

5 листопада 2013, 19:00 Зустріч-дискусія у Часописі

Сучасні Хмарні технології та технології Великих Даних: Можливості, тенденції та виклики для бізнесу та науки



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Теми презентації та питання для дискусії

- 1. Характеристики і можливості хмарних технологій, тенденції розвитку та стандартизація.
- 2. Приклади використання та типи впровадження комп'ютерних хмар: корпоративні, публічні, комунальні; міграція корпоративної ІТ інфраструктури на хмарну платформу, необхідні передумови і рівень "зрілості", переваги віртуалізації сервісів і ресурсів.
- 3. Законодавча та регуляторна база в Європі, програми підтримки впровадження хмар в Європі.
- 4. Глобальні провайдери хмарних послуг і ресурсів: Amazon AWS, Microsoft Azure, GoogleCloud: можливості, послуги, засоби розробки.
- 5. Великі Дані: Об'єм, Швидкість, Номенклатура, Мінливість, Цінність, Достовірність (Volume, Velocity, Variety, Variability, Value, Veracity).
- 6. Великі Дані та бізнес- аналітика: приклади використання і нові можливості.
- 7. Проблеми Великих Даних: зберігання, передача, обробка, контроль доступу, захист даних і персональної інформації.
- 8. Нові спеціальності для Хмарних технологій та Великих Даних: підготовка фахівців, тренінг та освіта.



Сучасні Хмарні технології та технології Великих Даних: Можливості, тенденції та виклики для бізнесу та науки

Cloud and Big Data Technologies:
Opportunities and Challenges for business and science

Yuri Demchenko SNE Group, University of Amsterdam

> Дискусія в Часописі, 5 листопада 2013, Київ



- Cloud Computing definition and features
- Cloud Computing Reference Architecture and standardisation
- Amazon AWS laaS cloud and cloud powered design principles
- European Cloud Computing strategy and legislation
- Big Data and Data Intensive Science as a new technology wave
 - Big Data 5+1 Vs: Volume, Velocity, Variety, Value, Variability, Veracity
- Big Data in Science, Industry and Business
 - Where do the data come from? What are Big Data drivers?
- Defining Big Data Architecture Framework (BDAF)
 - Big Data Infrastructure (BDI) and Big Data Analytics tools
- Data Scientist: New profession and need for Education&Training
- Summary and Discussion

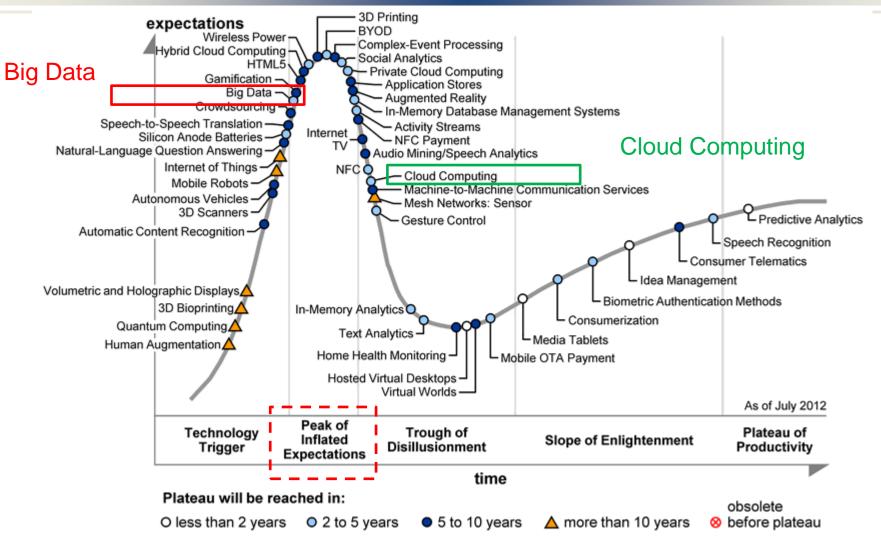


Big Data Research at System and Network Engineering (SNE), University of Amsterdam

- Long time research and development on Infrastructure services
 - High speed optical networking and data intensive applications
 - Application and infrastructure security services
 - Collaborative systems, Grid, Clouds and currently Big Data
- Focus on Infrastructure definition and services
 - Software Defined Infrastructure based on Cloud/Intercloud technologies
- Standardisation activity (IETF, OGF, ISO past)
 - NIST Cloud Computing Reference Architecture (past) and Big Data Working Group
 - Research Data Alliance: Education and Data Analytics Tools
- Big Data Interest Group at SNE, UvA
 - Non-formal but active, meets two-weekly for brainstorming sessions
 - Provides input to NIST BD-WG and RDA
- SNE Technical Report on Architecture Framework and Components for the Big Data Ecosystem. Draft Version 0.2, 12 September 2013
 http://www.uazone.org/demch/worksinprogress/sne-2013-02-techreport-bdaf-draft02.pdf



Gartner Technology Hypercycle



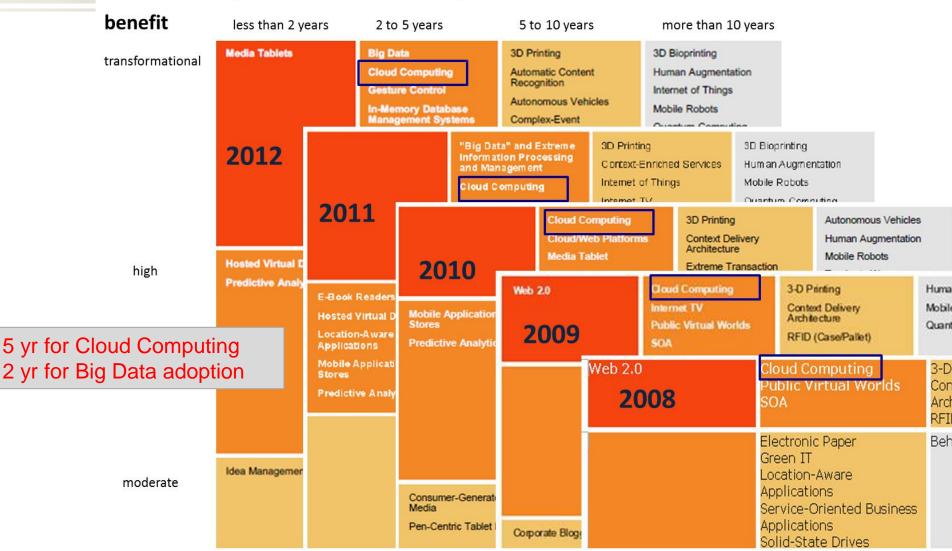
Source http://www.gartner.com/technology/research/methodologies/hype-cycle.jsp

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Gartner Priority Matrix

years to mainstream adoption





Technology Definitions and Timeline - Overview

- Service Oriented Architecture (SOA): First proposed in 1996 and revived with the Web Services advent in 2001-2002
 - Currently standard for industry, and widely used
 - Provided a conceptual basis for Web Services development
- Computer Grids: Initially proposed in 1998 and finally shaped in 2003 with the Open Grid Services Architecture (OGSA) by Open Grid Forum (OGF)
 - Currently remains as a collaborative environment
 - Migrates to cloud and inter-cloud platform
- Cloud Computing: Initially proposed in 2008 Now entering productive phase
 - Defined new features, capabilities, operational/usage models and actually provided a guidance for the new technology development
 - Originated from the Service Computing domain and service management focused
- Big Data and Data Intensive Science: Yet to be defined
 - Involves more components and processes to be included into the definition
 - Can be better defined as Ecosystem where data are the main driving component
 - Need to define the Big Data properties, expected technology capabilities and provide a guidance/vision for future technology development



Big Data and Clouds and Mobile Technologies – From disruptive to consolidating technologies

- Service Oriented Architecture (SOA) Industry
- Computer Grids, Distributed Computing Research community
- Cloud Computing: Initially proposed in 2008 Now entering productive phase (Industry)
 - Functional Cloud Computing definition provided a guidance for the technology development
 - Consolidating SOA, Distributed computing, SDN
 - Facilitated by Mobile technologies, Big Data
- Big Data and Data Intensive Science Originated from science
 - Consolidates Cloud Computing, Mobile technologies, High Performance computing,
 Data warehousing, Data analytics/science
 - Emerges new data centric models and technologies
 - Introduces new technical category Ecosystem where data are the main driving component
 - Need to define the Big Data properties, expected technology capabilities and provide a guidance/vision for future technology development



