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Applied Data Science Project

L3 - Model & data-centric data science projects



Giuseppe Rizzo Turin, October 3, 2022





PASSION FOR INNOVATION





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European Laboratory for Learning and Intelligent Systems



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iterative processes meant to refine the quality of the solution









TORINO

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artificial intelligence = data + model (software + algorithm)







Data-centric: the focus is on acquiring further examples or cleaning the collected ones to retrain the algorithm an generate a new model. The output of this activity is extending the dataset that is used for training

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Model-centric: the focus is on modifying the algorithm by extending the neural architecture (for instance having more layers, new residual connections) and then train it with the data at disposal





data is vital for creating any sort of artificial intelligence





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improving data has a big impact to artificial intelligence even more than model optimization

unless of radical changes in the code thus not optimization







Inspecting steel sheets for defects



Examples of defects

Baseline system: 76.2% accuracy Target: 90.0% accuracy

Andrew Ng





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	Steel defect detection
Baseline	76.2%
Model-centric	+0% (76.2%)
Data-centric	+16.9% (93.1%)



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0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0

	Solar panel	Surface inspection
Baseline	75.68%	85.05%
Model-centric	+0.04% (75.72%)	+0.00% (85.05%)
Data-centric	+3.06% (78.74%)	+0.4% (85.45%)

Inksfoundation.com

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Improving data turns out to be key for a better artificial intelligence solution

Note: Improving a code is different than designing a new, breakthrough, code however the effort for the latter is way higher than improving data and the return of the effort (may) be very high

Take home message: we consider the data improvement as an easier and necessary step when developing a machine intelligence before starting a new venture



Easier step



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Strategies for data improvement:

more examples augmentation
 completeness cleaning
 consistency







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Take or generate new examples







Task: Label cars









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Task: Label cars

Annotator 1









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Task: Label cars

Annotator 2









Consistency in annotation turns out to be crucial for the minimizing the potential error of the intelligence

However, ensuring a consistent dataset is a not obvious task

It involves:

TORINO

- how the task has been conceived
- how the intervention of the human has been designed
- how did human annotators perform their task
- how the dataset has been packaged





Andrew Ng



TORINO



Task: Label cars







TORINO



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Task: Label cars

Annotator 1









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Task: Label cars

Annotator 2









Completeness in annotation turns out to be crucial for improving coverage to the intelligence

However, ensuring a complete dataset is a not obvious task

It involves:

- how the task has been conceived
- how the intervention of the human has been designed
- how did human annotators perform their task
- how the dataset has been packaged





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Good data is:

Good data

- Defined consistently (definition of labels y is unambiguous)
- Cover of important cases (good coverage of inputs x)
- Has timely feedback from production data (distribution covers data drift and concept drift)
- Sized appropriately

We also refer to good data with the concept of clean data





Example: Clean vs. noisy data



Note: Big data problems where there's a long tail of rare events in the input (web search, self-driving cars, recommender systems) are also small data problems.

Andrew Ng



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model encapsulates the intelligence in an executable environment that embeds both training data and algorithm







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Improving a model is a hard task because it inherits the challenges related to optimize both data and algorithm







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Strategies for model optimization

- Any change in the data, if statistically relevant, is propagated to the final output of the model. This links to the previous topic
- Change in the algorithm, for instance the addition of a new layer in a neural architecture, or eventually, a brand new architecture
- Change in the hyperparameter set, for instance n_layers, or learning rate. This change modifies the parameter weights





PASSION FOR INNOVATION

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Thank you for your attention.

Questions?







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CONTACTS

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