

Learning **for** Big Data

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some sampled pages of my keynote talk
in IEEE BigData 2015 Taipei Satellite Session

About the Title

- “Learning **for** Big Data”

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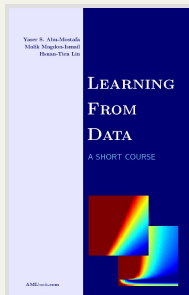
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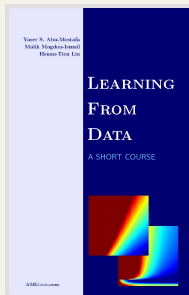
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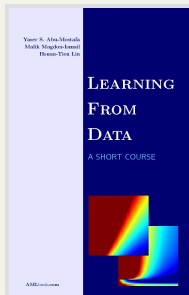


as machine learning
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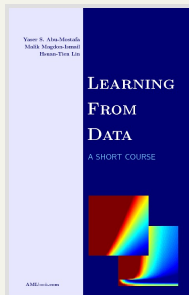


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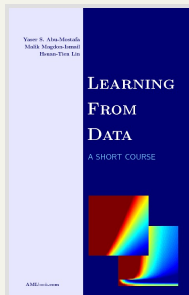
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will focus on **human** learning **for** big data

Human Learning for Big Data

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*I wish I had an answer to that
because I'm tired of answering that question.
—Yogi Berra (Athlete) 😊*

Appendix: ML Foundations on NTU@Coursera

<https://www.coursera.org/course/ntumlone>

When can machines learn?

- L1: the learning problem (😊)
- L2: learning to answer yes/no (😊)
- L3: types of learning (😊)
- L4: feasibility of learning

Why can machines learn?

- L5: training versus testing
- L6: theory of generalization
- L7: the VC dimension (😊)
- L8: noise and error

How can machines learn?

- L9: linear regression (😊)
- L10: logistic regression (😊)
- L11: linear models for classification (😊)
- L12: nonlinear transformation (😊)

How can machines learn better?

- L13: hazard of overfitting (😊)
- L14: regularization (😊)
- L15: validation (😊)
- L16: three learning principles (😊)

😊 \approx must-learn