Tamkang University

淡江大學

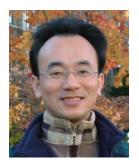


Big Data Mining

巨量資料探勘

Course Orientation for Big Data Mining (巨量資料探勘課程介紹)

1072DM01 MI4 (M2244) (2849) Wed 6, 7 (13:10-15:00) (B206)



<u>Min-Yuh Day</u> <u>戴敏育</u> Assistant Professor 專任助理教授

Dept. of Information Management, Tamkang University 淡江大學 咨訊答理學系

淡江大學 資訊管理學系



http://mail. tku.edu.tw/myday/ 2019-02-20



- 課程名稱: 巨量資料探勘 (Big Data Mining)
- 授課教師: 戴敏育 (Min-Yuh Day)
- 開課系級:資管四P(TLMXB4P)(M2244)(2849)
- 開課資料: 選修 單學期 2 學分 (2 Credits, Elective)
- 上課時間:週三 6,7 (Wed 13:10-15:00)
- 上課教室: B206

課程簡介

- 本課程介紹巨量資料探勘 (Big Data Mining) 的 基礎概念及應用技術。
- 課程內容包括
 - 巨量資料探勘 (Big Data Mining)
 - AI人工智慧與大數據分析 (Artificial Intelligence and Big Data Analytics)
 - 關連分析 (Association Analysis)
 - 分類與預測 (Classification and Prediction)
 - 分群分析 (Cluster Analysis)
 - 機器學習與深度學習 (Machine Learning and Deep Learning)
 - SAS企業資料採礦實務 (SAS Enterprise Miner)
 - 巨量資料探勘個案分析與實作

Course Introduction

- This course introduces the fundamental concepts and applications technology of big data mining.
- Topics include
 - Big Data Mining
 - Artificial Intelligence and Big Data Analytics
 - Association Analysis
 - Classification and Prediction
 - Cluster Analysis
 - Machine Learning and Deep Learning
 - Data Mining Using SAS Enterprise Miner (SAS EM)
 - Case Study and Implementation of Big Data Mining



• 瞭解及應用巨量資料探勘基本概念與技術。

 Understand and apply the fundamental concepts and technology of big data mining

課程大綱 (Syllabus)

週次(Week) 日期(Date) 內容(Subject/Topics)

- 1 2019/02/20 巨量資料探勘課程介紹 (Course Orientation for Big Data Mining)
- 2 2019/02/27 AI人工智慧與大數據分析 (Artificial Intelligence and Big Data Analytics)
- 3 2019/03/06 分群分析 (Cluster Analysis)
- 4 2019/03/13 個案分析與實作一(SAS EM 分群分析): Case Study 1 (Cluster Analysis - K-Means using SAS EM)
- 5 2019/03/20 關連分析 (Association Analysis)
- 6 2019/03/27 個案分析與實作二 (SAS EM 關連分析): Case Study 2 (Association Analysis using SAS EM)
- 7 2019/04/03 教學行政觀摩日 (Off-campus study)
- 8 2019/04/10 分類與預測 (Classification and Prediction)

課程大綱 (Syllabus)

週次(Week) 日期(Date) 內容(Subject/Topics)

- 9 2019/04/17 期中報告 (Midterm Project Presentation)
- 10 2019/04/24 期中考試週 (Midterm Exam)
- 11 2019/05/01 個案分析與實作三 (SAS EM 決策樹、模型評估): Case Study 3 (Decision Tree, Model Evaluation using SAS EM)
- 12 2019/05/08 個案分析與實作四 (SAS EM 迴歸分析、類神經網路): Case Study 4 (Regression Analysis, Artificial Neural Network using SAS EM)
- 13 2019/05/15 機器學習與深度學習
(Machine Learning and Deep Learning)
- 14 2019/05/22 期末報告 (Final Project Presentation)
- 15 2019/05/29 畢業考試週(Final Exam)

教學方法與評量方法

- 教學方法
 - 講述、討論、賞析、模擬、實作、問題解決
- 評量方法
 - -紙筆測驗、實作、報告、上課表現

教材課本

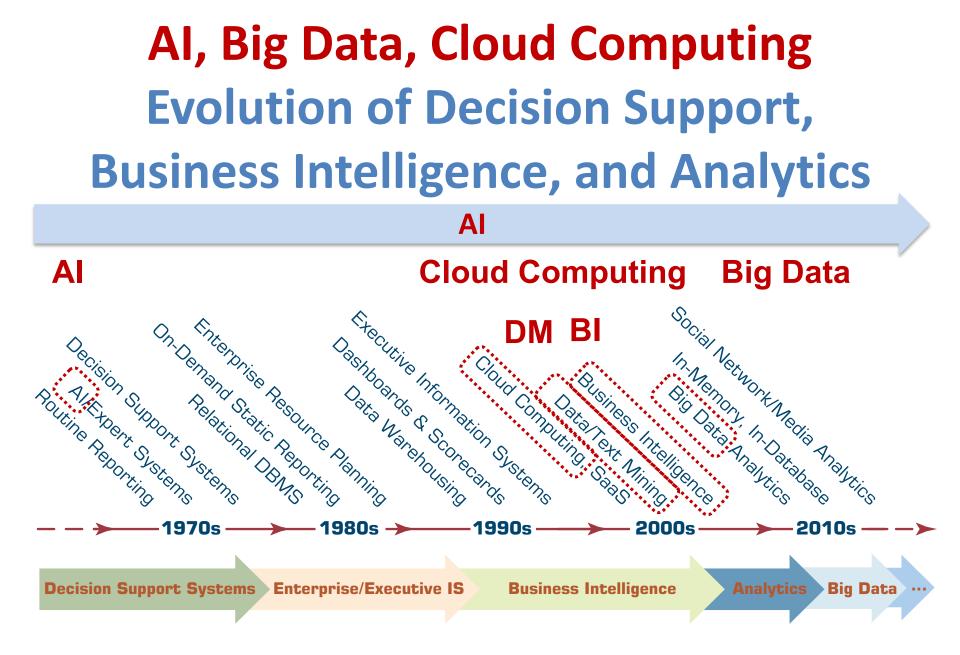
- 教材課本
 - 講義 (Slides)
 - 一資料採礦運用:以SAS Enterprise Miner為工具,
 李淑娟,2015,SAS賽仕電腦軟體
- 參考書籍
 - Big Data, Data Mining, and Machine Learning: Value Creation for Business Leaders and Practitioners, Jared Dean, Wiley, 2014
 - Data Science for Business: What you need to know about data mining and data-analytic thinking, Foster Provost and Tom Fawcett, O'Reilly, 2013
 - Applied Analytics Using SAS Enterprise Mining, Jim Georges, Jeff Thompson and Chip Wells, SAS, 2010
 - Data Mining: Concepts and Techniques, Third Edition, Jiawei Han,
 Micheline Kamber and Jian Pei, Morgan Kaufmann, 2011
 - Learning Data Mining with Python Second Edition, Robert Layton, Packt Publishing, 2017

作業與學期成績計算方式

- 作業篇數
 - -3篇
- 學期成績計算方式
 - 🗹 期 中 評 量: 30 %
 - 🗹 期末評量:30 %
 - ☑其他(課堂參與及報告討論表現):40 %

Team Term Project

- Term Project Topics
 - Big Data mining
 - Big Data Analytics
 - Business Intelligence
 - FinTech
- 3-4 人為一組
 - 分組名單於 2019/02/27 (三) 課程下課時繳交
 - 由班代統一收集協調分組名單



