数据科学好帮手 IBM数据科学及机器学习平台揭秘和案例分享

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Agenda 大纲

- ◆ 数据科学和机器学习概要
 - Data Science 101
 - Machine Learning 101
 - Data Science and ML Challenges
- ◆ IBM 数据科学平台介绍
 - IBM Data Science Experience
 - IBM Machine Learning
- ◆ 数据科学和机器学习案例演示



What is Data Science?

- Data science, also known as data-driven science, is an interdisciplinary field about scientific methods, processes and systems to extract <u>knowledge</u> or insights from <u>data</u> in various forms, either structured or unstructured,^{[1][2]} similar to <u>Knowledge Discovery in Databases</u> (KDD).
- Data science is a "concept to unify statistics, data analysis and their related methods" in order to "understand and analyze actual phenomena" with data.^[3] It employs techniques and theories drawn from many fields within the broad areas of <u>mathematics</u>, <u>statistics</u>, <u>information science</u>, and <u>computer science</u>, in particular from the subdomains of <u>machine learning</u>, <u>classification</u>, <u>cluster</u> <u>analysis</u>, <u>data mining</u>, <u>databases</u>, and <u>visualization</u>.

From Wikipedia



Data Scientist: The Sexiest Job of the 21st Century

MODERN DATA SCIENTIST

Tata Scientist, the sexiest jab of 20% contern requires a michan of multidisciplinary skills ranging from an intersection of mothematics, statistics, computer science, communication and business. Finding a data scientist is hard. Finding people who anderstand who a data scientist is, is equally hard. So here is a little cheat sheet as who the moders data scientist ranging in.

MATH & STATISTICS

- & Michine Isaming
- · Sanascal modeling
- ☆ Lournered design
- the Bayesian Interence
- Supervised learning decision lives random lavests, lagratic regression
- Unsupervised learning: clustering, fimereconality reduction
- Ophimization gradient descent and variants;

DOMAIN KNOWLEDGE & SOFT SKILLS

- Passenate about the business
- Corcon abmit state
- ☆ Influence without authority
- d Hadernmiller
- ☆ Probler soluti
- A Statege poactive analise analysisy and calibrative

PROGRAMMING & DATABASE

- Computer science fundamentals
- ☆ Scripting lingsuge s.g. Python
- ☆ Statistical comparing package e.g. It.
- 🖈 Datahases SQL and NeSQL
- 1 Relational algebra
- ☆ Parallel displaces and parallel every pricetsing
- MupRedure concepts
- ☆ Hadaoa Ant Have/Pig
- 🕁 Destan militera
- ✿ Expositece with was law JAVS

COMMUNICATION & VISUALIZATION

- Able to ensure with versor management.
- ☆ Stary talling shills
- Tomsule data driven multisinti decorrs aut actions
- 🖈 Visual art denym
- # R packages like gradel or failure
- Knowledge of any of visualizationtrols is g. Flam, D.S.s. Tobleau

What abilities make a data scientist successful?

Think of him or her as a hybrid of

- data hacker
- analyst
- communicator
- trusted adviser

The combination is extremely powerful—and rare.

-----Harvard Business Review Oct 2012 Issue

MarketingStatility const is a group at practitioners in the arms of a commerce marketing. Our fields of experise include marketing that any and approximation comment tracking and an air analytics pendicise analytics and economics: data womboring and thig data textment marketing downer lengths a Problem. Security 52(3) acade. (SMI and Strand.





数据科学家的硬技能 http://nirvacana.com/thoughts/becoming-a-data-scientist/ Using ETL Institut **Linearchine** How much Bala? Principal Google OpenAleTin Diels Serv Calif Ingention 6. Visualization Data Exploration in P (Hist, Exceptor etc) Percept Linest Regression Link (0) & Multivortate Viz Represeiter Shanking. sale of **Englatic Regrestite** Histogram & Pir Juni Tree & Tree Map 10 Toolbox Authon Swambaaaannaalia) an 4. Mathine Learning Scatter Piot (N) MS Excel of Address Engine Line Charts (6) Prob Dan Fe (PDF) 1. Fundamentals Java Pything panal Charts ANOVA R, ISSTudio, Walter Sarwy Plat Showned Mattices & Deep Algebra Forekamentals Woka Knine, Rapidlent **Hereiter Excisionants Displaying Store** al Poinson, Gaussian eri Trier Hash Furscheres, Binary Tree, G(H) Hindson Dist of Chester Onto Province Lines **Currul Dist Fe (DDF)** Relational Algebra, DG Basics Spirk, Spire Factors Alwarding CSV Bida Handom Valubles Flame, Scibe, Chakera Institu Galler, Cress, Thete Jon Arrays Barry Theorem Reading Raw Date CAP TRACKING Name, Talend, Schaperolike MANISTR Salassing Date Talvaur Dritt a Probability Theory Webscright, Flame, Spect DOGALO-MONT Data Pranes & Serves Venters Managed at a Auto Percentales & Dutates Name & Data Nodes Sharding C Storys Halling UR, HWURD, NUTH **HIMOGRAPHS** Sintup Harlosp (1814 / Clouthers / Hortonik/orks) Exections Variables DUMP Exploratory Data Analysis Data Myslegation Procepters itin o Expression (Multidenensional Data Model Descriptive Statistics HDES ears medium, ramps 525, Park R Battins, F ETL Hadoop Components ok a Detent Minister all Mag Medice Fundamentals ting Va ES in Analytica 7. Big Gath 5. Programmin







