

The Machine Learning Pipeline

CS 3244
Machine Learning

1

B



NUS | Computing

Instructor: Kan Min-Yen

kanmy@comp.nus.edu.sg

- Easy to remember: $\text{min}(x)$
- Ph.D./B.S. in CS, Columbia University

Research Interests

- Web
- Information Retrieval
- Natural Language Processing

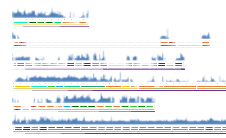




IoT Sensors

Health Behavior Change

Data Analytics



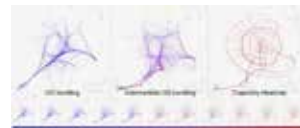
NUS Ubicomp Lab

Apps and Analytics for Smart Cities and Healthcare

<http://ubiquitous.comp.nus.edu.sg>



Explainable Artificial Intelligence



Interactive Data Visualization

[Instructor] Brian Lim

brianlim@comp.nus.edu.sg

- Asst. Prof. in Computer Science
- Ph.D. in HCI, Carnegie Mellon University
- B.S. in Engineering Physics, Cornell University

Research Interests

- HCI: understand people with tech, help people with tech
- Explainable Artificial Intelligence
- Ubiquitous Computing
- Data analysis and visualization
- Smart Health and Smart Cities

Forecast

Learning Outcomes for this week:

- Understand the rationale for our course's structure, the why of our module
- Recall the components of the course and their weightages and general deadlines
- Execute a basic machine learning workflow in Google Colab
- Conceptualize the division of labor in a typical machine learning project



You'll need this for this lecture.

<https://bit.ly/cs3244-join-slack>

- 👉 Use your NUS @nus.edu.sg or @u.nus.edu address.
- 👉 Use your LumiNUS name for your display name.





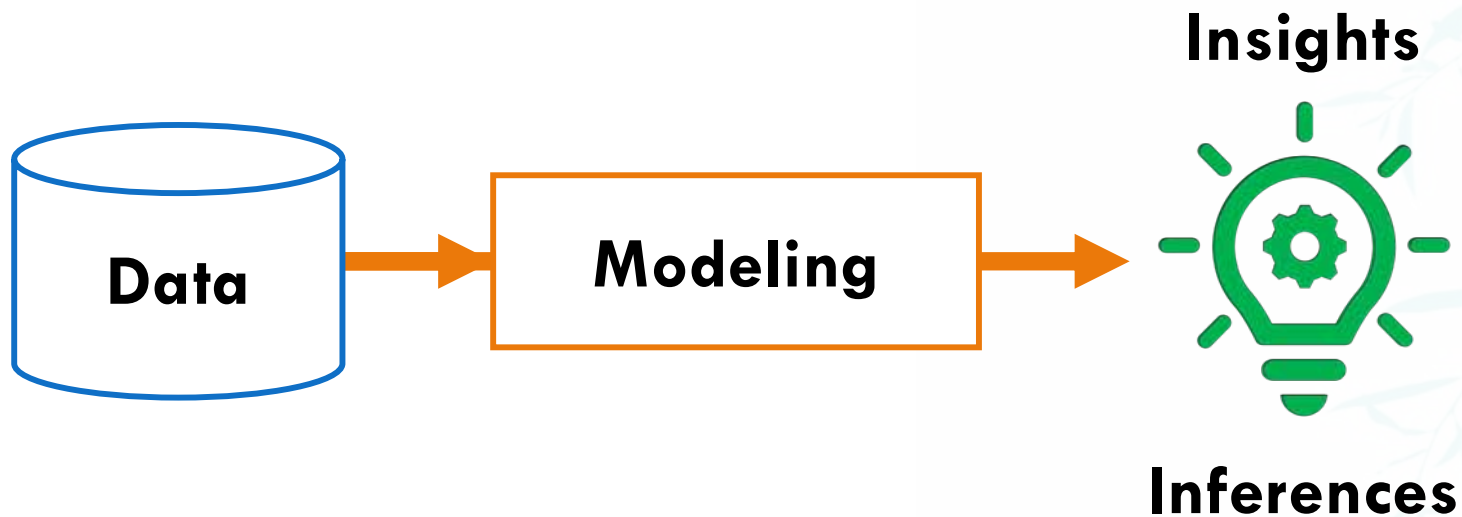
The ML Pipeline

CS3244 Machine Learning



Department of Computer Science
School of Computing

Machine Learning Pipeline



What do you think a ML Pipeline is?
What is its purpose?

Go to Slack, on the [#general](#) channel

Post an answer to our thread. A short 1-2 sentence or phrase will do.

If you'd prefer just to upvote another's answer, feel free to do that (too).