Data Mining Is a Blend of Multiple Disciplines



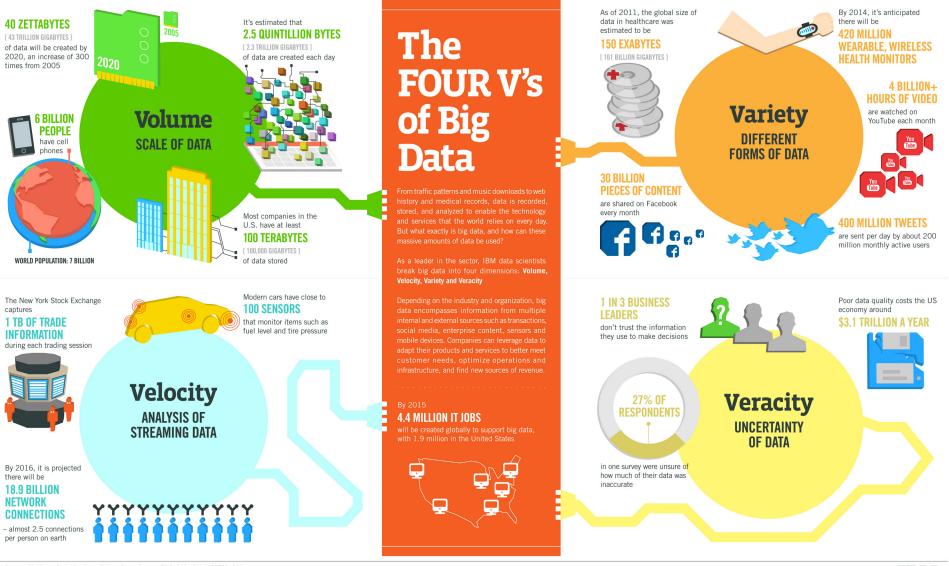
Source: Ramesh Sharda, Dursun Delen, and Efraim Turban (2017), Business Intelligence, Analytics, and Data Science: A Managerial Perspective, 4th Edition, Pearson

Data Mining Tasks & Methods

Data Mining Tasks & Methods	Data Mining Algorithms	Learning Type
Prediction		
Classification	Decision Trees, Neural Networks, Support Vector Machines, kNN, Naïve Bayes, GA	Supervised
Regression	Linear/Nonlinear Regression, ANN, Regression Trees, SVM, kNN, GA	Supervised
Time series	Autoregressive Methods, Averaging Methods, Exponential Smoothing, ARIMA	Supervised
Association		
Market-basket	Apriori, OneR, ZeroR, Eclat, GA	Unsupervised
Link analysis	Expectation Maximization, Apriori Algorithm, Graph-Based Matching	Unsupervised
Sequence analysis	Apriori Algorithm, FP-Growth, Graph-Based Matching	Unsupervised
Segmentation		
Clustering	k-means, Expectation Maximization (EM)	Unsupervised
Outlier analysis	k-means, Expectation Maximization (EM)	Unsupervised

Big Data Analytics and **Data Mining**

Big Data 4 V



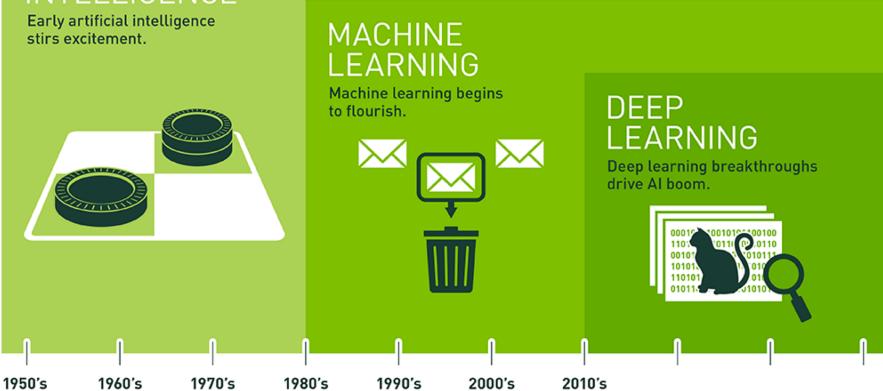
Sources: McKinsey Global Institute, Twitter, Cisco, Gartner, EMC, SAS, IBM, MEPTEC, QAS

TRM



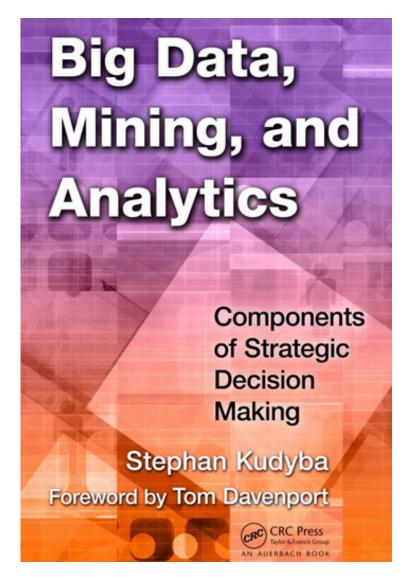
Artificial Intelligence Machine Learning & Deep Learning

ARTIFICIAL INTELLIGENCE



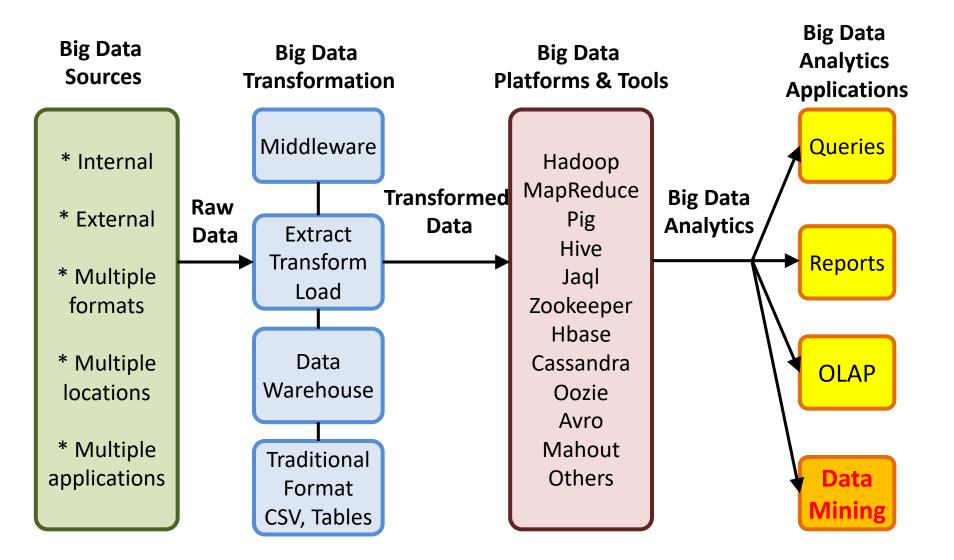
Since an early flush of optimism in the 1950s, smaller subsets of artificial intelligence – first machine learning, then deep learning, a subset of machine learning – have created ever larger disruptions.

Stephan Kudyba (2014), Big Data, Mining, and Analytics: Components of Strategic Decision Making, Auerbach Publications