Основы технологии Облачных Вычислений

Cloud Computing technology foundation



Анатомия Облачных Технологий

- Обьединние большого количества вычислительных ресурсов и запоминающих устройств в больших Центрах Обработки Даных (ЦОД)
 - Экономия размера/масштаба
 - Горизонтальное и вертикальное масштабирование: балансировагние нагрузки и эластичность
- Виртуализация сервисов, ресурсов и платформ
 - Virtualisation == (Pooling) Abstraction Composition Deployment (Lifecycle management)
 - Механизмы привязки: пространство имен и безопасность/доверительность
 - Миграция виртуальных машин (BM; VM Virtual Machine)
- Многоуровневая модель виртуализации сервисов и ресурсов
 - Много-пользовательская среда (multi-tenancy), учет, билинг
- Всеобщий и универсальный доступ через Интернет, всеобщая коннективность



Top Cloud Providers (July 2013) - Ranked 1-20

- Salesforce.com (revenue >\$3Bln)
- 2. Amazon AWS (>\$1.5Bln)
- 3. Microsoft (>\$1.5Bln)
- 4. Oracle
- 5. Google
- 6. SAP
- 7. SoftLayer (IBM since 2013) + IBM's SmartCloud
- 8. Terremart (Verizon Company)
- 9. Rackspace
- 10. NetSuite ERP cloud service provider (\$308MIn)

- 11. Workday HR and financial cloud services
- 12. Dropbox
- 13. Savvis (CenturyLink Company)
 Oracle and private clouds
- 14. Joyent
- 15. Navisite (Time Warner) enterprise applications from IBM, MS, Oracle
- 16. Citrix Systems
- 17. LogMeIn Remote Management platform provider
- 18. Zoho alternative to Salesforce, Offcie365, Google Apps
- 19. Dimension Data (NTT Group, ZA) managed hosting
- 20 Carbinite backup provider for SQL, MS Exchange



European Cloud Computing Strategy (2012+)

In September 2012, the European Commission adopted a strategy for "Unleashing the Potential of Cloud Computing in Europe"

To deliver 2.5 mln new European jobs, and an annual boost of EUR 160 billion to EU GDP (around 1%)

Key actions

- Safe and Fair Contract Terms and Conditions
 - data preservation after termination of the contract
 - data disclosure and integrity
 - data location and transfer
 - ownership of the data
 - direct and indirect liability change of service by cloud providers and subcontracting
- Cutting through the Jungle of Standards
 - EU cloud framework to emerge in next 18+ months, by 2015
- Establishing a European Cloud Partnership (ECP)
- Mixed reactions from the major cloud players in Europe (e.g., Microsoft, HP, IBM, majority from US)



NIST Cloud definition – NIST SP 800-145 (1)

NIST SP 800-145 The NIST Definition of Cloud Computing (Draft) http://csrc.nist.gov/publications/drafts/800-145/Draft-SP-800-145_cloud-definition.pdf

- Five Cloud characteristics
 - On-demand self-service
 - Broad network access
 - Resource pooling
 - Rapid elasticity
 - Measured Service
- 3 basic service models
 - Software as a Service (SaaS)
 - Platform as a Service (PaaS)
 - Infrastructure as a Service (laaS)
- Deployment models
 - Private clouds
 - Public clouds
 - Hybrid clouds
 - Community clouds



NIST Cloud definition – NIST SP 800-145 (1)

NIST SP 800-145 The NIST Definition of Cloud Computing (Draft) http://csrc.nist.gov/publications/drafts/800-145/Draft-SP-800-145_cloud-definition.pdf

- Пять основных характеристик Облачных технологий
 - Самообслуживание по требованию
 - Широкополосный сетевой доступ
 - Агрегатирование (пулинг) ресурсов
 - Быстрая эластичность
 - Измеряемые услуги (учет объема услуг)
- 3 базовые модели услуг
 - Software as a Service (SaaS)
 - Platform as a Service (PaaS)
 - Infrastructure as a Service (laaS)
- Модели реализации
 - Private clouds
 - Public clouds
 - Hybrid clouds
 - Community clouds

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NIST Cloud definition – Draft SP 800-145 (2)

Cloud computing is a model for enabling ubiquitous, convenient, on-demand network access to a shared pool of configurable computing resources (e.g., *networks*, servers, storage, applications, and services) that can be rapidly provisioned and released with minimal management effort or service provider interaction. This cloud model promotes availability and is composed of five essential characteristics, three service models, and four deployment models.

Cloud Infrastructure as a Service (laaS)

The capability provided to the consumer is to *provision processing, storage, networks*, and other fundamental computing resources where the consumer is able to *deploy and run arbitrary software, which can include operating systems and applications*. The consumer does not manage or control the underlying cloud infrastructure but has control over operating systems, storage, deployed applications, and *possibly limited control of select networking components (e.g., host firewalls)*.

Note: NIST Definition of Cloud – missing network provisioning, just "limited control over network".



NIST Cloud Definition - PaaS, SaaS

Platform as a Service (PaaS)

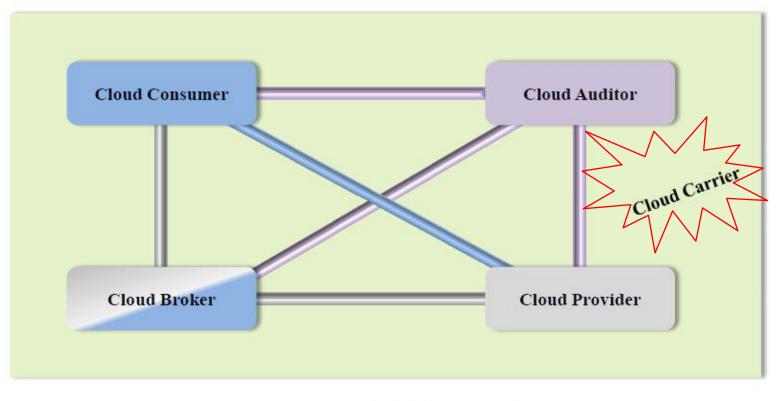
The capability provided to the consumer is to deploy onto the cloud infrastructure consumer-created or acquired applications created using *programming languages*, *libraries*, *services*, *and tools supported by the provider*. The consumer does not manage or control the underlying cloud infrastructure including network, servers, operating systems, or storage, but has *control over the deployed applications and possibly configuration settings for the application-hosting environment*.

Software as a Service (SaaS)

The capability provided to the consumer is to *use the provider's applications running* on a cloud infrastructure. The applications are accessible from various client devices through either a thin client interface, such as a web browser (e.g., web-based email), or a program interface. The consumer does not manage or control the underlying cloud infrastructure including network, servers, operating systems, storage, or even individual application capabilities, with the possible exception of *limited user-specific* application configuration settings.



NIST Cloud Computing Reference Architecture (CCRA) 2.0 - Main Roles (1)



The communication path between a cloud provider and a cloud consumer

The communication paths for a cloud auditor to collect auditing information

The communication paths for a cloud broker to provide service to a cloud consumer

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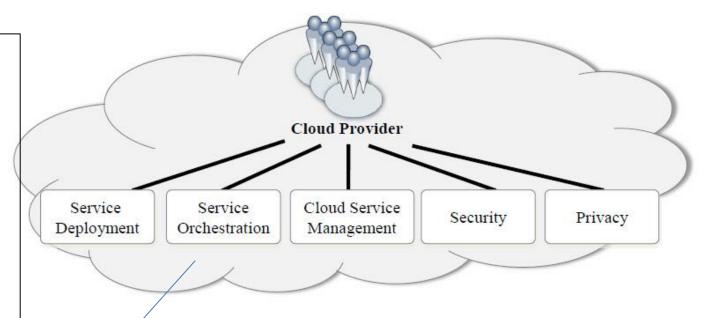


NIST Cloud Computing Reference Architecture (CCRA) 2.0 - Provider Functions (2)

Service provisioning stages:

Service Delivery (Framework)

- Request&SLA
- Reservation/ Composition
- Deployment
- Operation
- Decommissioning



