Walking through the projects

Handwritten digit recognition using Logistic regression, Feed Forward Neural Network

Smita Bhattacharya M.Sc. in Data Science & Artificial Intelligence Saarland University

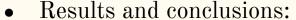
Handwritten digit recognition

- Goal: Handwritten digit recognition using a Logistic Regression Model
- Dataset: MNIST an inbuilt torchvision Image Classification Dataset
- Labels: 0, 1, 2, 3, 4, 5, 6, 7, 8, 9
- Model Architecture:
 - Logistic Regression model containing one fully connected linear layer
 - **Input:** Each training example is a vector, each 1x28x28 image tensor is *flattened* into a vector of size 784 (28*28) before being passed into the model
 - **Output:** The output for each image is a vector of size 10, with each element signifying the probability of a particular target label (i.e., 0 to 9). The predicted label for an image is simply the one with the highest probability. The **softmax function** is used.

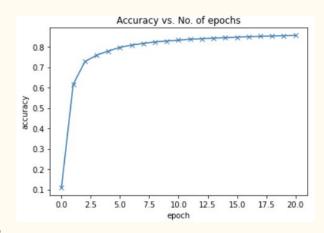
Handwritten digit recognition(Cont..)

• Training and evaluation

- Optimizer: Stochastic Gradient Descent
- Loss Function: Cross Entropy
- Accuracy: Computed as percentage of correct predictions



- Test accuracy: At the end of 20 epochs is approximately 83%
- Validation accuracy: Improved with each epoch
- Conclusion: Linear model does not take into consideration the non linear relationship which can be there in the data



Handwritten digit recognition(Cont..)

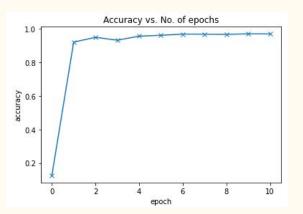
• Model Architecture:

- Fully connected Neural Network with following layers:
 - An Input layer
 - A hidden layer
 - An output layer.
- **Input:** Each training example is a vector, each 1x28x28 image tensor is *flattened* into a vector of size 784 (28*28) before being passed into the model
- Output: The output for each image is a vector of size 10, with each element signifying the probability of a particular target label (i.e., 0 to 9). The predicted label for an image is simply the one with the highest probability.

Handwritten digit recognition(Cont..)

• Training and evaluation

- Optimizer : Gradient Descent
- Loss Function: Cross Entropy
- Accuracy: Computed as percentage of correct predictions



• Results and conclusions:

- Test accuracy: At the end of 10 epochs is approximately 92%
- Validation accuracy: Improved with each epoch
- Conclusion: FFNN model does not improve beyond 92 %, we can add more hidden layers to learn more complex representation of the images, adding activation functions