



Fourth Industrial Summer School

Advanced Machine Learning

Tips for ML + KNIME

Session Objectives

- ✓ Tips on machine learning
- ✓ KNIME



Tips on Machine Learning

Important remarks



- Mainly for supervised learning but some aspects are general
- Not for academic research but for development

Some guidelines–1



- Understand your goal
 - What are you trying to achieve, translated into:
 - Classification
 - Regression
 - Unsupervised learning
 - ...
- Understand your data
 - What data you have
 - How much data you have
 - Characteristics of your data

Some guidelines–2

- Prepare your data
 - Clean your data
 - Select your features
 - Continuous vs. categorical features
 - How to encode the target
 - Data scaling
 - Data imbalance
 - Train/validation(cross-validation)/test sets
- Data collection and annotation
 - Realistic conditions
 - Dev. and test sets from the same distributions

Some guidelines–3

- Select measures appropriate for the task (get to an agreement)
 - Also depends on the data
 - Accuracy vs. precision & recall (unify it)
 - Prioritize your measures (accuracy vs. runtime), optimizing vs. satisficing
 - Change them later if needed
- What should be the size of train/dev/test partitions?
 - 70/15/15?, 60/20/20?
 - What if you huge amount of data?
- Significance interval of differences in performance
 - 95% percent confidence intervals
 - The smaller the intended progress, the larger the dev set needed

Some guidelines-4

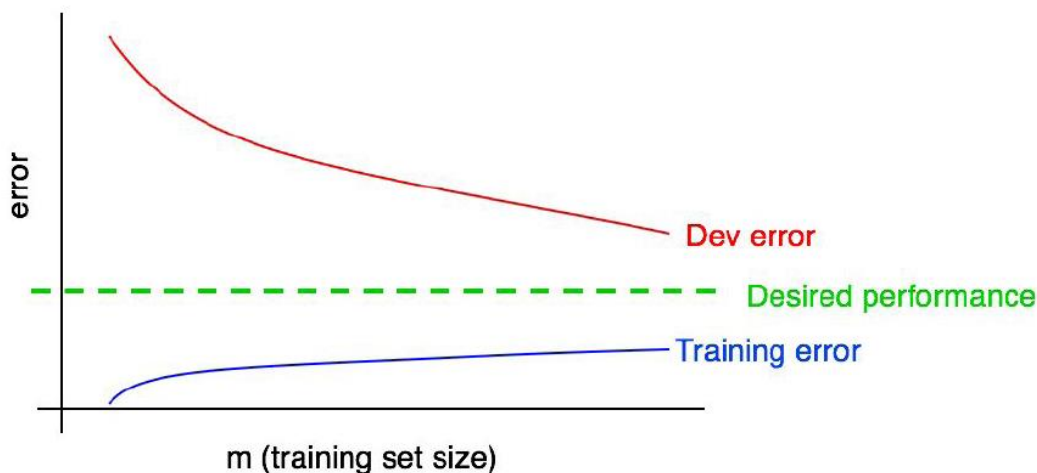
- Decide which algorithm to use
 - Start simple (linear/logistic regression) unless there is a clear case
 - Incrementally build complex ideas
 - Start small instead of making big goals from the very beginning
- Training
 - Trying to fit (overfit) and then,
 - Deal with variance
- Error analysis
 - Do error analysis to decide how to go forward by manually looking at the errors on the dev set
 - What if the dev set is large (create a small subset for manual analysis)

Some guidelines–5

- What about mislabeled data?
 - Correct them if they are a major cause of errors
 - Make sure to update the test set as well
 - What about errors in labels classified as correct category?
- Bias/variance and adding more data
 - What would be an optimal error rate?
 - All variance is avoidable (more data) but all bias is not
- High (avoidable) bias
 - Make your model more complex
 - More training
 - Error analysis on the training set

Some guidelines–6

- High variance
 - Add more training data
 - Regularization
 - Early stopping
 - Simpler model
 - Study the error curves



Andrew Ng, Machine Learning Yearning, deeplearning.ai

Some guidelines–7



- Data augmentation
- End-to-end recognition vs. a standard ML pipelines
- Use ensembles to give a final push to your results

KNIME

Data analysis and machine learning

KNIME (Konstanz Information Miner)



- It is a free and open-source data analytics, reporting and integration platform.
- Modular data pipelining concept.
- No programming needed.
- DEMO

References



- Andrew Ng, Machine Learning Yearning, deeplearning.ai
- Mehryar Mohri, Afshin Rostamizadeh, Ameet Talwalkar, Foundations of Machine Learning, second edition, The MIT Press
- Tom M. Mitchell, Machine Learning, McGraw-Hill, 1997