

CloudAEye Webinar series

How to become a data-driven enterprise with Al

(episode 1)

Introduction: Who are we

CloudAEye

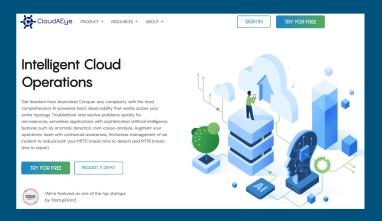
Al-driven observability for cloudnative applications

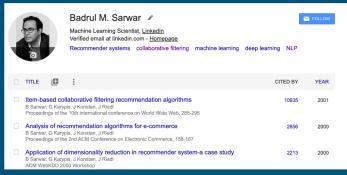
Presenter Bio

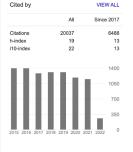
Co-Founder and CTO of CloudAEye

Al and recommender systems experience

Led the Al-academy at LinkedIn







Agenda

Organization of the webinar series

What to expect, topics covered, tools and contents provided

Data-driven Enterprise

Data-driven organizations and the role of AI

Introduction to Al concepts

AI/ML/DL/DS

Introduction to basics of ML: supervised and unsupervised learning

Supervised and unsupervised learning basics: Regression, classification and Clustering etc.

Code examples

Summary and QA

Organization of the Webinar series

Cadence: Bi-weekly meeting of 30 minutes

Duration: Approximately 15-20 episodes

Focus: Basic concepts of AI/ML and how organizations can utilize

them

Format: Interactive, presentation, code examples, Q/A, assignment

Resources: Github and colaboratory for code, slack-channel for discussion

Webinar series: Tentative Topics

- Introductory concepts
- Machine learning models: types of models, model fitting basics, supervised and unsupervised learning, regression and classification, evaluation
- Classification algorithms details: logistic regression, trees: bagging/boosting, SVMs, KNNs, Bayesian
- Time-series analysis, anomaly detection, forecasting

Practical and hands-on Example code and assignments

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pytorch Graph

Nanomaly fitting types reco algorithms

unsupervised analysis basics
logistic forecasting mage forward

Machine was detection

Machine was detection

Machine was detection

Time types reco algorithms

recurrent

Autoencoders was detection

trees NLP boosting

trees NLP boosting

trees NLP boosting

supervised details
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Webinar series: Tentative Topics (contd.)

- Introduction to deep learning: feed forward and recurrent neural networks, LSTMS, Autoencoders, Graph neural networks
- Personalization/recommendation: Matrix factorization, wide and deep models, GNN
- Practical applications: image processing,
 NLP, recommender systems, AlOps
- Transfer learning, self-supervised learning
- Emerging concepts: privacy, fairness, explainability

State-of-the-art models and techniques Practical large-scale examples

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pytorch Graph

Machine

Unsupervised Sologistic forecasting image forward

Works Model

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Unsupervised Sologistic forecasting image forward

Works Model

Works NLP boosting trees NLP boosting trees NLP boosting trees NLP supervised details
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Data-driven Enterprises

Data-driven culture

- We are submerged in data (Big data)
- Data can be leveraged to make better decisions, gauge and improve customer sentiment, streamline operations etc.

What are the traits of a data-driven enterprise?

- Fosters a data-driven culture starting from the top
- Metrics are carefully chosen
- Employees are encouraged to be data-centric and trained on latest tech
- Provides a robust infrastructure for collecting, analyzing and serving

Data-driven Enterprises: Challenges

Building a Data-driven culture is hard

- Cultural shift is harder than you think
- It's not about the technology it's about the people

Barriers of becoming a data-driven enterprise

- Shortage of talent–leadership to worker
- Sheer volume of data and lack of infrastructure
- Lack of clean usable data
- Concerns about privacy

How to overcome these challenges

There is no short term solution, it takes time

Think different

There is no shortage of analytic algorithms. These need to be matched by critical thinking, human judgement, and a view to creative innovation.

Fail fast, learn faster

Companies that are prepared for faster iterative learning — fail fast, learn faster — will gain insight and knowledge before their competitors.