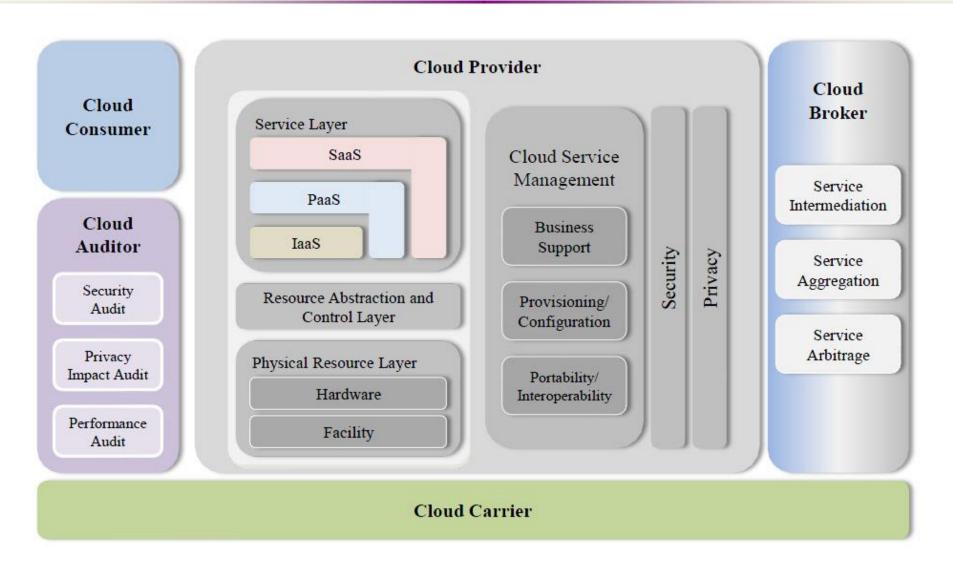


Функции провайдера облачных услуг

- Внедрение.установка сервисов и ресурсов
- Взаимодействие сервисов и ресурсов
- Управление ресурсами и сервисами
- Безопасность
- Приватность

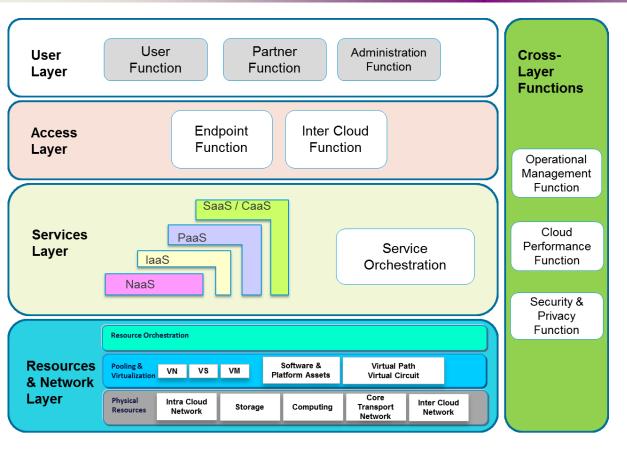


NIST Cloud Computing Reference Architecture (CCRA) 2.0 – Consolidated View (3)



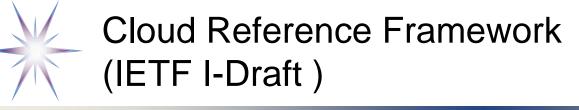


ITU-T FG on Clouds Part 2: Functional requirements and reference architecture



Layered Cloud computing architecture includes:

- User layer (including user functions, partner functions, administration functions)
- Access layer (including endpoint functions and inter-cloud functions). Network service providers role is to provide intercloud transport network
- Cloud services layer (including basic cloud services laaS, PaaS, SaaS, NaaS, CaaS and also Orchestration service)
- Resources and network layer (including physical resources, pooling and orchestration, pooling and virtualisation)



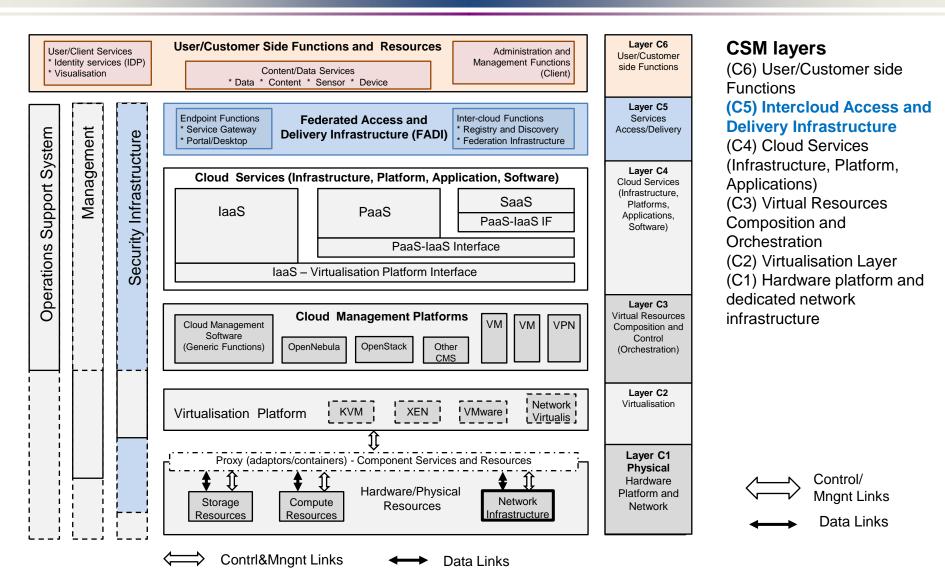
Cloud Reference Framework. Internet Draft, by B. Khasnabish, J. Chu, S. Ma, Y. Meng, N. So, P. Unbehagen, M. Morrow, M. Hasan, Y. Demchenko http://tools.ietf.org/html/draft-khasnabish-cloud-reference-framework-05.txt

- Multilayer Cloud Services Model (CSM)
 - Including Federated Access and Delivery Infrastructure layer and user side services layer
- Intercloud Architecture Framework (ICAF)
 - InterCloud Control and Management Plane (ICCMP)
 - Signaling, monitoring, synchronisation between heterogeneous clouds
 - InterCloud Federation Framework (ICFF)
 - · Protocols and mechanisms for heterogeneous clouds integration
 - InterCloud Operations and Management Framework (ICOMF)
 - Services and business processes management and operation
 - Intercloud Security Framework (ICSF)



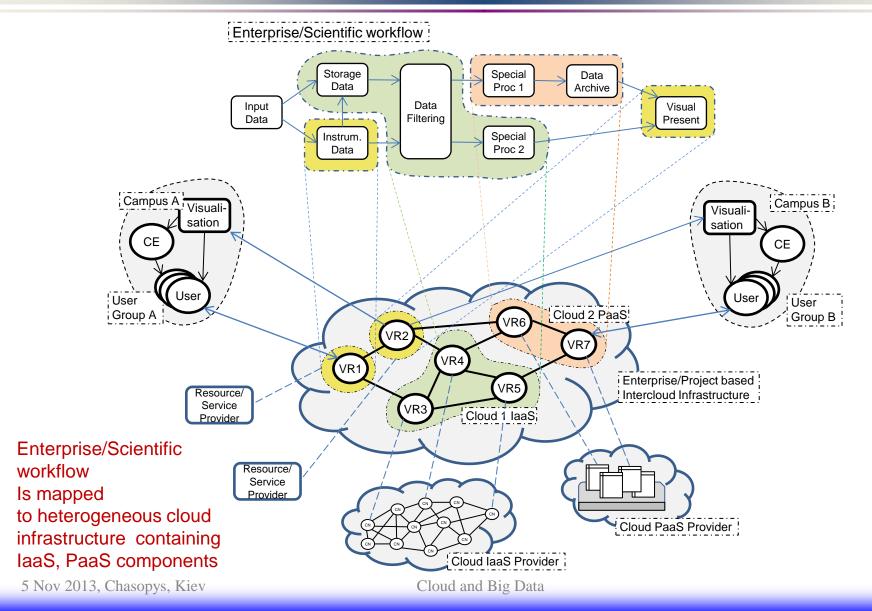
Multilayer Cloud Services Model (CSM)