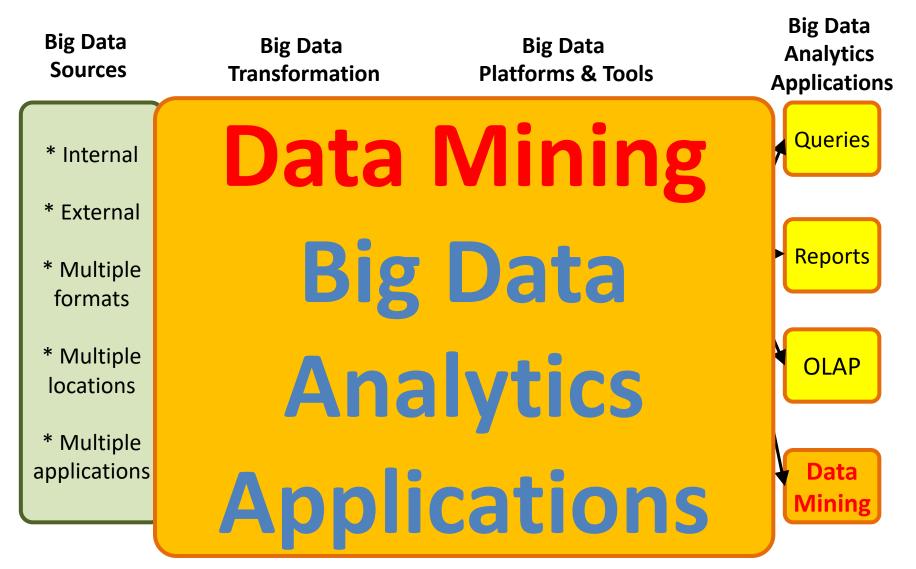


Source: http://www.amazon.com/gp/product/1466568704

Architecture of Big Data Analytics

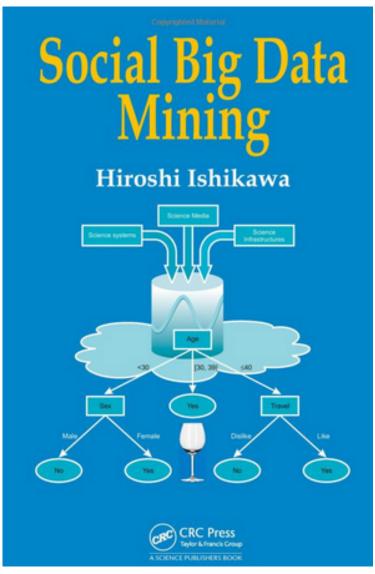


Architecture of Big Data Analytics



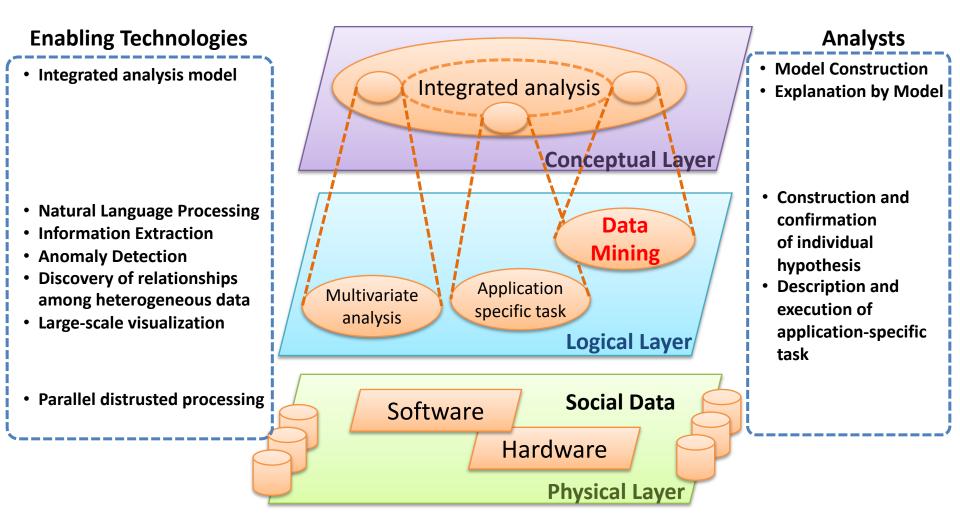
Social Big Data Mining

(Hiroshi Ishikawa, 2015)

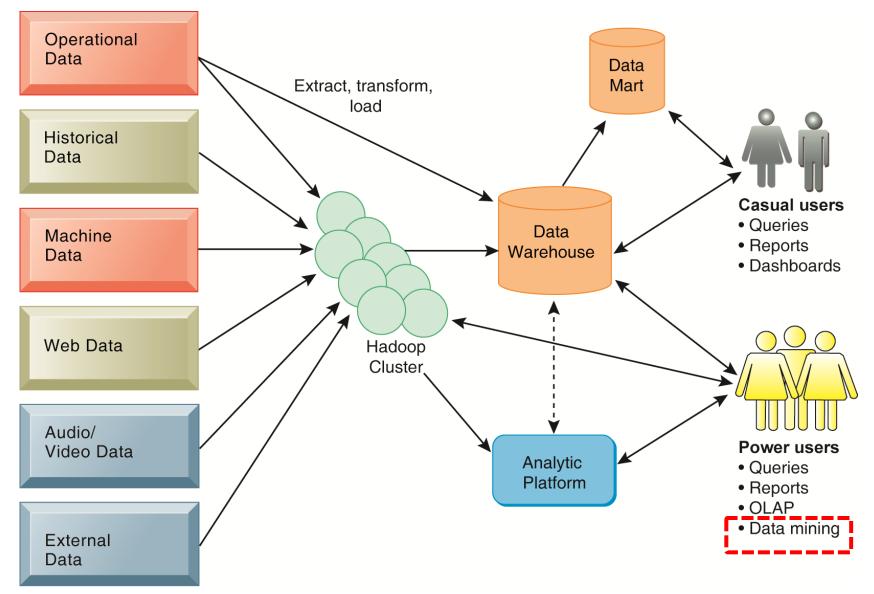


Architecture for Social Big Data Mining

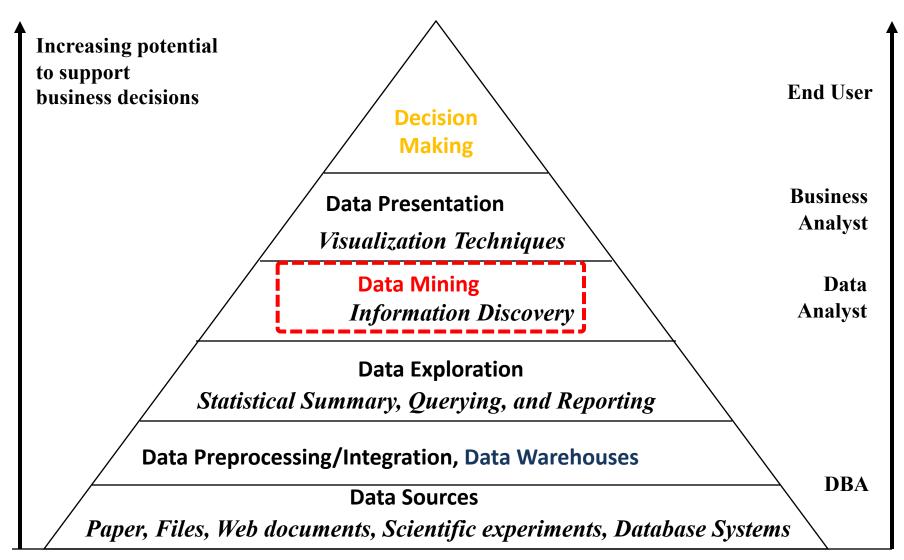
(Hiroshi Ishikawa, 2015)



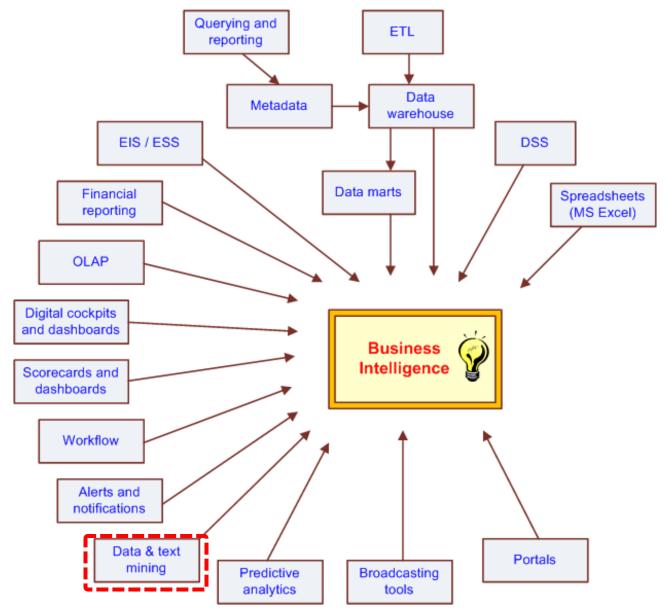
Business Intelligence (BI) Infrastructure



