

Deep Learning In Python using Keras

Subject: Early Stopping and Other Tips of CNN Models

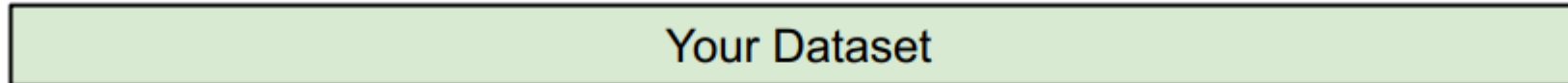
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Early Stopping

The best way to splitting dataset for implementing deep learning models

Idea #1: Choose hyperparameters that work best on the data

BAD: $K = 1$ always works perfectly on training data



Idea #2: Split data into **train** and **test**, choose hyperparameters that work best on test data

BAD: No idea how algorithm will perform on new data



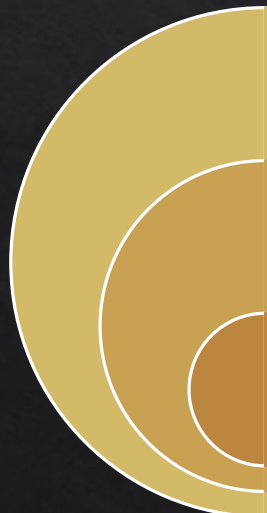
Idea #3: Split data into **train**, **val**, and **test**; choose hyperparameters on val and evaluate on test

Better!



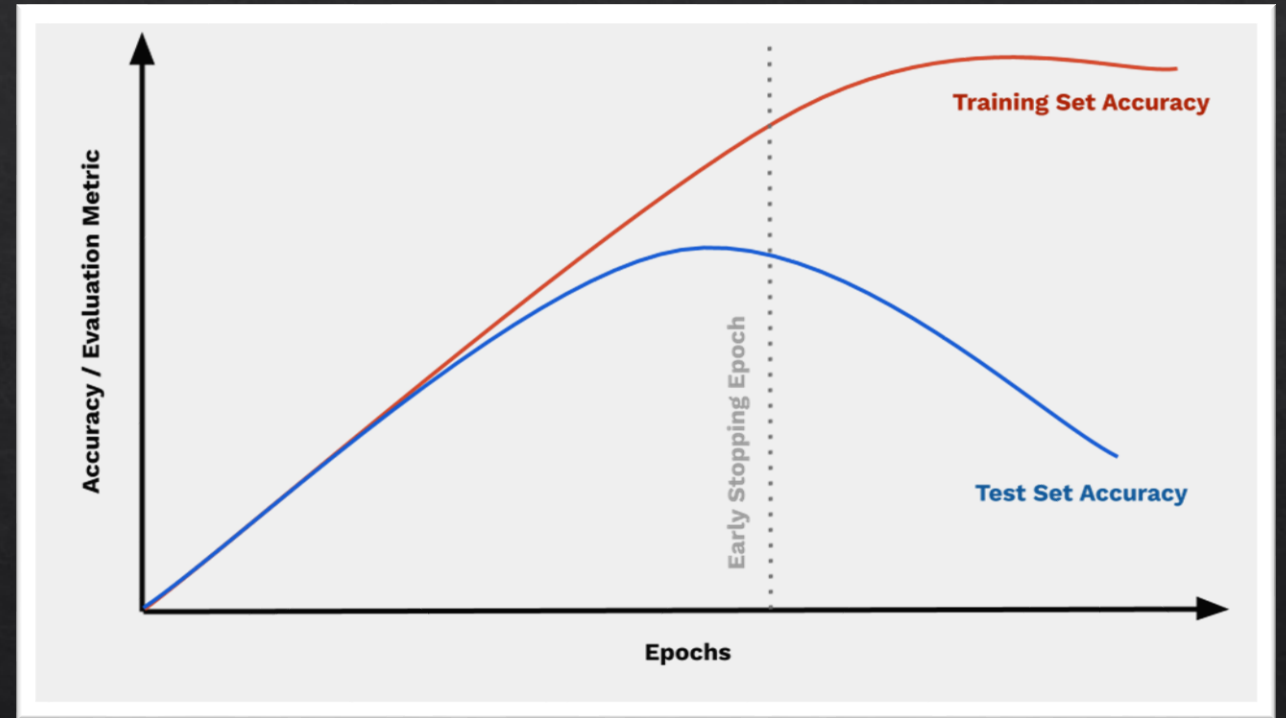
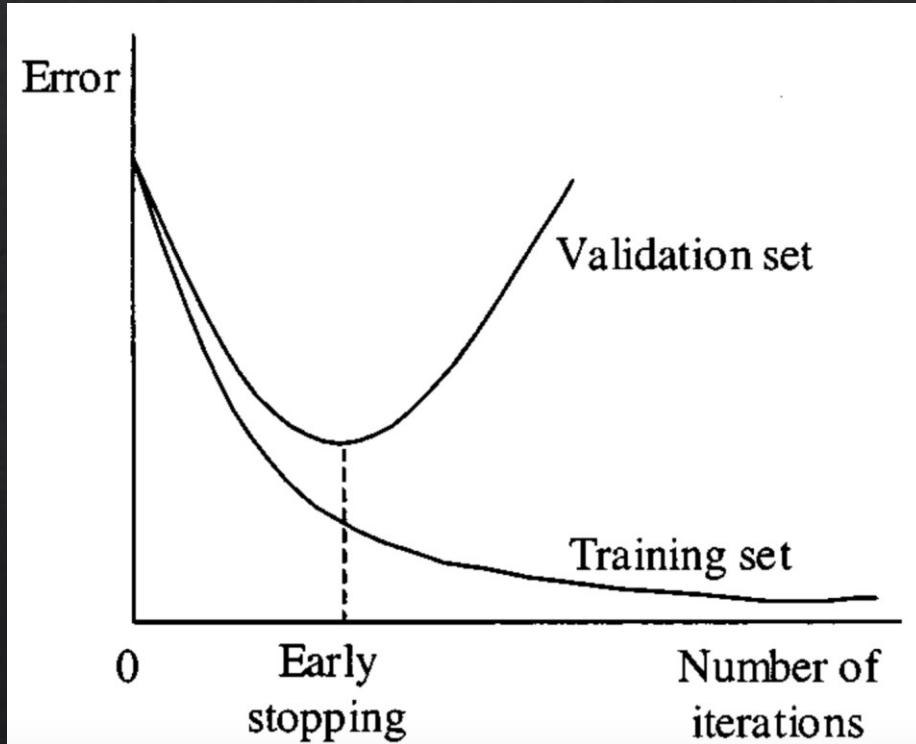
Early Stopping

