

Spark+Al Summit Europe - Oct. 2018

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#SAISEnt6



Nokia

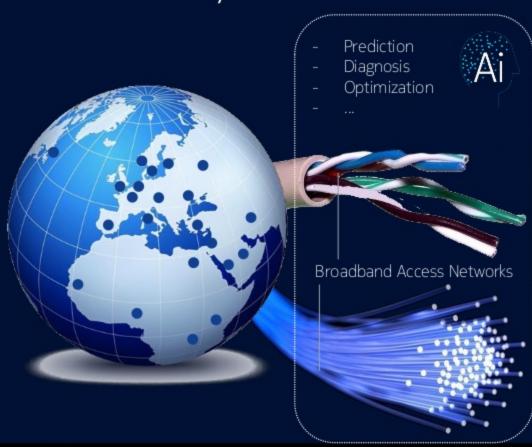


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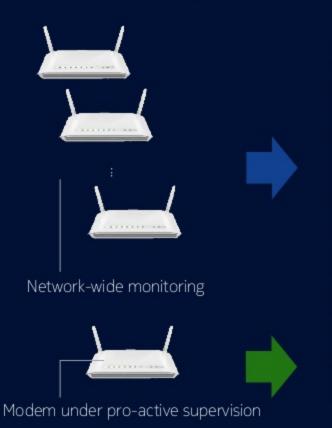
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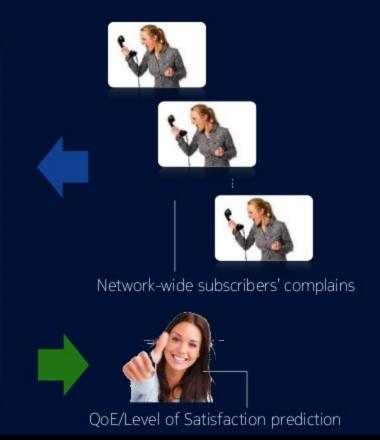


Digital Experience



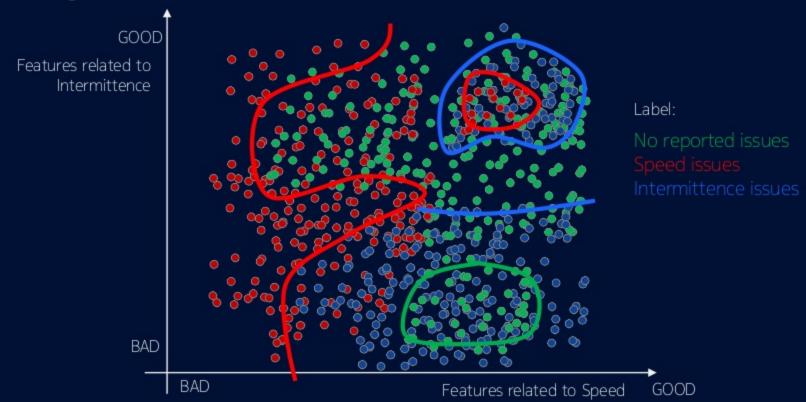






Pattern recognition Because of: Data transformation Domain expertise KPI's computation Lack of information a resampling step, based on pattern recognition, Performance validation Scope/requirements has **Lodg** added to the process. Model selection Time-series FE Modelling Balancing Machine Learning

Pattern recognition





Pattern recognition





- ETL
- Windowing/Pre-processing over >100M data rows
- Spark MLlib
- Efficient/distributed learning
- Execution



- Complete Spark MLlib API
- Java world
- Ability to produce JAR files



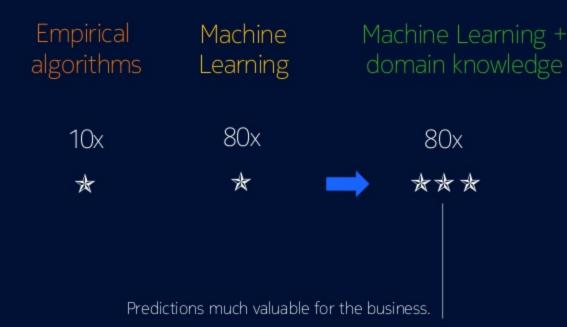
- Executed as Spark jobs
- Compiled code



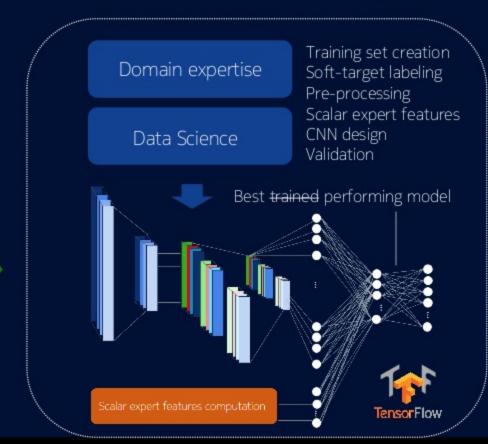


Prediction LIFT (gain)

Field performances









connection impairments



Modem under pro-active

supervision

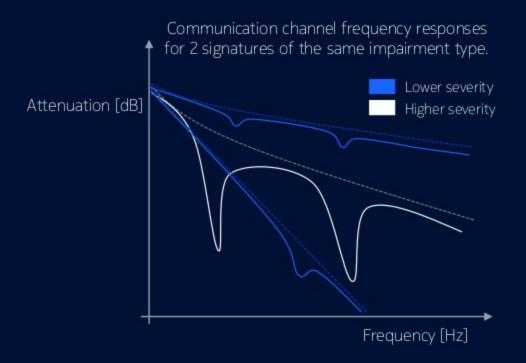
Soft-target labeling



- « Experts » (e.g. humans, authors) labelled formally each digit.
- No ambiguity is assumed during the training set creation (hard-labelling).

MNIST handwritten digits

Soft-target labeling



- In problem detection, there is the notion of « severity ».
- This means within the same class of problem, the returned confidence needs also to reflect such severity.
- Solution is to make use of softlabelling.
- Domain knowledge is required to assist deriving such soft-labels.
- Spark has been used to facilitate such processing over >20M curves.

Expert scalar features addition



- Having the « perfect trained model » would require to build the « perfect training set ».
- Building a large, various and unbiased training set is hard.
- The convolutional layers might get therefore biased, leading to extra sensitivity.
- Adding empirical quantities to the fully-connected layer have helped in gaining in robustness (conservative approach).







- CNN model design
- Training over >20M samples
- Expert scalar features computation
- Trained model/session storage

- Pre-processing
- TensorFlow Python API

- Soft-target labelling
- Hyper-parameter tuning (« grid search ») distribution
- Distributed execution



^{*} Narrower coverage.



Last advices...

- Data Science gives always better results when you know what is behind the data.
- Increasing your domain knowledge will save you a lot of time (and will make you a better data scientist ☺).
- ML/DL models that perform well in Notebooks may not give expected results in the field. Knowing how to move forward is the key!
- DL models usually performs better when guided with domain expertise.
- Try to get domain experts and data scientists in the same team.

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