MNIST在Keras下的 幾種學習架構

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Outlines

Machine learning basic

- underfitting/overfitting
- stochastic gradient descent
- softmax/cross-entropy

Program structure

MNIST and its network models

- linear based classifier
- multiple layer perceptron
- convolutional neural network

General resource

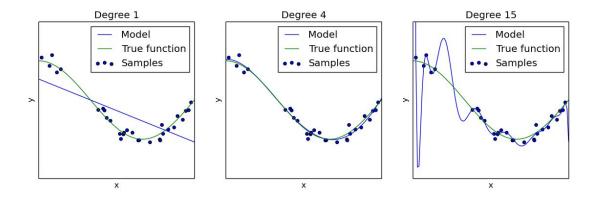
Tensorflow: https://www.tensorflow.org/get_started/

Keras: https://keras.io/getting-started/sequential-model-guide/

Deep learning book: http://www.deeplearningbook.org/

Machine learning basic - capacity

underfitting/overfitting



Source: http://scikit-learn.org/stable/auto_examples/model_selection/plot_underfitting_overfitting.html

Machine learning basic - stochastic gradient descent

Gradient descent

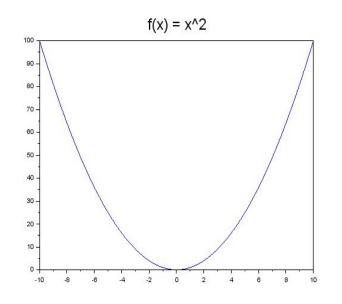
$$\mathbf{x} = \mathbf{x} - \eta \cdot \nabla_{\mathbf{x}} f(\mathbf{x})$$

Example:

$$f(x) = x^2$$

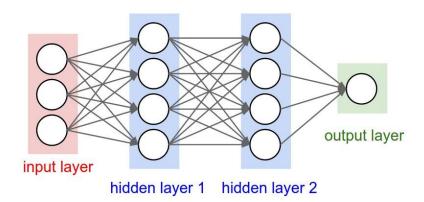
Stochastic gradient descent

$$\omega = \omega - \eta \cdot \sum_{i=1}^{n} \nabla_{\omega} f(\omega; \mathbf{x}_i)$$

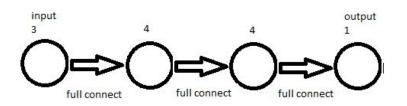


Machine learning basic - feedforward network

FNN (resource: http://cs231n.github.io/neural-networks-1/)

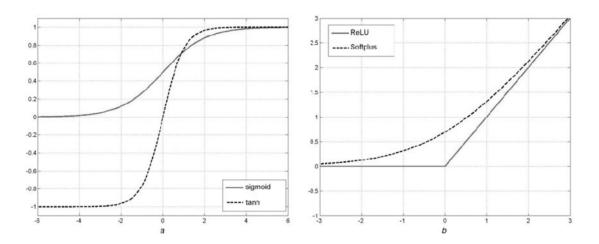


SImplified graph



Machine linearing basic - activation functions

sigmoid/tanh/relu/softplus



Resource:

https://www.researchgate.net/publication/283433254 Driving posture recognition by convolutional neural networks

Machine learning basic - categorical loss function

Sigmoid and negative log-likelihood

sigmoid:

$$y = \sigma(z) = \frac{1}{1 + e^{-z}}$$
$$\frac{d\sigma(z)}{dz} = \frac{e^{-z}}{(1 + e^{-z})^2} = \sigma(z)(1 - \sigma(z))$$

negative log-likelihood:

$$L(z) = -\hat{y}\log y - (1 - \hat{y})\log(1 - y)$$

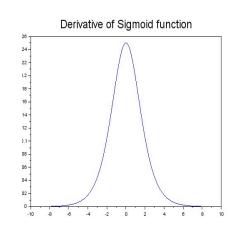
if $\hat{y} = 1$

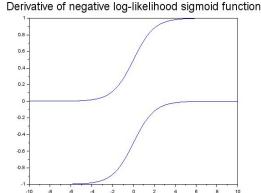
$$L(z) = -\log y = -\log \sigma(z) = \log(1 + e^{-z})$$

 $\frac{dL(z)}{dz} = \frac{-e^{-z}}{1 + e^{-z}} = \sigma(z) - 1$

if $\hat{y} = 0$

$$L(z) = -\log(1 - y) = -\log(1 - \sigma(z)) = \log(1 + e^{-z}) + z$$
$$\frac{dL(z)}{dz} = \frac{-e^{-z}}{1 + e^{-z}} + 1 = \sigma(z)$$