- Early stopping is a method for preventing overfitting in deep learning models.
- In early stopping, the dataset is categorized in training, validation, and testing subsets.
- In the end of each epoch, the validation set is applied to the model and the validation accuracy (validation loss) is compared with training accuracy (training loss)
- If the training and validation values (accuracy or loss) are such that they are getting away from each other, the model training process will be stopped.



| | |
|----------------|--|
| Training Set | <ul style="list-style-type: none">• It is used in training procedure for training model (Updating weights) |
| Validation Set | <ul style="list-style-type: none">• It is used in training procedure for preventing overfitting model |
| Testing Set | <ul style="list-style-type: none">• It is used in testing procedure to evaluate the trained model |
| | |

Early Stopping



Other tips

**How to prevent overfitting in
deep learning models**

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graph TD; A[How to prevent overfitting in deep learning models] --> B[Using dropout layers in models]; A --> C[Early Stopping]; A --> D[Regulation in layers which contains weights and biases]
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Using dropout layers in
models

Early Stopping

Regulation in layers
which contains weights
and biases

Normalization Layer

“you want zero-mean unit-variance activations? just make them so.”

consider a batch of activations at some layer. To make each dimension zero-mean unit-variance, apply:

$$\hat{x}^{(k)} = \frac{x^{(k)} - \mathbb{E}[x^{(k)}]}{\sqrt{\text{Var}[x^{(k)}]}}$$

this is a vanilla
differentiable function...

Normalization Layer

Comparison of Normalization Layers