http://www.ietf.org/id/draft-khasnabish-cloud-reference-framework-05.txt



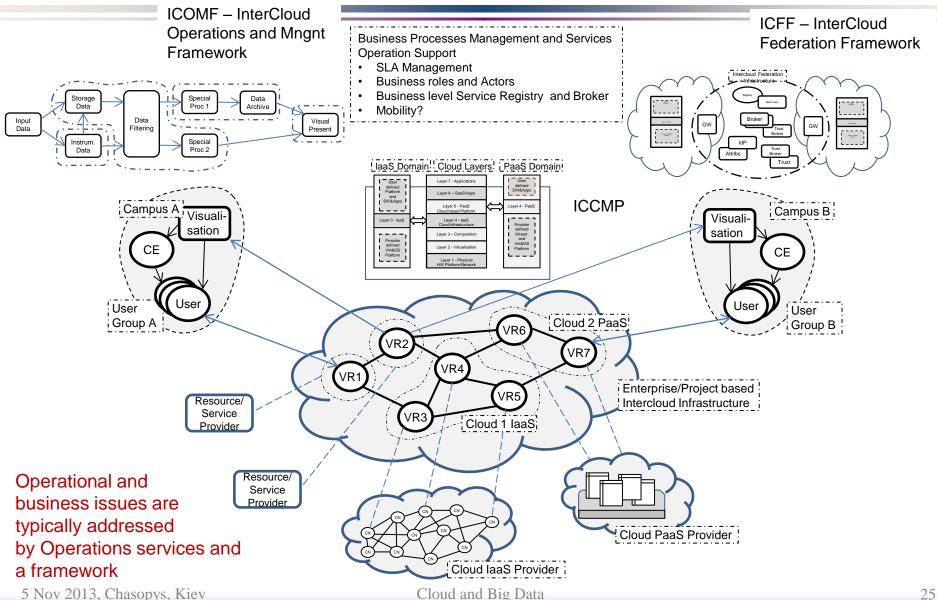


General use case for infrastructure provisioning: Workflow => Logical (Cloud) Infrastructure





Intercloud Applications Integration: ICCM, ICFF, ICOMF



5 Nov 2013, Chasopys, Kiev Cloud and Big Data



Cloud laaS – облачные инфраструктурные сервисы



Cloud laaS – Основные свойства и характеристики облачных систем и инфраструктур

• Все фундаментальные свойства и характеристики компьютерных облаков должны быть адресованы и внедреныв laaS

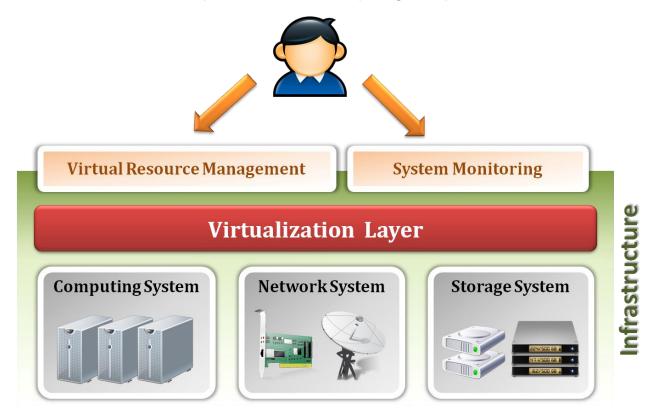


- Масштабируемость и эластичность
 - Динамичное внедрение и установка
 - Много-пользовательская среда
- Доступность и надежность
 - Устойчивость к сбоям
 - Устойчивость системы
 - Безопасность
- Управляемость и совместимость
 - Управляющая информация
 - Мониторинг систем
 - Биллинг
- Производительность и оптимизация
 - Параллельная обработка
 - Баланс нагрузки
 - Планирование работ
- Доступ и переносимость
 - Стандартный доступ (через Интернет)
 - Тонкий клиент
 - Стандартизация



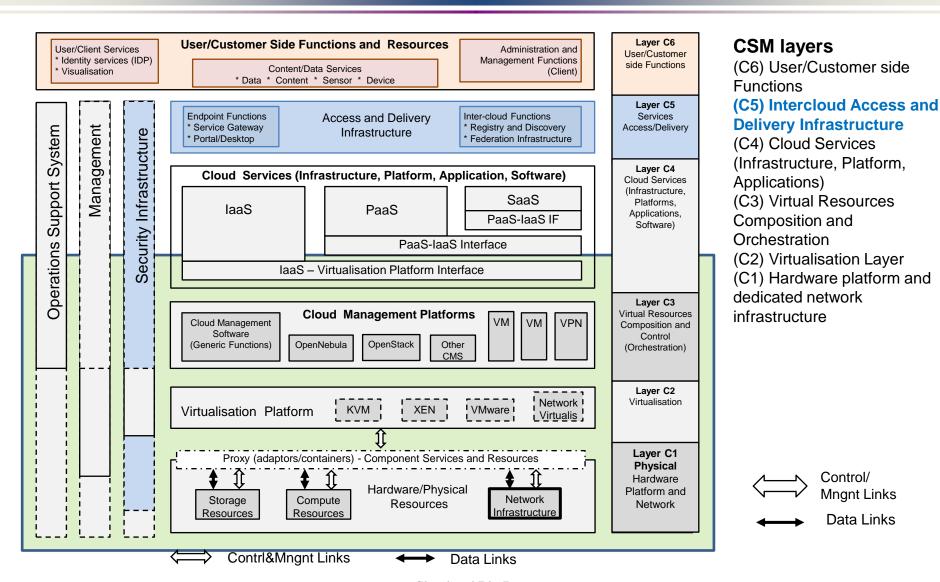
Cloud IaaS Architecture

- Infrastructure as a Service (laaS) delivers computer infrastructure for cloud user, typically a platform virtualization environment as a service.
- Virtualization is an enabling technique to provide an abstraction of logical resources away from underlying physical resources.



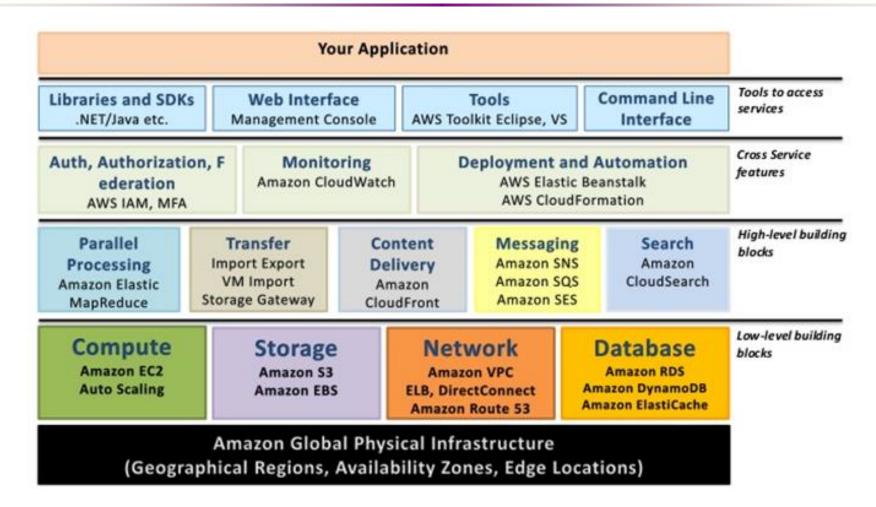


Multilayer Cloud Services Model (CSM)





Amazon AWS Cloud Architecture



Credits "Building Powerful Web Applications in the AWS Cloud" by Louis Columbus http://softwarestrategiesblog.com/2011/03/10/building-powerful-web-applications-in-the-aws-cloud/

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Amazon EC2 User Application Component Services

- EBS Elastic Block Store
- Elastic IP Address dynamically assigned to user VMs
- VPC (Virtual Private Cloud) allows organizations to use AWS resources along with their existing infrastructure in a VPN (Virtual Private Network) to increase compute capabilities beyond the local resources.
- CloudWatch monitoring service
- Auto Scaling dynamic resource provision
- Elastic Load Balancing between multiple VMs located within a single availability zone or multiple zones
- VM Import/Export for custom VM images store and load
- Cloud Formation services composition and deployment



Amazon S3 (Simple Storage Service)

S3 is a service that stores large amounts of data, and is accessible via the Internet