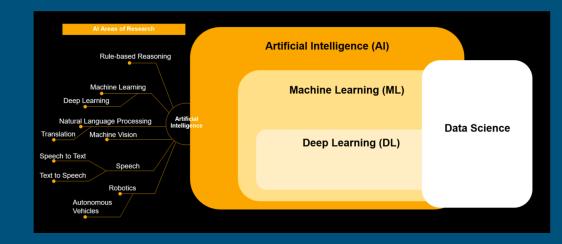
Focus on the long-term

Becoming data-driven is a process. It takes time

AI, ML, DL, DS Demystified

Al: Intelligence exhibited by machines. Broadly defined to include any simulation of human intelligence

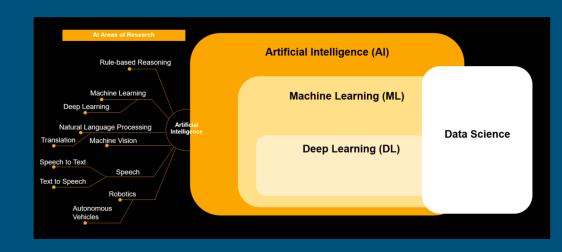
ML: Subfield of AI that aims to teach computers the ability to do tasks with data, without explicit programming



AI, ML, DL, DS Demystified

DL: Subfield of ML that uses specialized techniques involving multi-layer (2+) artificial neural networks

DS: Scientific methods, algorithms and systems to extract knowledge or insights from big data



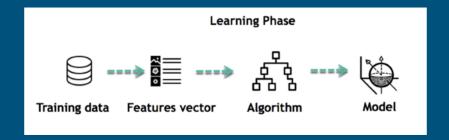
For this webinar series we will focus mostly on ML/DL algorithms

Machine learning algorithms

Machine learning algorithms aim to build a model or hypothesis based on the input data

Depending on the type on input ML algorithms can be

Supervised or unsupervised



Supervised ML algorithms

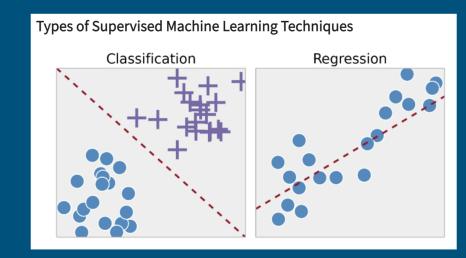
When the data is "labeled" or "tagged" for the correct outcome or value:

Input looks like (X, y)

Where X is the set of input features and y is the outcome variable that is provided

Based on y we have two classes of supervised learning:

- 1. Regression, where y is continuous ex: linear regression, tree-based regression, SVR
- 2. Classification, where y is discrete. ex: logistic regression, decision trees, SVM, KNN, deep networks



Unsupervised ML algorithms

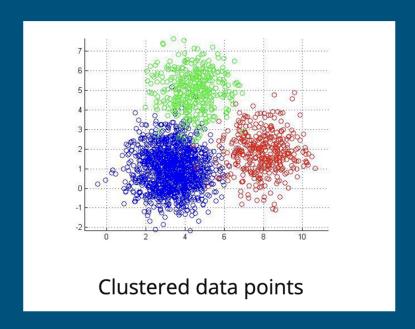
When the data is NOT "labeled" or "tagged"

Input looks like (X)

Where X is the set of input features

Types of unsupervised learning:

Clustering, density estimation, dimensionality reduction, representation learning, auto-encoders



Other types of ML

1. Semi-supervised learning

Some labeled data and mostly unlabeled data are provided, the algorithm uses the labeled data to reason about the unlabeled data

1. Self-supervised learning

Common-sense learning using lots of unsupervised data e.g., text, images

Reinforcement learning

Learning by trials and errors, given a state, a set of actions and reward function these algorithms learn to achieve intelligence through exploration

Coding example

Link to the colab Notebook

https://colab.research.google.com/drive/15xzHZU9G2eiJlyhabFqPFlvAAWki68l 1?usp=sharing

Thank You

You have DevOps pain points?

We like to hear from you

Looking for exciting opportunities?

We are hiring

scientists

AI/ML engineers and applied

Cloud Engineers

UI/UX designers