Data Warehouse Data Mining and Business Intelligence

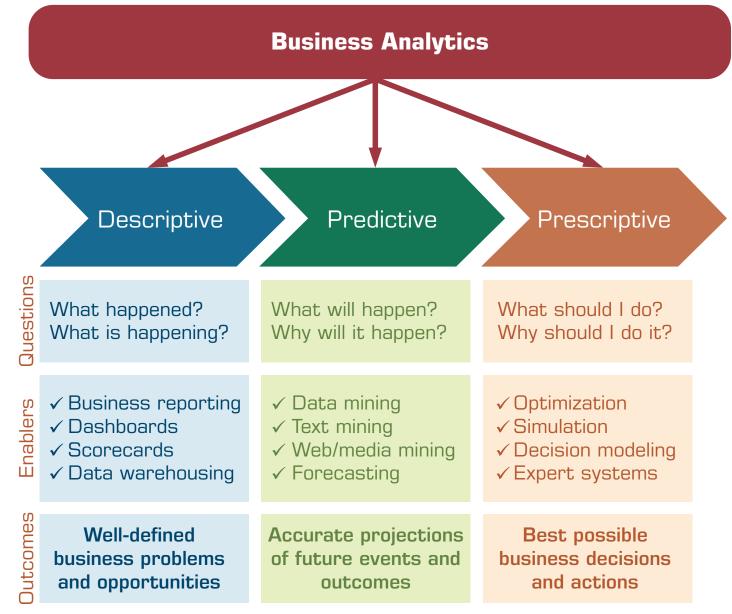


The Evolution of BI Capabilities



Source: Turban et al. (2011), Decision Support and Business Intelligence Systems

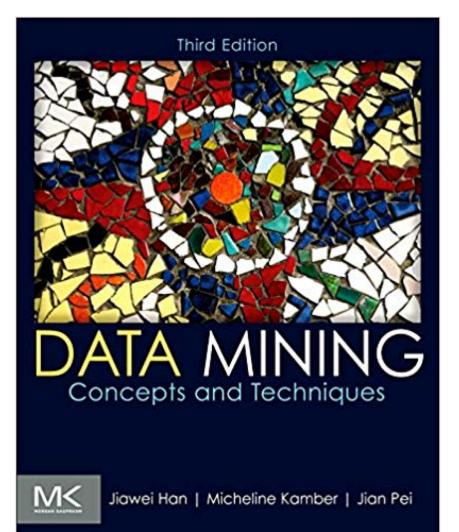
Three Types of Analytics



Source: Ramesh Sharda, Dursun Delen, and Efraim Turban (2017),

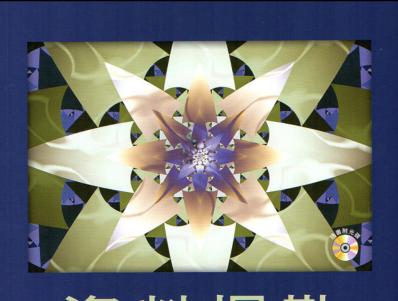
Business Intelligence, Analytics, and Data Science: A Managerial Perspective, 4th Edition, Pearson

Data Mining: Concepts and Techniques, Third Edition, Jiawei Han, MichelineKamber and Jian Pei, Morgan Kaufmann, 2011



28

郝沛毅,李御璽,黃嘉彦編譯,資料探勘 (Jiawei Han, Micheline Kamber, Jian Pei, Data Mining - Concepts and Techniques 3/e), _{高立圖書}, 2014



資料探勘 DATA MINING Concepts and Techniques 3/e

Jiawei Han・Micheline Kamber・Jian Pei 原著 郝沛毅 李御璽 黃嘉彦 編譯

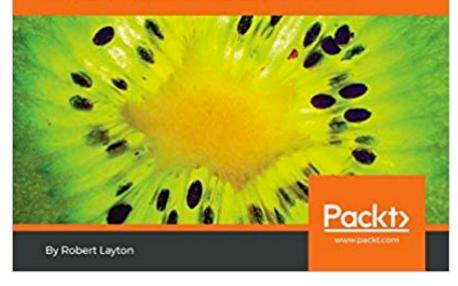
ELSEVIER TAIWAN LLC · 高立圖書 合作出版

Learning Data Mining with Python - Second Edition, Robert Layton, Packt Publishing, 2017



Second Edition

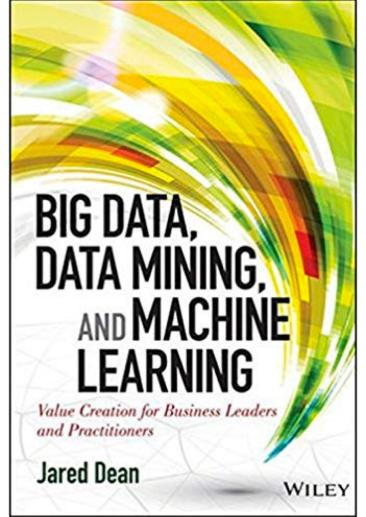
Use Python to manipulate data and build predictive models



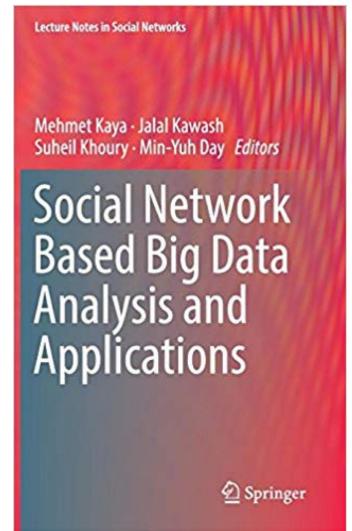
Source: https://www.amazon.com/Learning-Data-Mining-Python-Second/dp/1787126781

Big Data, Data Mining, and Machine Learning: Value Creation for Business Leaders and Practitioners, Jared Dean,

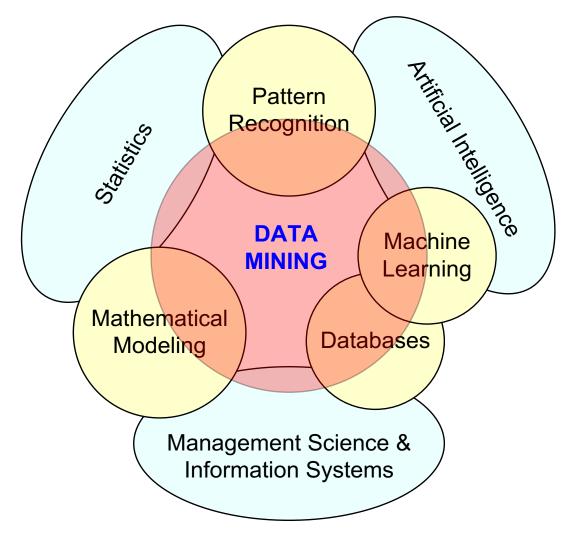
Wiley, 2014.



Social Network Based Big Data Analysis and Applications, Lecture Notes in Social Networks, Mehmet Kaya, Jalal Kawash, Suheil Khoury, Min-Yuh Day, Springer International Publishing, 2018.



Data Mining at the Intersection of Many Disciplines



Source: Turban et al. (2011), Decision Support and Business Intelligence Systems





Data Mining: Core Analytics Process

The KDD Process for Extracting Useful Knowledge from Volumes of Data

Source: Fayyad, U., Piatetsky-Shapiro, G., & Smyth, P. (1996). The KDD Process for Extracting Useful Knowledge from Volumes of Data. Communications of the ACM, 39(11), 27-34.

Fayyad, U., Piatetsky-Shapiro, G., & Smyth, P. (1996). The KDD Process for **Extracting Useful Knowledge** from Volumes of Data. Communications of the ACM, 39(11), 27-34.

Knowledge Discovery in Databases creates the context for developing the tools needed to control the flood of data facing organizations that depend on ever-growing databases of business, manufacturing, scientific, and personal information.