Machine learning basic - categorical loss function

Softmax and cross-entropy

softmax:

$$y_i = s_i(\mathbf{z}) = \frac{e^{z_i}}{\sum_{k=1}^n e^{z_k}}$$

cross-entropy

$$L(\mathbf{z}) = -\sum_{k=1}^{n} \hat{y}_k \log y_k$$

one-hot code of \hat{y} , let one to be index d:

$$\frac{\partial L(\mathbf{z})}{\partial z_i} = \frac{\partial - \log y_d}{\partial z_i}$$

if
$$i = d$$

$$\frac{\partial L(\mathbf{z})}{\partial z_d} = s_d(\mathbf{z}) - 1$$

if
$$i \neq d$$

$$\frac{\partial L(\mathbf{z})}{\partial z_i} = s_i(\mathbf{z})$$

Machine learning basic - backpropagation

Gradient + chain rule

Reference:

https://www.facebook.com/groups/1948109392087807/permalink/1980996765465 736/

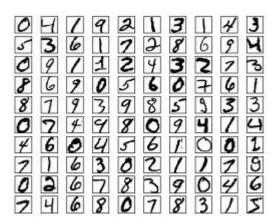
Program structure

- Dataset preprocessing
- Model construction
- Loss function/Optimization
- Training/Validation/Test
- Display

MNIST data set

MNIST is a data set of images of handwritten digits.

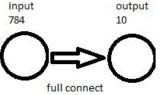
Every digit is represented by 28x28 pixels; each pixel is 256 gray levels.



Resource: http://neuralnetworksanddeeplearning.com/chap1.html

Linear based classifier

Linear based classifier model:



Modified from:

https://github.com/fchollet/keras/blob/master/examples/mnist_mlp.py

Reference: https://www.tensorflow.org/get_started/mnist/beginners

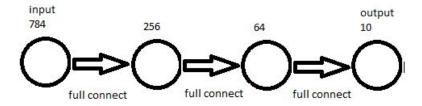
Key technology: softmax/cross-entropy/stochastic gradient descent

Keras model: # model

```
model = Sequential()
model.add(Dense(10, activation='softmax', input_shape=(784,)))
```

Multiple layer perceptron

MLP model:



Modified from:

https://github.com/fchollet/keras/blob/master/examples/mnist_mlp.py

Key technology: backpropagation/ReLU

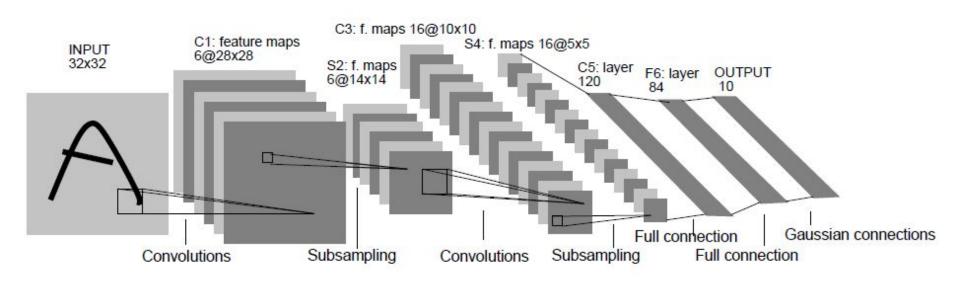
Keras model: # model

```
model = Sequential()
model.add(Dense(256, activation='relu', input_shape=(784,)))
model.add(Dense(64, activation='relu'))
model.add(Dense(10, activation='softmax'))
```

Convolutional neural network - LeNet5

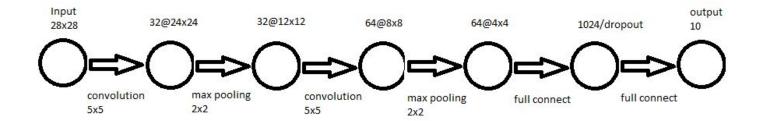
Reference:

http://www.iro.umontreal.ca/~lisa/bib/pub_subject/finance/pointeurs/lecun-98.pdf



CNN complexity

Our CNN model:



Complexity:

$$1 \cdot (5^{2} + 1) \cdot 24^{2} \cdot 32 = 479,232$$

$$32 \cdot (5^{2} + 1) \cdot 8^{2} \cdot 64 = 3,407,872$$

$$(4^{2} \cdot 64 + 1) \cdot 1024 = 1,049,600$$

$$(1024 + 1) \cdot 10 = 10,250$$

$$total = 4,946,964$$

Convolutional neural network

Modified from:

https://github.com/fchollet/keras/blob/master/examples/mnist_cnn.py

Reference: https://www.tensorflow.org/get_started/mnist/pros

Key technology: convolutional kernel/padding/stride/pooling/features/dropout

Kears model: # model

Appendix - more study notes

Backpropagation by Computational Graph:

https://www.facebook.com/groups/1948109392087807/permalink/1980996765465736/

Feedforword Networks:

https://www.facebook.com/groups/1948109392087807/permalink/1982550641977015/

Convolutional Neural Network:

https://www.facebook.com/groups/1948109392087807/permalink/1987843988114347/

Recurrent Neural Networks (RNN) and Its Applications:

https://www.facebook.com/groups/1948109392087807/permalink/2004972959734783/

Generative Adversarial Network (GAN) and Its Applications:

https://www.facebook.com/groups/1948109392087807/permalink/2025433097688769/