Machine Learning in the Real World — Part 2

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Contents

- Improving accuracy & performance:
 - -Tweaking the library
 - Processing the inputs
- Deploying to the real world
- Scaling with the data
- Monitoring
- Summary



Making Improvements

- Crucial to have tests in place first
- Improving accuracy vs improving performance
- Types of classifier:
- Decision Tree, Naive Bayes, Max Ent
 - Is your classifier a black/white box?
 - A DT can be debugged more easily



Explore the Library

- Variants on theme of MaxEnt:
- L1 and L2 regularization
- Number of corrections
- What is the tradeoff performance vs accuracy?
- -> Must have tests in place first



Modify the Inputs?

- Input is text, human readable
- Temptation is to process, use our knowledge of domain/language to give the classifier a helping hand
- Resist this the mind is an imperfect classifier
- Aim to model how the world is, not how we think of it
- Do however look at regularisation of inputs



Process the Inputs

- Remove irrelevant differences
 - Capitalisation, punctuation
 - Stemming "accounts", "accounting", "accounted"
 - Spelling conventions "license" vs "licence"
- Discard "invalid" data
- Synonyms?
 - "expense" == "costs"? In all contexts?
- Antonyms?
 - Profit/Loss can be the same thing
- Depends on the domain



Real World Deployment

- Fast, reliable, adaptive
- Must ship with built-in data
- Classifier slow to build, quick to respond
- Need to be able to serialize results
 - Gives fast startup
 - Custom serialization vs java Serializable



Real World Deployment

- Other real world concerns
- Sharing data
- Migrating data
- Fallback
 - What if you have little or no data?
 - What if confidence is low
 - Our tactic, use search / composite



Scaling - Problems

- Data increasing constantly, over 100k items
- Diversity of data increasing, both number of features (f) and size of universe (u)
- Critical value is product of f * u.
- Reached 15.6 million in live service
- Problems
 - Classifier build prohibitively slow (4:35hrs)
 - Classifier build uses > 2GB memory
 - Serialized classifiers > 1GB, problems storing to db
 - The whole service is at risk if any of the above goes wrong



Scaling - Solutions

- Limit classifier building to "quiet" times
- Separate classifier building from web application
- Stemming etc. reduces the number of distinct features in play
- Store classifiers as discrete units not a collection
- Always use compression!
- Limit the amount of training data used prioritise most recent
- How do you know what effect this is having on accuracy?



Monitoring Performance

- Realistic data is vital for testing hard to generate
- Developers won't necessarily have access to this data

Logging

- Record stats, how much data, how long, how much memory
- Assumptions about the shape of your data invalid?
- Have these logs mirrored, convenient to view
- Record the confidential inputs, access on request



Monitoring Accuracy

- How do you know if you are giving the right answers?
- Much harder to track than performance
- Customer feedback
 - Time consuming, subjective
- Track how many suggestions were accepted
 - self-perpetuating, if users are lazy
- Get the real data and perform x fold tests
 - Not allowed the data
 - Obfuscate the real data and then test?



Monitoring Accuracy Solution

- Run tests on the live data in original format and only report results
- Report statistics why is our feature set so large?
- Report accuracy what is success rate/confidence when run on existing data
- Currently not using x fold tests, just on everything
 - Not as fair, as we are comparing to seen content
 - Much quicker as don't need to rebuild classifiers



Summary

- Test first, record performance and accuracy
- Regularize your input data
- Save your work serialize classifiers
- Perform batch tasks offline
- Monitor what's happening



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