- Amazon S3 is intentionally built with a minimal feature set
- Data stored as objects associated with unique keys
- Objects can store up to 5 TB of data and are bound to specific buckets which can only be stored in a particular region (availability zone)

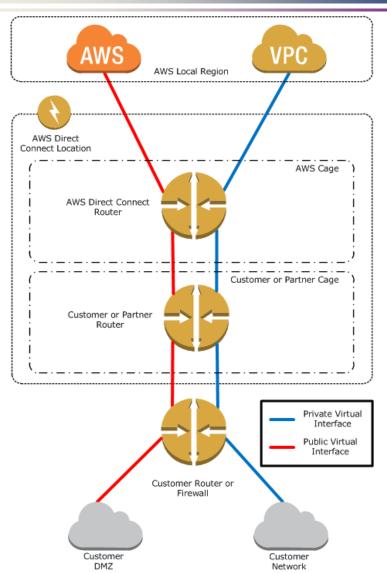
"Data stored in Amazon S3 is secure by default" :-); only bucket and object owners have access to the Amazon S3 resources they create

- Amazon S3 supports multiple access control mechanisms, as well as encryption for both secure transit and secure storage on disk
- With Amazon S3's data protection features, you can protect your data from both logical and physical failures, guarding against data loss from unintended user actions, application errors, and infrastructure failures
- For customers who must *comply with regulatory standards such as PCI and HIPAA*, Amazon S3's data protection features can be used as part of an overall strategy to achieve compliance
- Cost issues: free upload, paid download

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New Service: Amazon Direct Connect



- AWS Direct Connect links
 customer internal network to an
 AWS Direct Connect location over
 a standard 1 Gbps or 10 Gbps
 Ethernet fiber-optic cable
 - One end of the cable is connected to customer router, the other to an AWS Direct Connect router
 - Allows creating virtual interfaces directly to the AWS cloud (Amazon EC2, S3) and to Amazon Virtual Private Cloud (Amazon VPC), bypassing Internet service providers in your network path
- Access is limited to Amazon Web Services in the region



Example: Cloud powered services design with AWS

 Based on AWS seminar presentation "Building Powerful Web Applications in the AWS Cloud" by Louis Columbus

http://softwarestrategiesblog.com/2011/03/10/building-powerful-web-applications-in-the-aws-cloud/

Pattern #1: Design for failure and nothing will fail

Pattern #2: Edge cache static content

Pattern #3: Implement Elasticity

Pattern #4: Leverage Multiple Availability Zones

Pattern #5: Isolate read and write traffic; Isolate static and dynamic traffic

Pattern #6: Hardening security at every stage

Pattern #7: Parallel Processing

Pattern #8: Go global quickly (with single API)

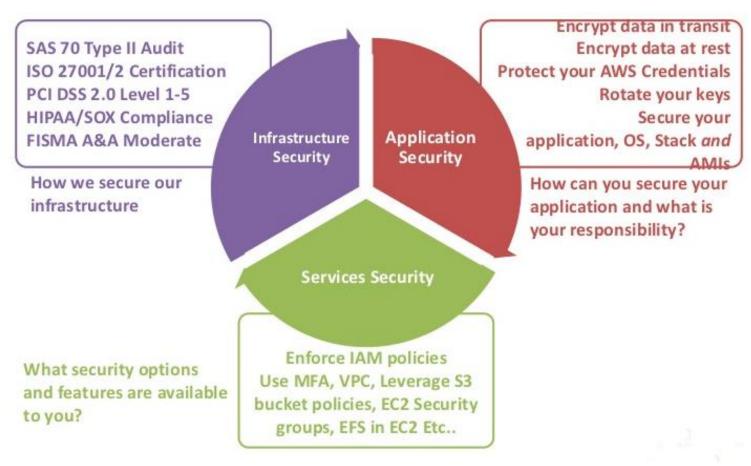
Pattern #9: Automate your in-cloud Development and Deployment Lifecycle

Pattern #10: Keep optimizing and see the savings in the next month's bill



Hardening Security

In the cloud, Security is a Shared Responsibility

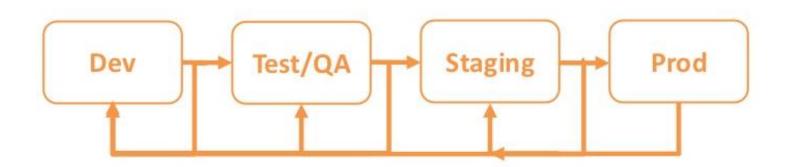


Credits "Building Powerful Web Applications in the AWS Cloud" by Louis Columbus http://softwarestrategiesblog.com/2011/03/10/building-powerful-web-applications-in-the-aws-cloud/

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Cloud powered development lifecycle



For the Test/QA stage:

Speed: quickly get on-demand resources

Variety: test more demo configurations

Real world load and stress testing; easy simulates

100s of clients

Repeatability: preconfigured shareable Test DB in

minutes

Reproducibility: "Save As" Productions Environment

and re-launch in Test Environment

Savings: "Turn off" Testing Environment

- Environmental separation
- Environmental consistency

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- Variable resource
- Different control levels

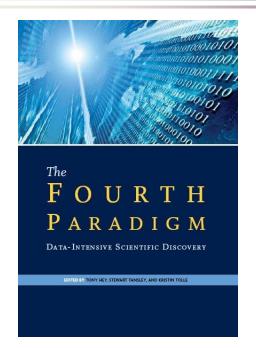
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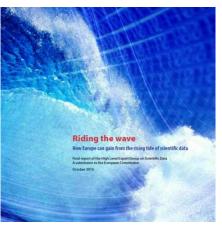
Технологии Больших Даных



Visionaries and Drivers: Seminal works and High level reports



The Fourth Paradigm: Data-Intensive Scientific Discovery. By Jim Gray, Microsoft, 2009. Edited by Tony Hey, et al. http://research.microsoft.com/en-us/collaboration/fourthparadigm/



Riding the wave: How Europe can gain from the rising tide of scientific data.

Final report of the High Level Expert Group on Scientific Data. October 2010.

http://cordis.europa.eu/fp7/ict/e-infrastructure/docs/hlg-sdi-report.pdf



Research Data Sharing without barriers

https://www.rd-alliance.org/

NIST Big Data Working Group (NBD-WG) https://www.rd-alliance.org/



AAA Study: Study on AAA Platforms For Scientific data/information Resources in Europe, TERENA, UvA, LIBER, UinvDeb.



The Fourth Paradigm of Scientific Research

- Theory and logical reasoning
- 2. Observation or Experiment
 - E.g. Newton observed apples falling to design his theory of mechanics
 - But Gallileo Galilei made experiments with falling objects from the Pisa leaning tower
- 3. Simulation of theory or model
 - Digital simulation can prove theory or model
- 4. Data-driven Scientific Discovery (aka Data Science)
 - More data beat hypnotized theory



Big Data and Data Intensive Science - The next technology focus

Scientific and Research Data – e-Science

- Big Data is/has becoming the next buzz word
 - Not much academic research and papers Read seminal works, Dive into blogs and tweets
- Based on e-Science concept and entire information and artifacts digitising
 - Requires also new information and semantic models for information structuring and presentation
 - Requires new research methods using large data sets and data mining
 - · Methods to evolve and results to be improved
- Changes the way how the modern research is done (in e-Science)
 - Secondary research, data re-focusing, linking data and publications
- Big Data require a new infrastructure to support both distributed data (collection, storage, processing) and metadata/discovery services
 - High performance network and computing, distributed storage and access
 - Cloud Computing as native platform for distributed dynamic virtualised (data supporting) infrastructure
 - Demand for trusted/trustworthy infrastructure