What is Machine Learning?

Computers that ...

Learn without being explicitly programmed Grow and change when exposed to new data Deliver personalized and optimized customer interactions

Identify Patterns not readily foreseen by humans



Build Models of behavior from those patterns



Achieving Business Value through Watson Machine Learning Capabilities



Machine Learning helps...

- Constantly learns and adapts
- Avoids making the same mistakes
- Faster, deeper, improved insights

Resulting in...

- ✓ Smarter business outcomes
- ✓ Lower business risks and costs
- ✓ New business opportunities





Churn analysis helps identify the cause of the churn and implement effective strategies for retention.



Detect and understand lifethreatening medical conditions and design ever more effective treatment programs



Learn, predict weather patterns and energy production from renewable sources and integrate into grid more



effectively Product recommendation, next purchase prediction, targeted offers – individual tailored shopping experience.



Identify suspicious behavior, predict and prevent threats / fraud – continually reduce business risks and costs



Machine Learning 101 : Types of machine learning

Classification

- Data points are labeled and are being used to predict a category
- Two-class vs multi-class
- Example:
 - Fraud detection (fraud vs non-fraud)
 - Spam email detection (spam vs non-spam)

Regression

- When a value is being predicted
- Example:
 - Stock prices prediction

Clustering

- Data points are not labeled.
- Goal is to group data into clusters to better organize the data



Machine Learning 101 : feature engineering

- A feature is a piece of information that might be useful for prediction – Example, predict the churn probability of a customer
- Labeled data is the desired output data
 - Example, CHURN_LABEL false representing a churn sample



What is Machine Learning (机器学习概要)

a TrainOps (DevOps) story



The (incomplete) machine learning process Takes significant development, deployment and management efforts



数据科学及机器学习新挑战

- 降低数据科学入门门槛 (Citizen Data Scientist)
- 管控机器学习全生命周期
- ■提高持续交付能力
- 数据科学的可重复性



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ruary 2017)

Data Science Experience (DSx) 主要特性



IBM Data Science Experience

社区

- 教材与数据集
- 连接数据科学家
- 提问
- 文章与论文
- 复制与分享项目

开源

- Scala/Python/R/SQL
- Jupyter and Zeppelin* Notebooks
- RStudio IDE and Shiny apps
- Apache Spark
- · Your favorite libraries

DSx Cloud Service <u>http://datascience.ibm.com</u>

DSx Local Edition

IBM 提供的能力

- 数据预处理/Pipeline UI*
- 自动数据准备与建模*
- 高级可视化*
- 模型管理与部署
- 模型API文档*
- Spark云服务/Packaged Spark

IBM Machine Learning for z/OS - 企业级机器学习平台



Feature Highlights – CADS 数据科学认知助手

- What is CADS?
 - Cognitive Assistant for Data Scientist which helps select the best fit algorithm for training
- Why Data Scientists need CADS?
 - Many algorithms for classification/regression tasks: SVM, Decision Trees/Forests, Naïve Bayes, Logistic Regression, etc.
 - -Substantial cost in user and compute time to select the best algorithm
 - · User spends time on trying various learners
 - Computational cost for training a single SVM can exceed 24h
 - Selection commonly based on data scientist bias and experience

Feature Highlights – CADS/HPO

- Minimize amount of data to be considered to make an informed selection of most suitable learner
- Given a data set try to select best approach by directly considering part of actual data



Feature Highlights

- Integrated Notebook Interface with flexible APIs



Feature Highlights - Data Visualization with Brunel (https://github.com/Brunel-Visualization/Brunel)

-504





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Feature Highlights – Visual Model Builder, the guided Machine Learning Interface

