

Regularizations

- Regularizations

- L1: Lasso

- Function: $CostFunction = Loss + \frac{\lambda}{2m} * \sum ||w||$
 - penalize the absolute value of the weights
 - useful for feature engineering
 - Explainable: select the important features whose weights do not equal to 0, while deletes those unimportant features' weights are equal 0
 - Laplace distribution

- L2: Ridge

- Function: $CostFunction = Loss + \frac{\lambda}{2m} * \sum ||w||^2$
 - λ : regularization parameter
 - Also called *weight decay* 权重衰减, owing to it forces the weights to decay towards 0.
 - Gaussian distribution

- Coding in Keras:

- `from Keras import regularizers`
 - `model.add(Dense(64, input_dim=64, kernel_regularizer=regularizers.L2(0.01))`

- Dropout (more useful in DL)

- Concepts: At every iteration, it randomly selects some nodes, and removes them along with all of their incoming and outgoing connections.
 - use **backpropagation** to learn and update the parameters in neural network, while the nodes which are deleted not engage in the update progress.
 - the hyperparameter of dropout function: the probability(p) to choose how many nodes should be dropped.
 - Apply to hidden layers and input layers.
 - Ensemble learning
 - Coding:
 - `from keras.layers.core import Dropout`
 - `model = Sequential([Dense(output_dim=hidden1_num_units, input_dim=input_num_units, activation='relu'), Dropout(0.25), # we define $p = 0.25$ as the probability of dropping Dense(output_dim=output_num_units, input_dim=hidden5_num_units, activation='softmax'),])`

- Data Augmentation
 - increase the size of the training data
- Early Stopping
 - cross-validation strategy, keep one part of training dataset as validation set
 - If we see the performance of validation set is getting worse, we can stop the training on model
 - Coding
 - `from keras.callbacks import EarlyStopping`
 - `EarlyStopping(monitor='val_err', patience=5)`
 - **monitor**: the quantity that needs to be monitored
 - **'val_err'** denotes the validation error.
 - **Patience**: the number of epochs with no further improvement after which the training will be stopped.
- Concepts: to prevent the model too complex and overfitting, add penalization elements after the original loss function, improving the model's performance on the unseen data.→ `Cost Function=Loss + Regularization Term`
 - In deep learning, it penalizes the weight matrices of the nodes.
 - If the `regularization coefficient` is too high,
 - some weight matrices nearly equal zero
 - a simpler linear network
 - slight underfitting for the training data

以上内容整理于 [幕布文档](#)