CIFAR-10

Logistic Regression

```
In[46]:= training1000 = RandomSample[trainingData, 1000];
                                testing100 = RandomSample[testData, 100]
\mathsf{Out}[47] = \left\{ \begin{array}{c} \mathsf{out}[47] = \\ \mathsf{out}[47] = \end{array} \right. \rightarrow \mathsf{ship}, \qquad \mathsf{out}[47] \rightarrow \mathsf

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```

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AbsoluteTiming[(logit = Classify[training1000, Method → "LogisticRegression"])]

In[53]:= predLabel = logit[Keys@testing100];

```
In[54]:= realLabel = Values[testing100];
In[55]:= Counts[MapThread[Equal, {predLabel, realLabel}]]
Out[55]= \langle | \text{True} \rightarrow 78, \text{ False} \rightarrow 22 | \rangle
ln[57] = accuracy = \frac{78}{100} // N
\mathsf{Out}[57] = \ \textbf{0.78}
In[73]:= logitFunction = Association @@ Normal[logit]
In[74]:= Keys[logitFunction]
Out[74]= {Basic, Input, Output, Combiner, Decision, Models, Log}
In[80]:= ts = logitFunction["Models"] [[1]] ["Theta"];
In[81]:= Dimensions[ts]
Out[81]= \{10, 440\}
In[67]:= ClassifierInformation[logit]
         Classifier information
                                Method
                                              Logistic regression
                    Number of classes
                                              10
                   Number of features
                                              1
Out[67]=
         Number of training examples
                                              1000
          L1 regularization coefficient
                                              0
          L2 regularization coefficient
                                              10.
```

```
In[66]:= ClassifierInformation[logit, "Models"]
```

Out[66]= Missing[PropertyNotAvailable, Models]

Naive CNN model

```
naiveCNN = NetChain[{ConvolutionLayer[20, 5], 10, SoftmaxLayer[]},
         "Output" → NetDecoder[{"Class", classes}],
         "Input" → NetEncoder[{"Image", {32, 32}}]]
                                      image
                                      3-tensor (size: 3 x 32 x 32)
                       Input
                       ConvolutionLayer
                                      3-tensor (size: 20 x 28 x 28)
Out[23]= NetChain
                    2 LinearLayer
                                      vector (size: 10)
                       SoftmaxLayer
                                      vector (size: 10)
                       Output
                                      class
                                   (uninitialized)
```

```
In[24]:= cnn1layer =
```

NetTrain[naiveCNN, trainingData, ValidationSet → testData, MaxTrainingRounds → 3];

In[25]:= predClass = cnn1layer /@ (Keys@testData)

{frog, airplane, ship, ship, ship, ship, ship, airplane, airplane, ship, ship, truck, airplane, ship, airplane, ship, ship, ship, ship, dog, ship, ship, ship, ... 9952 ..., automobile, truck, truck, truck, ship, frog, truck, ship, truck, truck, ship, truck, truck, ship, truck, automobile, frog, truck, automobile, truck, automobile, automobile, truck}

In[26]:= realClass = Values[testData]

{airplane, airplane, airpl

In[27]:= Counts[MapThread[Equal, {predClass, realClass}]]

 $Out[27]= \langle | False \rightarrow 6100, True \rightarrow 3900 | \rangle$

$$ln[28]:= accuracy = \frac{3900}{10.000} // N$$

Out[28]= 0.39

In[29]:= randomTestData = RandomSample[testData, 10]

$$\text{Out} [29] = \left\{ \begin{array}{c} \longrightarrow \text{ frog, } \longrightarrow \text{ automobile, } \longrightarrow \text{ deer, } \longrightarrow \text{ automobile, } \longrightarrow \text{ ship, } \\ \longrightarrow \text{ frog, } \longrightarrow \text{ airplane, } \longrightarrow \text{ frog, } \longrightarrow \text{ cat, } \longrightarrow \text{ horse} \right\}$$

In[30]:= Keys@randomTestData

In[31]:= cnn1layer[Keys@randomTestData]

Outsile {cat, truck, dog, automobile, ship, deer, ship, cat, cat, frog}

LeNet model

```
In[32]:= lenet = NetChain[{ConvolutionLayer[20, 5], Ramp, PoolingLayer[2, 2],
        ConvolutionLayer[50, 5], Ramp, PoolingLayer[2, 2], FlattenLayer[], 500,
        Ramp, 10, SoftmaxLayer[]}, "Output" \rightarrow NetDecoder[{"Class", classes}],
       "Input" → NetEncoder[{"Image", {32, 32}}]]
```

image Input 3-tensor (size: 3 x 32 x 32) ConvolutionLayer 3-tensor (size: 20 x 28 x 28) 2 Ramp 3-tensor (size: 20 x 28 x 28) 3 PoolingLayer 3-tensor (size: 20 x 14 x 14) ConvolutionLayer 3-tensor (size: 50 x 10 x 10) 3-tensor (size: 50 x 10 x 10) Ramp Out[32]= NetChain PoolingLayer 3-tensor (size: $50 \times 5 \times 5$) FlattenLayer vector (size: 1250) LinearLayer 8 vector (size: 500) Ramp vector (size: 500) 10 LinearLayer vector (size: 10) SoftmaxLayer vector (size: 10) Output class (uninitialized)

In[34]= lenet = NetTrain[lenet, trainingData, ValidationSet → testData, MaxTrainingRounds → 3]

image Input 3-tensor (size: 3 x 32 x 32) ConvolutionLayer 3-tensor (size: 20 x 28 x 28) Ramp 3-tensor (size: 20 x 28 x 28) PoolingLayer 3-tensor (size: 20 x 14 x 14) ConvolutionLayer 3-tensor (size: 50 x 10 x 10) Ramp 3-tensor (size: 50 × 10 × 10) Out[34]= NetChain PoolingLayer 3-tensor (size: $50 \times 5 \times 5$) FlattenLayer vector (size: 1250) LinearLayer vector (size: 500) 9 Ramp vector (size: 500) 10 LinearLayer vector (size: 10) SoftmaxLayer vector (size: 10) Output class

In[35]:= predClass = lenet /@ (Keys@testData)

{ship, airplane, ship, airplane, airplane, airplane, airplane, airplane, airplane, airplane, dog, airplane, airplane, ship, airplane, airplane, 9966 ... , truck, automobile, truck, truck, truck, truck, truck, horse, Out[35]= automobile, truck, truck, truck, automobile, truck, truck, truck} large output show less show more show all set size limit...

```
In[36]:= realClass = Values[testData]
```

{airplane, airplane, truck, truck,

large output

show less

show more

show all

set size limit...

In[37]:= Counts[MapThread[Equal, {predClass, realClass}]]

$$_{\text{Out[37]=}}$$
 $\langle\,\big|\,\,\text{False}\rightarrow3883\,,\,\,\text{True}\rightarrow6117\,\big|\,\rangle$

$$ln[38] = accuracy = \frac{6117}{10000} // N$$

Out[38] = 0.6117

Out[36]=