Big Data Definitions Overview

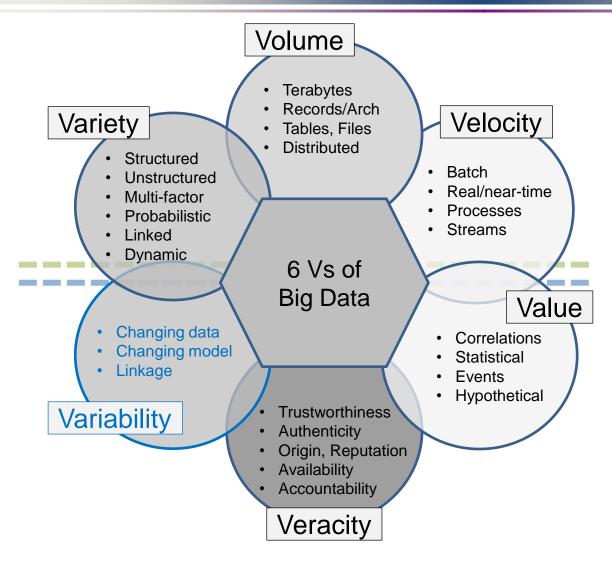
- IDC definition of Big Data (conservative and strict approach):
 "A new generation of technologies and architectures designed to economically extract value from very large volumes of a wide variety of data by enabling high-velocity capture, discovery, and/or analysis"
- Gartner definition
 Big data is high-volume, high-velocity and high-variety information assets that demand costeffective, innovative forms of information processing for enhanced insight and decision
 making. http://www.gartner.com/it-glossary/big-data/
 - Termed as 3 parts definition, not 3V definition
- Big Data: a massive volume of both structured and unstructured data that is so large that
 it's difficult to process using traditional database and software techniques.
 - From "The Big Data Long Tail" blog post by Jason Bloomberg (Jan 17, 2013). http://www.devx.com/blog/the-big-data-long-tail.html
- "Data that exceeds the processing capacity of conventional database systems. The data is too big, moves too fast, or doesn't fit the structures of your database architectures. To gain value from this data, you must choose an alternative way to process it."
 - Ed Dumbill, program chair for the O'Reilly Strata Conference
- Termed as the Fourth Paradigm *)

 "The techniques and technologies for such data-intensive science are so different that it is worth distinguishing data-intensive science from computational science as a new, fourth paradigm for scientific exploration." (Jim Gray, computer scientist)

^{*)} The Fourth Paradigm: Data-Intensive Scientific Discovery. Edited by Tony Hey, Stewart Tansley, and Kristin Tolle. Microsoft, 2009.



Improved: 5+1 V's of Big Data



Generic Big Data Properties

- Volume
- Variety
- Velocity

Acquired Properties (after entering system)

- Value
- Veracity
- Variability

Commonly accepted 3V's of Big Data





Improved: 5+1 V's of Big Data



Базовые свойства

- Обьем
- Скорость
- Номенклатура (разнообразие)

Приданные свойства (после обработки)

- Ценность
- Достоверность
- **Изменчивость**Сомморы ассер

Commonly accepted 3V's of Big Data





Big Data Definition: From 5+1V to 5 Parts (1)

(1) Big Data Properties: 5V

- Volume, Variety, Velocity, Value, Veracity
- Additionally: Data Dynamicity (Variability)

(2) New Data Models

- Data Lifecycle and Variability
- Data linking, provenance and referral integrity

(3) New Analytics

Real-time/streaming analytics, interactive and machine learning analytics

(4) New Infrastructure and Tools

- High performance Computing, Storage, Network
- Heterogeneous multi-provider services integration
- New Data Centric (multi-stakeholder) service models
- New Data Centric security models for trusted infrastructure and data processing and storage

(5) Source and Target

- High velocity/speed data capture from variety of sensors and data sources
- Data delivery to different visualisation and actionable systems and consumers
- Full digitised input and output, (ubiquitous) sensor networks, full digital control



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- Data Lifecycle and Variability/Evolution

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Big Data Definition: From 5V to 5 Parts (2)

Refining Gartner definition

"Big data is (1) high-volume, high-velocity and high-variety information assets that demand (3) cost-effective, innovative forms of information processing for (5) enhanced insight and decision making"

- Big Data (Data Intensive) Technologies are targeting to process (1) high-volume, high-velocity, high-variety data (sets/assets) to extract intended data value and ensure high-veracity of original data and obtained information that demand (3) cost-effective, innovative forms of data and information processing (analytics) for enhanced insight, decision making, and processes control; all of those demand (should be supported by) (2) new data models (supporting all data states and stages during the whole data lifecycle) and (4) new infrastructure services and tools that allows also obtaining (and processing data) from (5) a variety of sources (including sensor networks) and delivering data in a variety of forms to different data and information consumers and devices.
 - (1) Big Data Properties: 5V
 - (2) New Data Models
 - (3) New Analytics
 - (4) New Infrastructure and Tools
 - (5) Source and Target

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Big Data Nature: Origin and consumers (target)

Big Data Origin

- Science
- Internet, Web
- Industry
- Business
- Living Environment, Cities
- Social media and networks
- Healthcare
- Telecom/Infrastructure



Big Data Target Use

- Scientific discovery
- New technologies
- Manufacturing, processes, transport
- Personal services, campaigns
- Living environment support
- Healthcare support
- Social Networking



Volume, Velocity, Variety – Examples e-Science

- Volume Terabyte records, transactions, tables, files.
 - LHC 5 PB a month (now is under re-construction)
 - LOFAR, SKA 5 PB every hour, requires processing asap to discard non-informative data
 - Large Synoptic Survey Telescope (LSST) 10 Petabytes per year
 - Genomic research x10 TB per individual
 - Earth, climate and weather data
- Velocity batch, near-time, real-time, streams.
 - LHC ATLAS detector generates about 1 Petabyte raw data per second, during the collision time about 1 ms
- Variety structures, unstructured, semi-structured, and all the above in a mix
 - Biodiversity, Biological and medical, facial research
 - Human, psychology and behavior research
 - History, archeology and artifacts



Volume, Velocity, Variety – Examples Industry

- Volume Terabyte records, transactions, tables, files.
 - A Boeing Jet engine produce out 10TB of operational data for every 30 minutes they run
 - Hence a 4-engine Jumbo jet can create 640TB on one Atlantic crossing. Multiply that to 25,000 flights flown each day
- Velocity batch, near-time, real-time, streams.
 - Today's on-line ads serving requires 40ms to respond with a decision
 - Financial services (i.e., stock quotes feed) need near 1ms to calculate customer scoring probabilities
 - Stream data, such as movies, need to travel at high speed for proper rendering
- Variety structures, unstructured, semi-structured, and all the above in a mix
 - WalMart processes 1M customer transactions per hour and feeds information to a database estimated at 2.5PB (petabytes)
 - There are old and new data sources like RFID, sensors, mobile payments, in-vehicle tracking, etc.



Big Data technology drivers (1)

- Modern e-Science in search for new knowledge
 - Scientific experiments and tools are becoming bigger and heavily based on data processing and mining
- Traditional data intensive industry
 - Genomic research, drugs development, Healthcare
 - High-tech industry, CAD/CAM, weather/climate, etc.
- Intelligence and security
- Network/infrastructure management
 - Network monitoring, Intrusion detection, troubleshooting

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Big Data technology drivers (2)

- Consumer facing companies like Google and Facebook have driven many of the recent advances in Big Data efficiency
 - Facebook has some 900+ million users and is still growing
 - Google handles number of search queries at 3 billion per day
 - Twitter handles some 400 million tweets per day count for 12 terabytes per day
 - Used also for market sentiments prediction
 - Power companies: process up to 350 billion annual meter readings to better predict power consumption
- Processes/activity data recording and analysis
 - Flight data, log data, intelligence, traffic
- Business (retail) uses Big Data technologies "to search" for customers
 - Modern business concept (multi-channel) of delivering directly to customers requires prediction of customer behavior
 - Data volumes What cause(s) and what effect?
 - Big Data gives companies a fighting chance in the battle over the customer



Big Data technology drivers (3) - Advertising

- "... this new course of big data, gleaned from a wealth of unstructured information on the web, has the ability to turn advertising on its head— at least enough to make media people rethink algorithms for maximizing performance." HessieJ.com
 - Traditional Ad Model: User profiles
 - More Sophisticated Ad Model: Behavioral targeting "smart ads"
 - Future Ad Model: Enter Social Data

Example:

- Mary Brown searches for information about a future trip to Mallorca
 - She also goes to travel sites, reads hotel reviews and has excitedly spoken to close friends on Twitter and Facebook about her plans and preparations
- Now we have not only recent behavioral activity where she's been on the internet, but we also are aware of her conversations that validate her behavior
- It is safe to assume that Mary will "definitely" be going to Mallorca
- What this information does for a travel company?
 - They now have MORE information on that user that will allow them to not only serve an ad, or even respond to that user with relevant offers, but DO so with a certain degree of confidence that Mary will at the very least click on the ad.



