracy: 0.7674
epochs: 18	cost: 0.0078	accuracy: 0.7705
epochs: 19	cost: 0.0076	accuracy: 0.7674
epochs: 20	cost: 0.0065	accuracy: 0.7683

Accuracy: 0.7708

Training with Neurons: 500, Learning Rate: 0.0001

epochs: 1	cost: 0.7755	accuracy: 0.6316
epochs: 2	cost: 0.2999	accuracy: 0.7128
epochs: 3	cost: 0.1288	accuracy: 0.7451
epochs: 4	cost: 0.0819	accuracy: 0.7418
epochs: 5	cost: 0.0511	accuracy: 0.7518
epochs: 6	cost: 0.0348	accuracy: 0.7577
epochs: 7	cost: 0.0291	accuracy: 0.7560
epochs: 8	cost: 0.0199	accuracy: 0.7624
epochs: 9	cost: 0.0212	accuracy: 0.7613
epochs: 10	cost: 0.0150	accuracy: 0.7658
epochs: 11	cost: 0.0134	accuracy: 0.7638
epochs: 12	cost: 0.0110	accuracy: 0.7660
epochs: 13	cost: 0.0102	accuracy: 0.7694
epochs: 14	cost: 0.0108	accuracy: 0.7683
epochs: 15	cost: 0.0087	accuracy: 0.7677
epochs: 16	cost: 0.0071	accuracy: 0.7674
epochs: 17	cost: 0.0078	accuracy: 0.7727
epochs: 18	cost: 0.0068	accuracy: 0.7691
epochs: 19	cost: 0.0073	accuracy: 0.7691
epochs: 20	cost: 0.0070	accuracy: 0.7691

Accuracy: 0.7836

Training with Neurons: 700, Learning Rate: 0.0001

epochs: 1	cost: 0.6491	accuracy: 0.6729
epochs: 2	cost: 0.2533	accuracy: 0.7412
epochs: 3	cost: 0.1106	accuracy: 0.7571
epochs: 4	cost: 0.0618	accuracy: 0.7803
epochs: 5	cost: 0.0432	accuracy: 0.7800
epochs: 6	cost: 0.0364	accuracy: 0.7844
epochs: 7	cost: 0.0244	accuracy: 0.7797
epochs: 8	cost: 0.0226	accuracy: 0.7833
epochs: 9	cost: 0.0184	accuracy: 0.7856
epochs: 10	cost: 0.0151	accuracy: 0.7822
epochs: 11	cost: 0.0127	accuracy: 0.7833
epochs: 12	cost: 0.0095	accuracy: 0.7833

epochs: 13	cost: 0.0089	accuracy: 0.7828
epochs: 14	cost: 0.0087	accuracy: 0.7844
epochs: 15	cost: 0.0084	accuracy: 0.7853
epochs: 16	cost: 0.0081	accuracy: 0.7847
epochs: 17	cost: 0.0063	accuracy: 0.7853
epochs: 18	cost: 0.0061	accuracy: 0.7844
epochs: 19	cost: 0.0053	accuracy: 0.7864
epochs: 20	cost: 0.0058	accuracy: 0.7839

Accuracy: 0.7934

Best Model Accuracy: 0.8224

Test your model on Random data from your test set



Predicted value is: m, real value is:m