The KDD Process for Extracting Useful Knowledge from Volumes of Data

As we march into the age of digital information, the problem of data overload looms ominously ahead. Our ability to analyze and Gregory Piatetsky-Shapiro, understand massive datasets lags far behind our ability to gather and store the data. A new gen-

and Padhraic Smyth

Usama Fayyad,

eration of computational techniques and many more applications generate the rapidly growing volumes of data. data warehouses. These techniques and tools are the Current hardware and database tech-



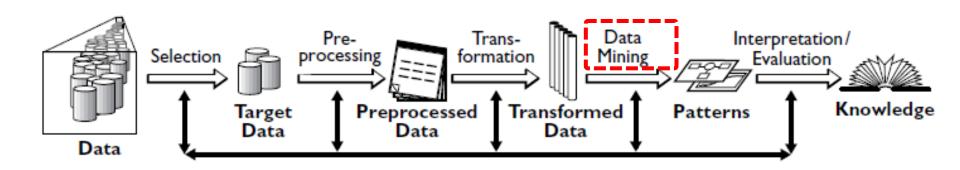
and tools is required to support the streams of digital records archived in extraction of useful knowledge from huge databases, sometimes in so-called

subject of the emerging field of knowl- nology allow efficient and inexpensive edge discovery in databases (KDD) and reliable data storage and access. However er, whether the context is business Large databases of digital informa- medicine, science, or government, the tion are ubiquitous. Data from the datasets themselves (in raw form) are of neighborhood store's checkout regis- liule direct value. What is of value is the ter, your bank's credit card authoriza- knowledge that can be inferred from tion device, records in your doctor's the data and put to use. For example, office, patterns in your telephone calls, the marketing database of a consumer

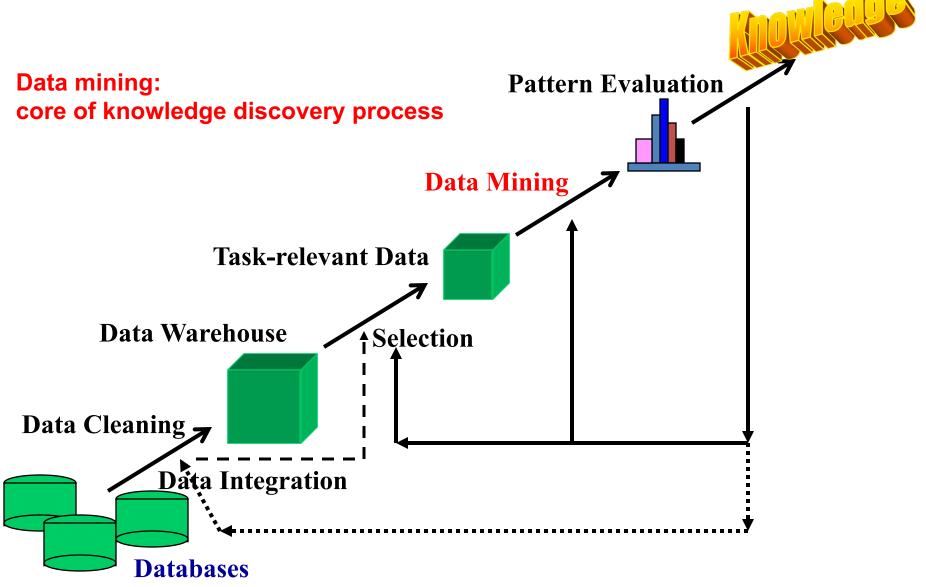
Data Mining

Knowledge Discovery in Databases (KDD) Process

(Fayyad et al., 1996)

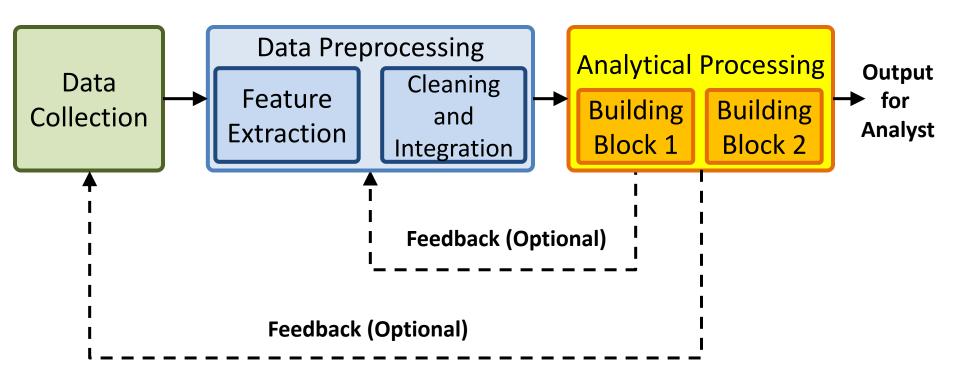


Knowledge Discovery (KDD) Process



Data Mining Processing Pipeline

(Charu Aggarwal, 2015)



Source: Charu Aggarwal (2015), Data Mining: The Textbook Hardcover, Springer

BIG DATA, DATA MINING, AND MACHINE LEARNING

Value Creation for Business Leaders and Practitioners

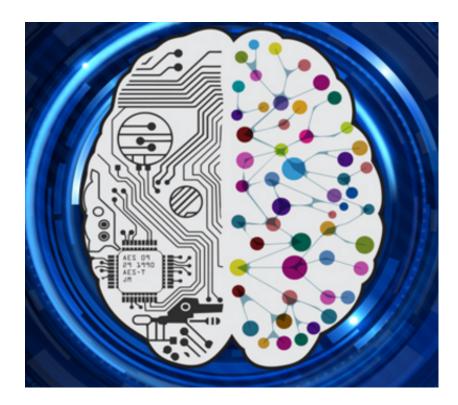


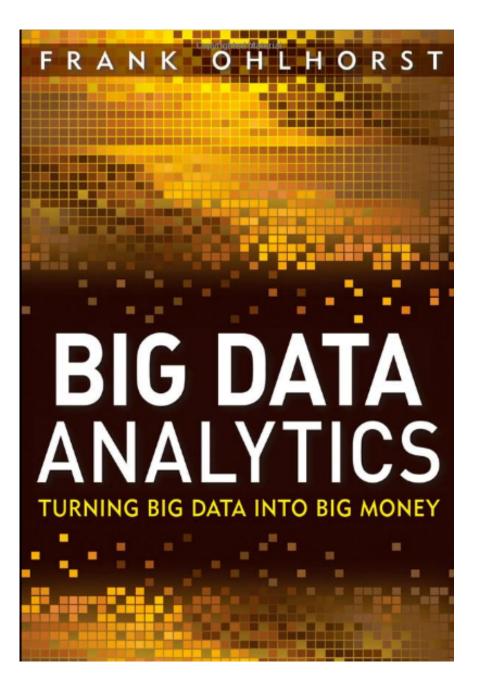
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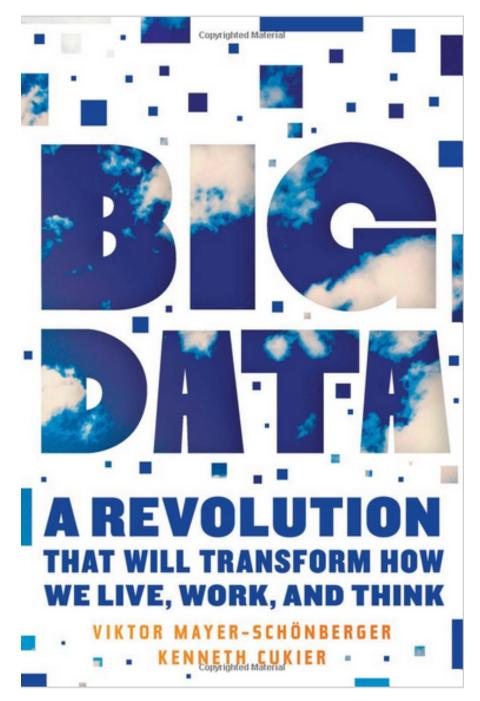
Source: http://www.amazon.com/Data-Mining-Machine-Learning-Practitioners/dp/1118618041