Big Data technology drivers (3a) – Service Delivery

- Consumer products and services delivery
 - Netflix already captures movie genre preferences by the user and makes recommendations based on recent shows/movies watched
 - Announced \$2mln prize for effective customer targeting in 2003
 - Netflix recommender system in use as a reference technology implementation
 - It is already capturing which devices the user is watching recent programs/shows and when
 - Marrying that data with GetGlue (news feed on movies), for example, validates the original information and supplements the usage information

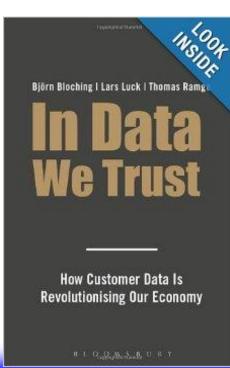
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- Combined and correlating, allows Netflix to optimize the movie offering to keep you a satisfied customer
- It can also capture the comments and shares from those watching the movie in order to drive messaging to attract new users



Big Data technology drivers (3b) – Managing public campaigns, e.g. election, public relations

- The rise of public opinion stored in platforms like Twitter,
 Google, Facebook, etc. provide enough intelligence to
 influence the campaign development, timing, geography and
 even the colour of the campaign signs
 - Twitter was a major source of data aggregation for the Republican Race in the US
 - Multimillion-dollar contract for data management and collection services awarded May 1, 2013 to Liberty Work to build advanced list of voters
 - Article "In Data we trust" by T.Edsall in The New York Times
 - Book: In Data We Trust: How Customer Data is Revolutionising Our Economy (Aug 2012)
 - A strategy for tomorrow's data world



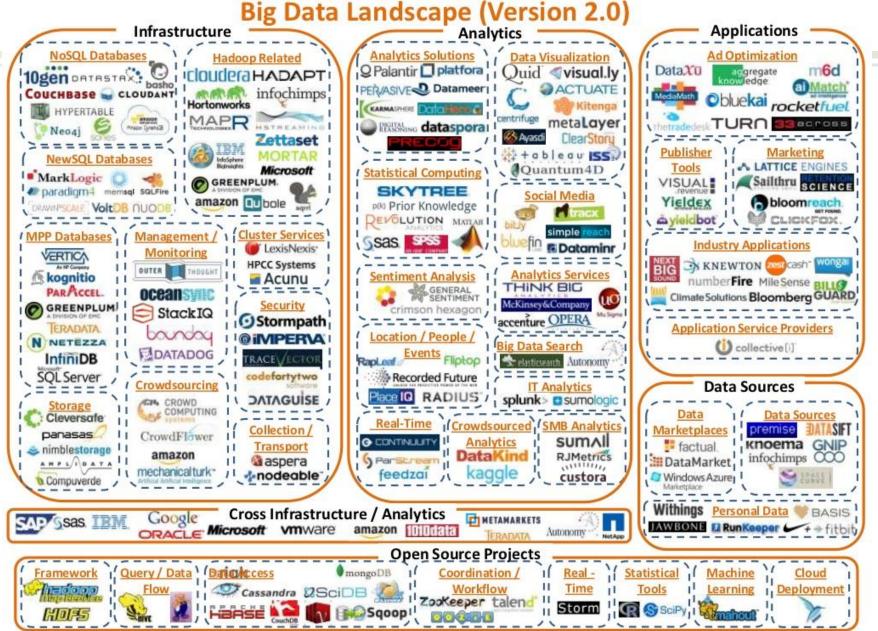


Big Data technology drivers (4) - Emerging

- Social media itself share and socialise/collaborate
 - Facebook, Twitter, YouTube, Flickr, etc.
- Workplace improvement
 - Means more data will be collected and monitored on the personnel
- Healthcare, health/physiological and medical information
 - Human health monitoring not just for ill or aged people
 - Early diagnostics, proactive care advising



http://mattturck.com/2012/10/15/a-chart-of-the-big-data-ecosystem-take-2/





Foreseen Big Data Innovations in 2013+

- Cloud-Based Big Data Solutions
 - Amazon's Elastic Map Reduce (EMR) is a market leader
 - Expected new innovative Big Data and Cloud solutions
- Real-Time Hadoop
 - Google's Dremel-like solutions that will allow real-time queries on Big Data and be open source
- Distributed and deep Machine Learning
 - Mahout iterative scalable distributed back propagation machine learning and data mining algorithm
 - New algorithms Jubatus, HogWild
- Big Data Appliances (also for home)
 - Raspberry Pi and home-made GPU clusters
 - Hardware vendors (Dell, HP, etc.) pack mobile ARM processors into server boxes
 - Adepteva's Parallella will put a 16-core supercomputer into range of \$99
- Easier Big Data Tools
 - Open Source and easy to use drag-and-drop tools for Big Data Analytics to facilitate the BD adoption
 - Commercial examples: Radoop = RapidMiner + Mahout, Tableau, Datameer, etc.
 - LexisNexis: from ECL (Enterprise Control Language) to KEL (Knowledge Engineering Language)

Source: Big Data in 2013 by Mike Guattieri, Forrester



NIST Big Data Working Group (NBD-WG)

- Deliverables target September 2013
 - 26 September initial draft documents
 - 30 September Workshop and F2F meeting
- Activities: Conference calls every day 17-19:00 (CET) by subgroup - http://bigdatawg.nist.gov/home.php
 - Big Data Definition and Taxonomies
 - Requirements (chair: Geoffrey Fox, Indiana Univ)
 - Big Data Security
 - Reference Architecture
 - Technology Roadmap
- BigdataWG mailing list and useful documents
 - Input documents http://bigdatawg.nist.gov/show_InputDoc2.php
 - Big Data Reference Architecture <u>http://bigdatawg.nist.gov/_uploadfiles/M0226_v7_2611176301.docx</u>
 - Big Data Architectures Survey http://bigdatawg.nist.gov/_uploadfiles/M0151_v2_7447424902.docx
 - Requirements based on 51 usecases
 http://bigdatawg.nist.gov/_uploadfiles/M0224_v1_1076079077.xlsx

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Defining Big Data Architecture Framework

- Existing attempts don't converge to something consistent: ODCA, TMF, NIST
 - See http://bigdatawg.nist.gov/_uploadfiles/M0151_v2_7447424902.docx
- Architecture vs Ecosystem
 - Big Data undergo a number of transformations during their lifecycle
 - Big Data fuel the whole transformation chain
 - Data sources and data consumers, target data usage
 - Multi-dimensional relations between
 - Data models and data driven processes
 - Infrastructure components and data centric services
- Architecture vs Architecture Framework
 - Separates concerns and factors
 - Control and Management functions, orthogonal factors
 - Architecture Framework components are inter-related

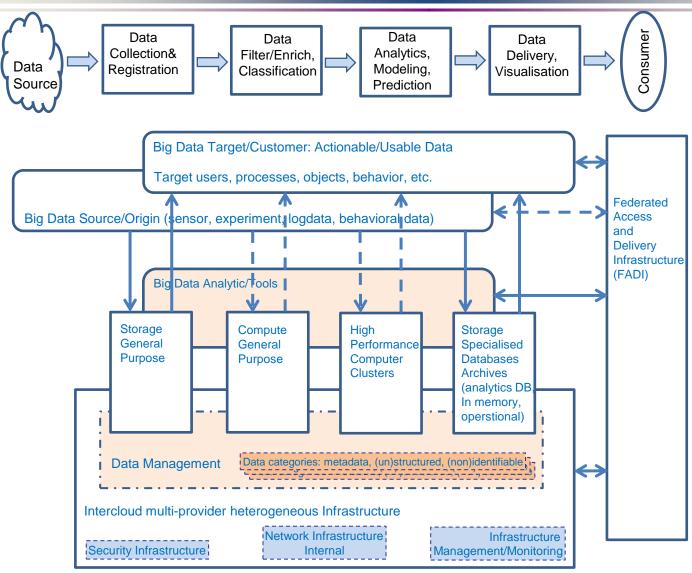


Big Data Architecture Framework (BDAF) (1)