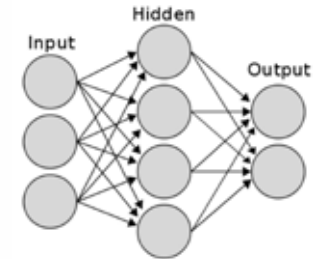
What is a model?

$$y = ax + b$$

Equation (Line)

If Rain, then Stay Home
Else Go Out

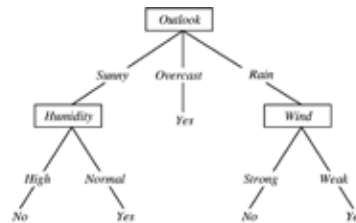
Code? Rules



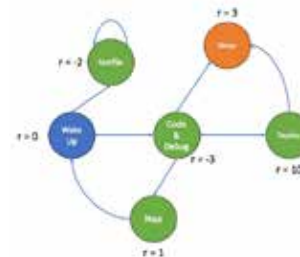
Neural Network (NN)

$$y = ax^2 + bx + c$$

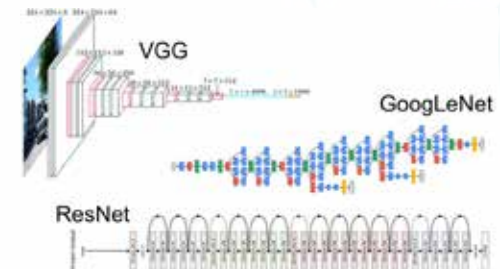
Equation (Nonlinear)



Decision Tree



Bayesian Graph



Deep NN

A **model** is a mathematical representation of a behavior

Machine Learning Pipeline

1. Data collection
2. Data extraction (Feature engineering)
3. Data understanding (with Visualization)
4. Data pre-processing
5. Model choice / design
6. Model training
7. Model validation (Evaluation)
8. Model understanding (Visualization / Explainability)
9. Model deployment



Machine Learning Pipeline

1. Data collection
2. Data extraction (Feature engineering)
3. **Data** understanding (with Visualization)
4. **Data** pre-processing
5. **Model** choice / design
6. **Model** training
7. **Model** validation (Evaluation)
8. **Model** understanding (Visualization / Explainability)
9. Model deployment



Machine Learning Pipeline

1. Data collection
2. Data extraction (Feature engineering)
3. Data understanding (with Visualization) [W08]
4. Data pre-processing [W08]
5. Model choice / design [W01–06, 10–12]
6. Model training [with Step 5]
7. Model validation (Evaluation) [W07]
8. Model understanding (Explainability [W10] / Visualization [W08])
9. Model deployment

“Machine learning”

Machine Learning Pipeline

5. Model choice / design [W01–06, 10–12]

- Paradigms of ML
 - Supervised Learning
 - Unsupervised Learning
 - ~~Reinforcement Learning~~
- Bias and Variance

Machine Learning Pipeline

5. Model choice / design

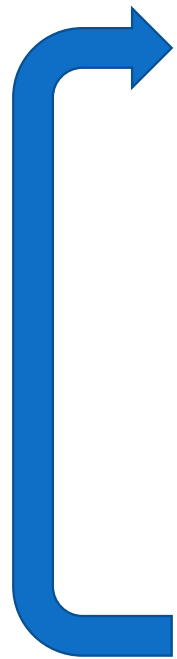
- Paradigms of ML [W02]
 - Supervised Learning [W02...]
 - Unsupervised Learning [W11]
 - ~~Reinforcement Learning~~
- Bias and Variance [W05]

Machine Learning Pipeline

5. Model choice / design

- Paradigms of ML [W02]
 - Supervised Learning
 - Decision trees [W03]
 - Linear models [W04]
 - Support Vector Machines [W04]
 - Perceptron and Neural Networks [W09]
 - Deep Learning [W10]
 - Unsupervised Learning [W11]
- Bias and Variance [W05]

(Iterative) Machine Learning Pipeline



1. Data collection
2. Data extraction (Feature engineering)
3. Data understanding (with Visualization)
4. Data pre-processing
5. Model choice / design
6. Model training
7. Model validation (Evaluation)
8. Model understanding (Visualization / Explainability)
9. Model deployment



Hands on with Colab

CS3244 Machine Learning



Department of Computer Science
School of Computing

Hands-On ML Model w/ Colab (30 minutes)

Let's Go!

<http://www.comp.nus.edu.sg/~cs3244/AY2122S1/01.colab.html>



Machine Learning

NUS SoC, 2021/2022, Semester I, Hybrid: Physically Mondays, 16:00-18:00 (i3 Auditorium) and Thursdays, 11:00-12:00 (LT15); Virtually on Zoom via LumiNUS Conferencing.

What did we learn?

We ran a ML classifier over a famous dataset, MNIST, for digit classification.

Procedurally, we also had a first try at:

- Using Google Colab.
- Using LumiNUS Quiz to submit your answers to in-Colab exercises.

Wrapping up Week 01

CS3244 Machine Learning



Department of Computer Science
School of Computing

What did we learn this week?

- Understand the rationale for our course's structure, the why of our module
- Recall the components of the course and their weightages and general deadlines
- Execute a basic machine learning workflow in Google Colab
- Conceptualize the division of labor in a typical machine learning project

Outlook for next week

Image credits <http://scott.fortmann-roe.com/docs/BiasVariance.html>

Assigned Readings (due before next Mon)

First, let's do some reading to set the stage:

- Paradigms of ML: <https://medium.datadriveninvestor.com/learning-paradigms-in-machine-learning-146ebf8b5943> (5 minutes)
- 5 Tribes of ML: <https://medium.com/dummykoders/5-tribes-of-ai-the-perspective-6b9e0334638c> (10 minutes)

Assigned Task (due before next Mon)

Let's intersect the two readings together!

Post a 1-2 sentence (not long!) answer to the below question in your Slack tutorial group [#tg-xx](#) or [#tg-na](#) (if you don't have a tutorial group yet).

Give an example of an ML application (a URL to it is good).
Describe which paradigm and tribe you think it belongs to.

[Don't worry if you're not sure, we'll cover this again in Week 02.]