Deep Learning Intelligence from Big Data

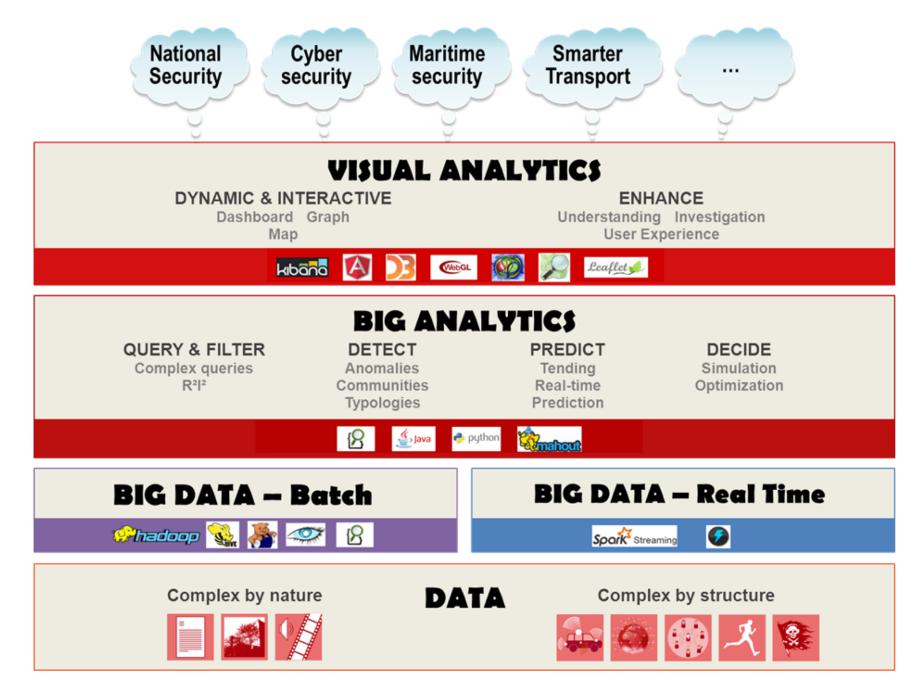




Source: http://www.amazon.com/Big-Data-Analytics-Turning-Money/dp/1118147596

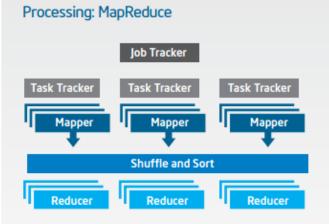


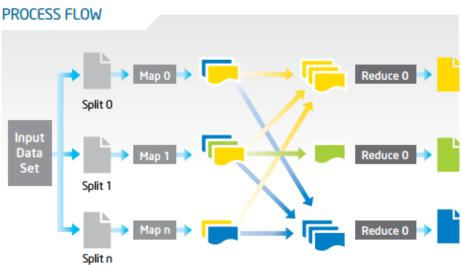
Source: http://www.amazon.com/Big-Data-Revolution-Transform-Mayer-Schonberger/dp/B00D81X2YE



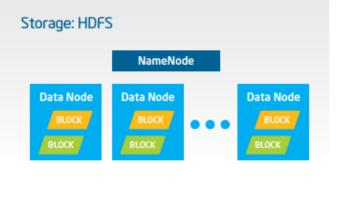
Big Data with Hadoop Architecture

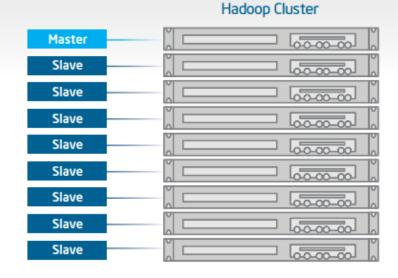
LOGICAL ARCHITECTURE



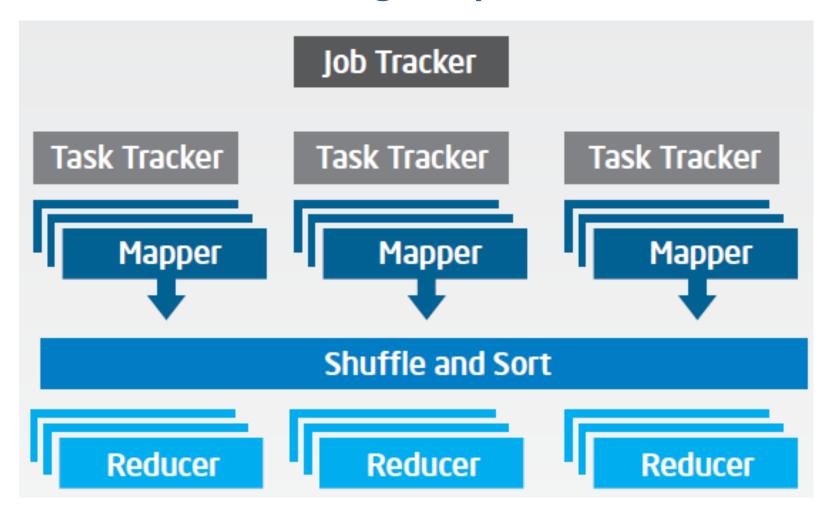


PHYSICAL ARCHITECTURE

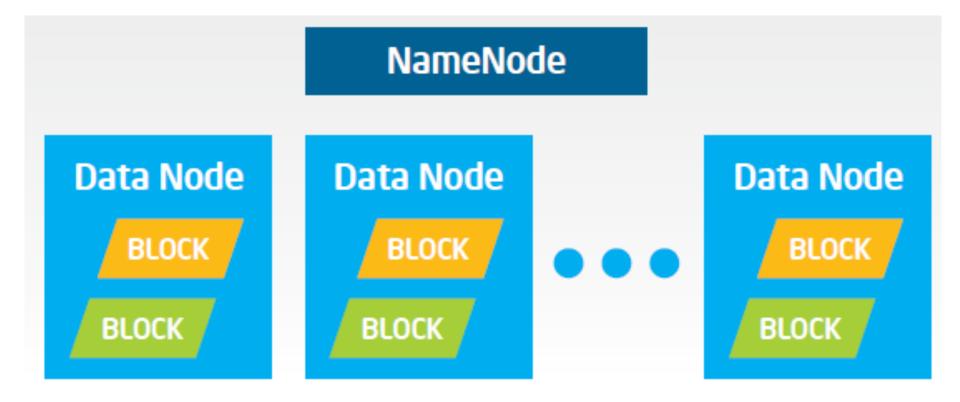




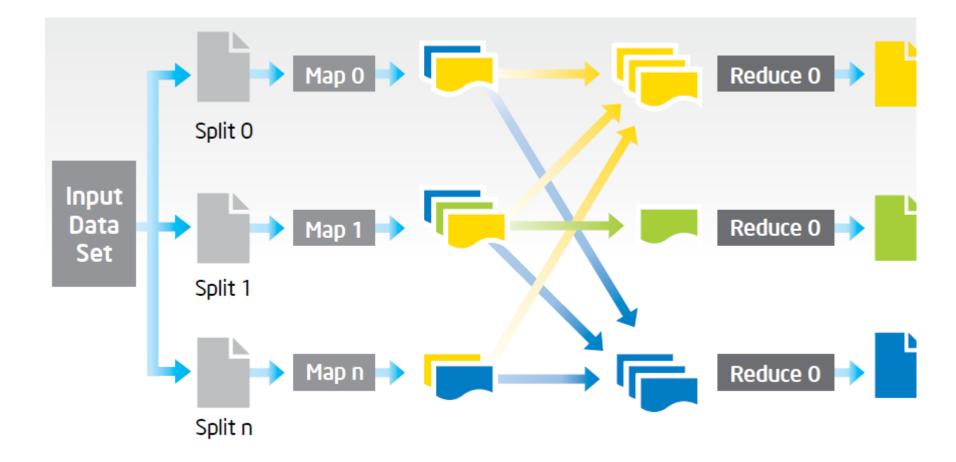
Big Data with Hadoop Architecture Logical Architecture Processing: MapReduce



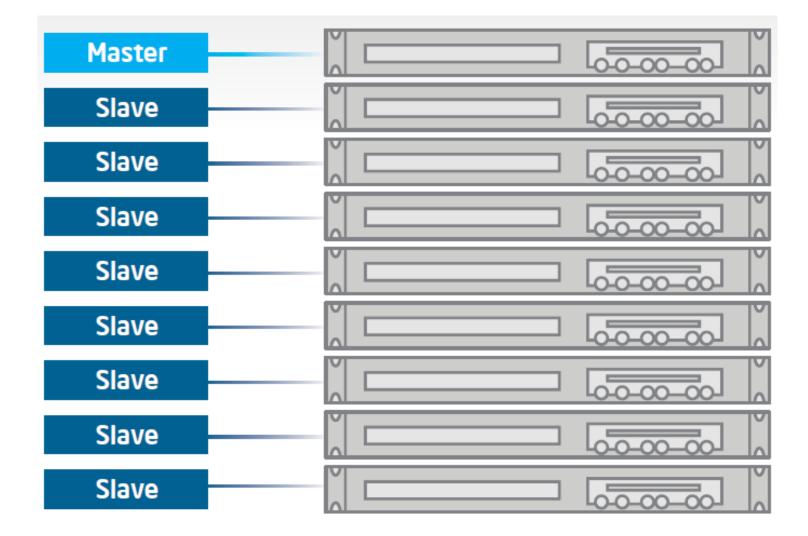
Big Data with Hadoop Architecture Logical Architecture Storage: HDFS