Select Data	Prepare da	ata set					Ingest transfo	data a orm	nd	Add a Transformer
Prepare	TWITTERID	CUST_ID	AVG_DAILY_TX	EDUCATION	EDUCATION_GROUP	-	AVO_TX_AMT	CHURNULA	Configured transfor	mers
Train	0	1009530860	0.9178079962730408	2	Bachelors degree	114368	2090.320068359375	false	StringIndexer	
Select Model	0	1009544000	0.9506850242614746	2	Bachelors degree	90298	2095.0400390625	falso	GenderCode	
Evaluate	0	1009534260	0.9205480217903655	2	Bachelors degree	94881	1723.4599609375	true .	ChurnLabel	
	0	1009574010	0.9945210218429565	2	Bachelors degree	112099	1297.419921875	trué	StringIndexer	
	0	1009578620	0.9178079962730408	5	Doctorate	84638	1333.179931640625	faise	StateCode	
	0	1009575250	0.9452049732208252	5	Doctorate	80194	1175.570068359375	true.	VectorAssembler AllFeatures	•
	A				-					

Select Data	Select	model			Training	andevaluatio	on	
Prepare	Binary 0	Zassification						
Train		MODEL NAME	ESTIMATOR TYPE	PERFORMACE	AREA UNDER ROC CURVE	AREA UNDER PR CURVE	LAST VALIDATION	ACTIONS
Select Model	•	BankingChumLRModel	logistic_regression	Excellent	0.96481	0.96867	19 Feb 2017, 7:48 PM	
Evaluate	0	BankingChurnLRWIth4FeaaturesModel	logistic_regression	Good	0.84418	0.79224	8 Mar 2017, 9:32 AM	
	0	BankingChurnLRWIth3FeaturesModel	logistic_regression	Poor	0.68375	0.64866	8 Mar 2017, 9:33 AM	

Feature Highlights – Model Management

Models Dashboard Deployments Manage model, Q Churn create deployment MODEL NAME DATE UPDATED ACTIONS MODEL ID OWNED BY DATE CREATED ChunCADS 12 2017-03-06 13:22 2017-03-06 13:22 ACTIVE steve View details ChurnModel 11 wmlz16 2017-03-06 06:53 2017-03-06 06:53 ACTIVE Create deployment AdarshChurnCADSModel 7 adarsh 2017-02-28 19:16 2017-02-28 19:16 ACTIVE Update Delete ChumCADSModelLR 4 steve 2017-02-19 11:55 2017-02-19 11:55 ACTIVE BankingChurnLRModel 3 steve 2017-02-19 11:50 2017-02-19 11:50 ACTIVE 1 2 ACTIVE ChurnCADSModel steve 2017-02-19 11:49 2017-02-19 11-49 **ChurnCADSWithNotebook** 2017-02-19 11:46 ACTIVE 1 steve 2017-02-19 11:46 Dashboard Models Deployments Manage Q deployment Notebook DEPLOYMENT NAME DEPLOYMENT ID MODEL EVALUATION SCHEDULED EATED ~ ACTIONS TentCADSNotebookModelD02 TentCADSNotebookModel NO 4:36 11 View details 2017-02-19 20:03 TentCADSNotebookModel 4 TentCADSNotebookModel NO Test **ChumCADSWithNotebook** 2 **ChumCADSWithNotebook** YES 2017-02-19 20:02 Schedule evaluation

Delete

Feature Highlights

- Easily consumable RESTful API for online Scoring within Application Code



Feature Highlights

- Feedback and Continuous Monitoring

valuation Highlights					Model Health
MODEL	EVALUATION TIME ~	METRICS	THRESHOLD	RESULT	
GCSalesTontLR4StoveD01	2017-01-23 02:01:23	aresUnderROC	0.7	A 0.64	92%
GOSalesTentLR4SteveD01	2017-01-22 02:03:35	areaUnderRCC	0.7	0.34	Health Index
TentFeedback	2017-01-20 07:33:02	areaUnderROC	0.95	4 0.64	\sim
TentFeedback	2017-01-20 06:57:03	areaUnderROC	0.05	A 0.64	24 MODELS IN TOTAL 2 models have been highlighted
Spark Status		Top mode	els by number	of invocations	Top models by average elapsed time
				Number of invocation	Average elapsed time(ms)
Cores in use: 4 50% Memory in use: 75%		140			140
		120	r		120
		100			100
Andertina 2 Duning 199	Completed	80			30
ODBORNIN OLUBRING ISA	L HAR STAL CHARTER CON L TOTT				60
portations. Straining 120	101101100110011000				79



数据科学和机器学习的方法学



机器学习平台演示

客户流失预测





案例: "认知银行" - 了解你的客户

•业务问题:

我愿意提供优惠来防止客户流失,问题是我不知道谁会流失,过去都是流 失之后我才知道的

●解决方案:

用机器学习找到现有已流失客户的特点,用来预测现有客户的流失可能性

•知道之后应该怎么做:

推荐系统之"认知银行",以后介绍完整场景

IBM Machine Learning Platform