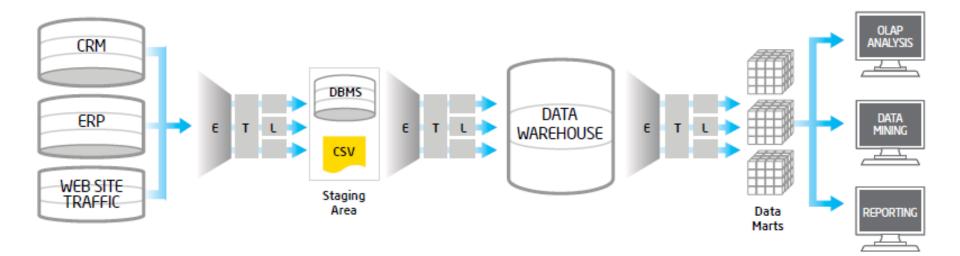
Big Data with Hadoop Architecture Process Flow



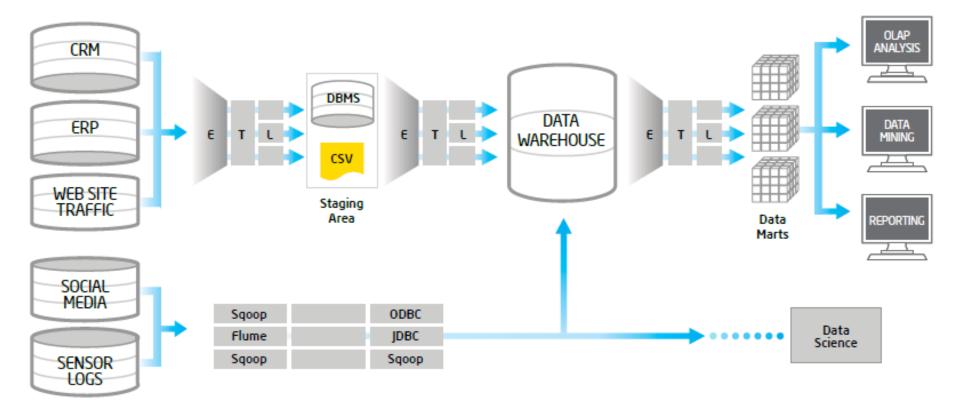
Big Data with Hadoop Architecture Hadoop Cluster



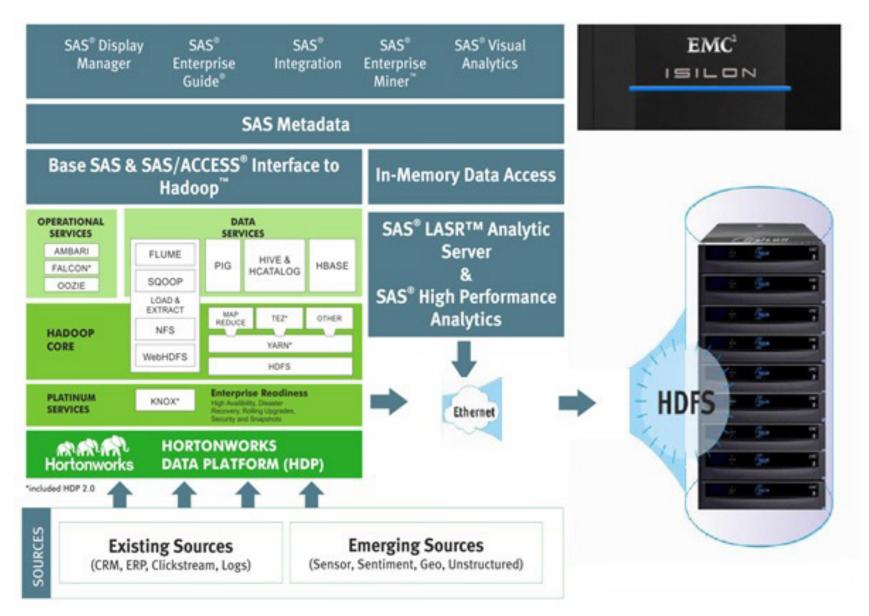
Traditional ETL Architecture



Offload ETL with Hadoop (Big Data Architecture)



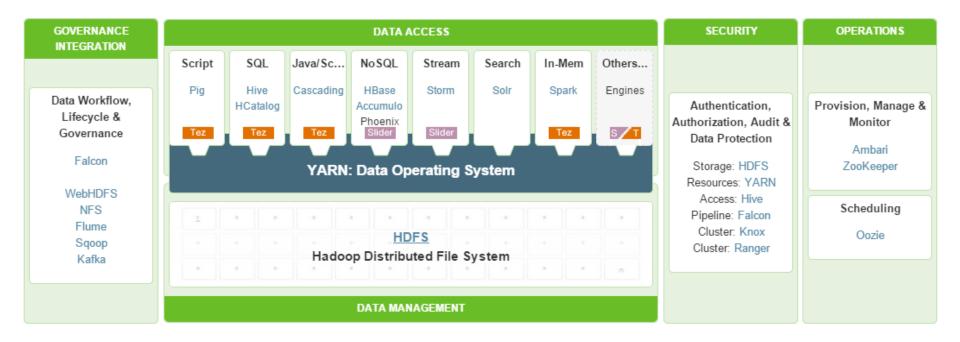
Big Data Solution



Source: http://www.newera-technologies.com/big-data-solution.html

HDP

A Complete Enterprise Hadoop Data Platform



Spark and Hadoop











Spark Ecosystem

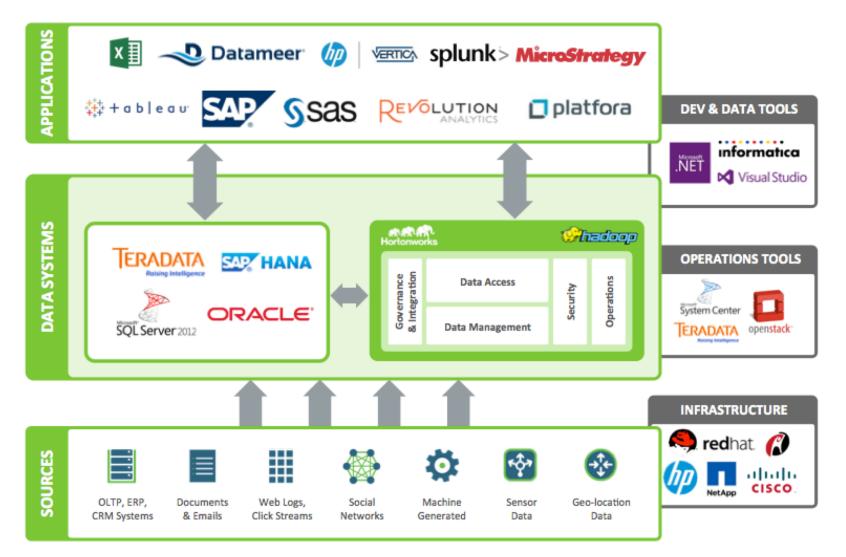
Spark SQL Streaming

MLlib (machine learning)

GraphX (graph)

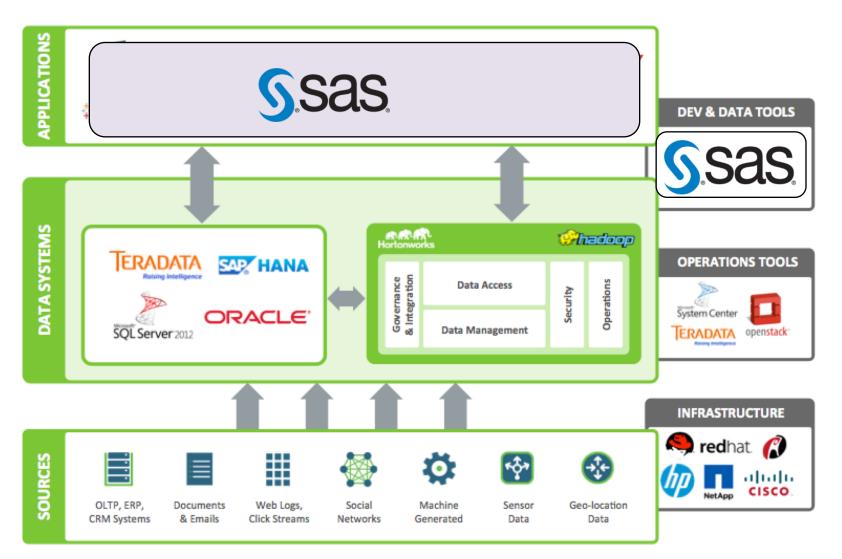
Apache Spark

SAS Big data Strategy - SAS areas



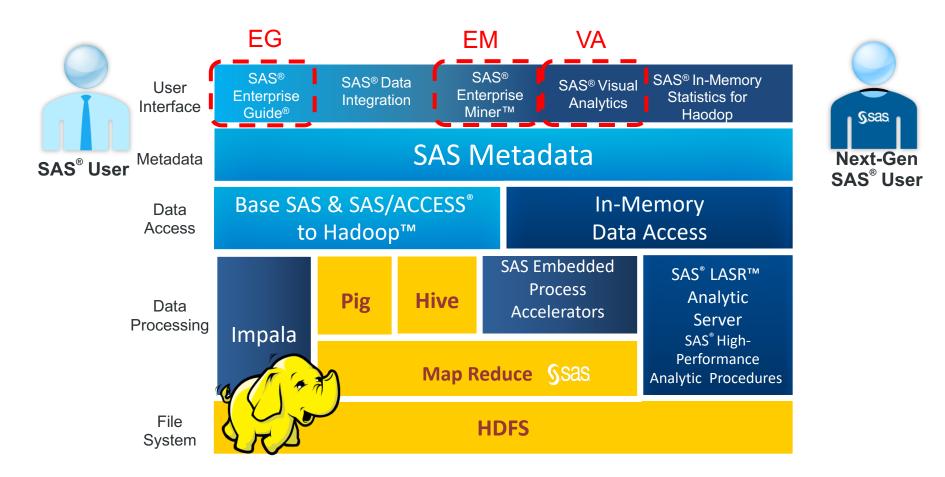
Source: Deepak Ramanathan (2014), SAS Modernization architectures - Big Data Analytics

SAS Big data Strategy - SAS areas



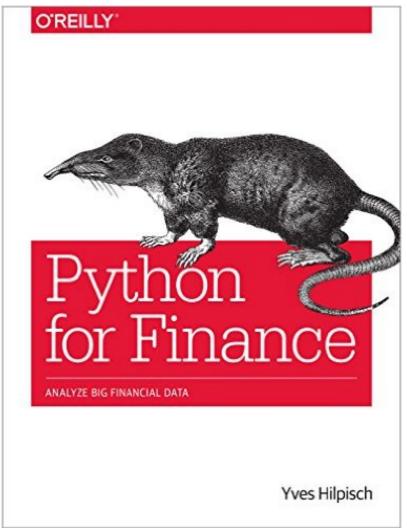
Source: Deepak Ramanathan (2014), SAS Modernization architectures - Big Data Analytics

SAS[®] Within the HADOOP ECOSYSTEM



Source: Deepak Ramanathan (2014), SAS Modernization architectures - Big Data Analytics

Yves Hilpisch, Python for Finance: Analyze Big Financial Data, O'Reilly, 2014



Business Intelligence Trends

- 1. Agile Information Management (IM)
- 2. Cloud Business Intelligence (BI)
- 3. Mobile Business Intelligence (BI)
- 4. Analytics
- 5. Big Data

Business Intelligence Trends: Computing and Service

- Cloud Computing and Service
- Mobile Computing and Service
- Social Computing and Service

Business Intelligence and Analytics

- Business Intelligence 2.0 (BI 2.0)
 - Web Intelligence
 - Web Analytics
 - Web 2.0
 - Social Networking and Microblogging sites
- Data Trends
 - Big Data
- Platform Technology Trends

- Cloud computing platform

Source: Lim, E. P., Chen, H., & Chen, G. (2013). Business Intelligence and Analytics: Research Directions. ACM Transactions on Management Information Systems (TMIS), 3(4), 17

Business Intelligence and Analytics: Research Directions

- **1. Big Data Analytics**
 - Data analytics using Hadoop / MapReduce framework
- 2. Text Analytics
 - From Information Extraction to Question Answering
 - From Sentiment Analysis to Opinion Mining
- 3. Network Analysis
 - Link mining
 - Community Detection
 - Social Recommendation

Source: Lim, E. P., Chen, H., & Chen, G. (2013). Business Intelligence and Analytics: Research Directions. ACM Transactions on Management Information Systems (TMIS), 3(4), 17

Big Data & Al Landscape 2018

INFRASTRUCTURE	ANALYTICS	APPLICATIONS – ENTERPRISE
HADOOP ON-PREMISE Cloudera Hotomworks WNPR Pivotal IBM InfoSphere' BigInsighter bluedata' jethro HADOOP IN THE CLOUD WS Microsoft Azure Coogle Cloud BigInsighter Coogle Cloud	DATA ANALYST PLATFORMS Microsoft @ pentoho alteryx @ pentoho alteryx @ guAVUS AYASDI ATTIV/O Datameer Quid Incorta. inter ana. ClearStory Origoni @ ENDOR MODE Bottlense @chabased	SALES MARKETING - B28 RADIUS App Aixie INSIDE SALES.COM Clari A aviso tact.ai fuse (machines () TROOPS MARKETING - B28 RADIUS App Aixie MARKETING - B28 RADIUS App Aixie Lattice MARKETING - B22 Zela () bioomreach "SendGrid BlueYonder (PERSADO] Image () CUSTOMER SERVICE Marketing - BLUECORE () CUSTOMER SERVICE MARKETING - B28 Zela () bioomreach "SendGrid BlueYonder (PERSADO] Image () CUSTOMER SERVICE Clari A aviso tact.ai fuse (machines () TROOPS MARKETING - B28 MARKETING - B28 MARKETING - B28 MARKETING - B28 Zela () bioomreach "SendGrid BlueYonder (PERSADO] Image () CUSTOMER SERVICE CLARABRIDGE () MARKETING - B28 MARKETING - B28
NoSOL DATABASES Google Cloud CRACLE Microsoft Azure mongoDB MarkLogic AtmageDB Couchbase redisidbs SCYLLA.	BIPLATFORMS VISUALIZATION Microsoft WS Wise Analysis Google Cloud Löcker Google Cloud A T S C A LE Olik Q Consumer Consumer Olik Q Microsoft Distance Microsoft Olik Q Distance Distance Microsoft Distance Consumer Distance Microsoft Distance Visualization Distance Consumer Distance Consumer Distance Visualization Distance Visualization Distance Visualization Visualization Microsoftranegy Distance Visualization Provisition	HUMAN CAPITAL HUMAN
DATA TRANSFORMATION atterny @ TRIATA () torm @ Paxata () torm	COMPUTER VISION Microsoft Azure American Satisgrafica contraint	APPEICATIONS - INDUSTRY Appeirosas Criteol. Part distance Criteol. Part dis
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	ATA SOURCES & APIs JONES JON	Sector Qualifics QPalantir QPalant

Final 2018 version, updated 07/15/2018

© Matt Turck (@mattturck), Demi Obayomi (@demi_obayomi), & FirstMark (@firstmarkcap)

mattturck.com/bigdata2018

FIRSTMARK

Summary

- This course introduces the fundamental concepts and applications technology of big data mining.
- Topics include
 - Big Data Mining
 - Artificial Intelligence and Big Data Analytics
 - Association Analysis
 - Classification and Prediction
 - Cluster Analysis
 - Machine Learning and Deep Learning
 - Data Mining Using SAS Enterprise Miner (SAS EM)
 - Case Study and Implementation of Big Data Mining

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