- (1) Data Models, Structures, Types
 - Data formats, non/relational, file systems, etc.
- (2) Big Data Management
 - Big Data Lifecycle (Management) Model
 - Big Data transformation/staging
 - Provenance, Curation, Archiving
- (3) Big Data Analytics and Tools
 - Big Data Applications
 - Target use, presentation, visualisation
- (4) Big Data Infrastructure (BDI)
 - Storage, Compute, (High Performance Computing,) Network
 - Sensor network, target/actionable devices
 - Big Data Operational support
- (5) Big Data Security
 - Data security in-rest, in-move, trusted processing environments



Big Data Ecosystem: Data, Transformation, Infrastructure



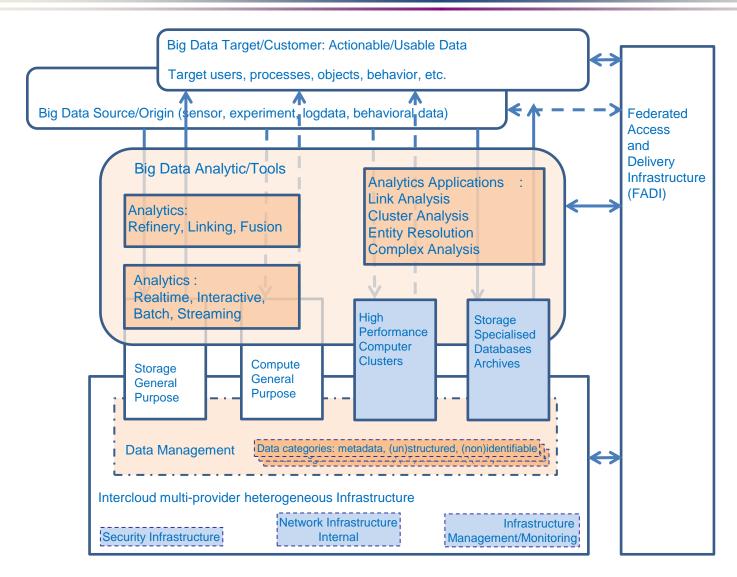


General BDI services and components

- Data management infrastructure and tools
- Registries, search/indexing, ontologies, schemas, namespace
- Collaborative Environment (user/groups managements)
- Heterogeneous multi-provider Inter-cloud infrastructure
 - Compute, Storage, Network (provisioned on-demand dynamically scaling)
 - Federated Access and Delivery Infrastructure (FADI)
- Advanced high performance (programmable) network
- Security infrastructure (access control, Identity and policy management, confidentiality, privacy, trust)



Big Data Infrastructure and Analytic Tools





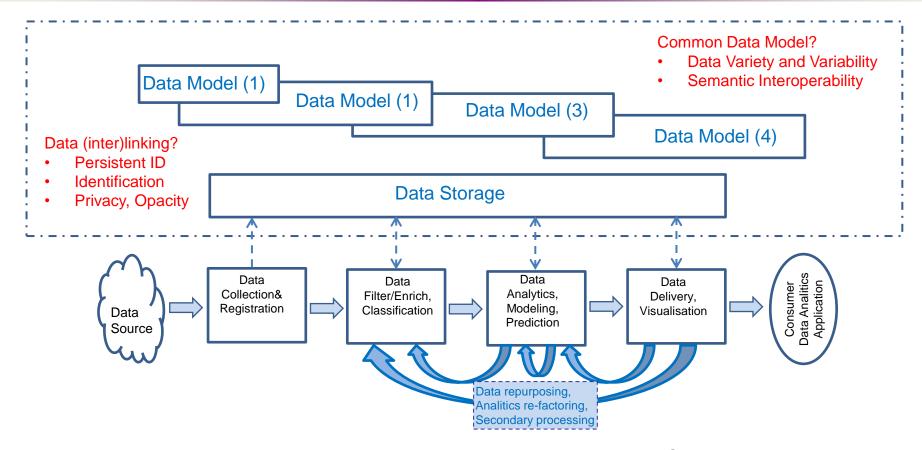
Big Data Analytics Infrastructure

- High Performance Computer Clusters (HPCC)
- Specialised Storage, Distributed/Replicated, Archives, Databases, SQL/NoSQL
- Big Data Analytics Tools/Applications
- Analytics/processing: Real-time, Interactive, Batch, Streaming
- Link Analysis, Graph analysis
- Cluster Analysis
- Entity Resolution
- Complex Analysis

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Data Transformation/Lifecycle Model

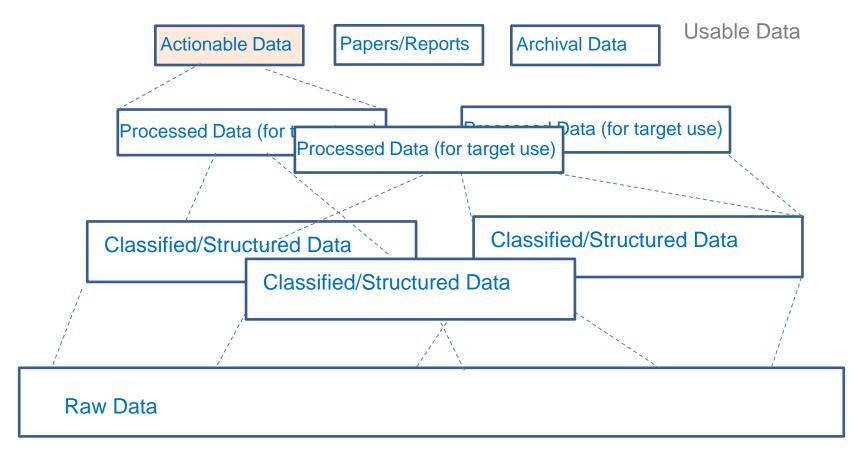


- Does Data Model changes along lifecycle or data evolution?
- Identifying and linking data
 - Persistent identifier

- Traceability vs Opacity
- Referral integrity



Evolutional/Hierarchical Data Model



- Common Data Model?
- Data interlinking?
- Fits to Graph data type?
- Metadata

- Referrals
- Control information
- Policy
- Data patterns



Data Science: Research and Education



Horizon2020: Challenge (2.1): Development, deployment and operation of e-infrastructures

Call on Specific Challenge 2.1: Development, deployment and operation of einfrastructures (based on discussion draft summer 2013)

CHALLENGE 1 – High Performance Computing (HPC)

CHALLENGE 2 - CONNECTIVITY

Topic 4: Research and Education Networking – GÉANT

CHALLENGE 3 - DATA

Topic 5: Community data services

Topic 6: Managing, preserving and computing with big research data

Topic 7: e-Infrastructure for Open Access

Topic 8: Towards global data e-infrastructures - RDA

CHALLENGE 4 – e-INFRASTRUCTURE INTEGRATION

Topic 9: e-Infrastructures for virtual research environments (VRE)

Topic 10: Provisioning of core services across e-Infrastructures

Topic 11: Skills and new professions for e-infrastructures

CHALLENGE 5 – POLICY AND INTERNATIONAL

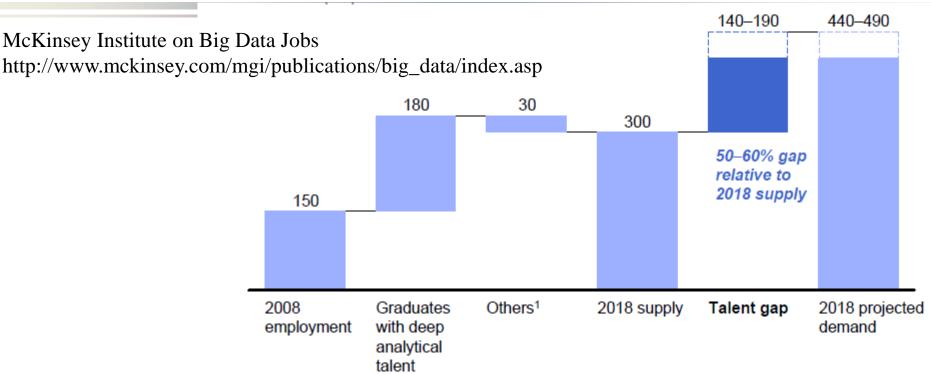
Topic 12: Policy development and international cooperation

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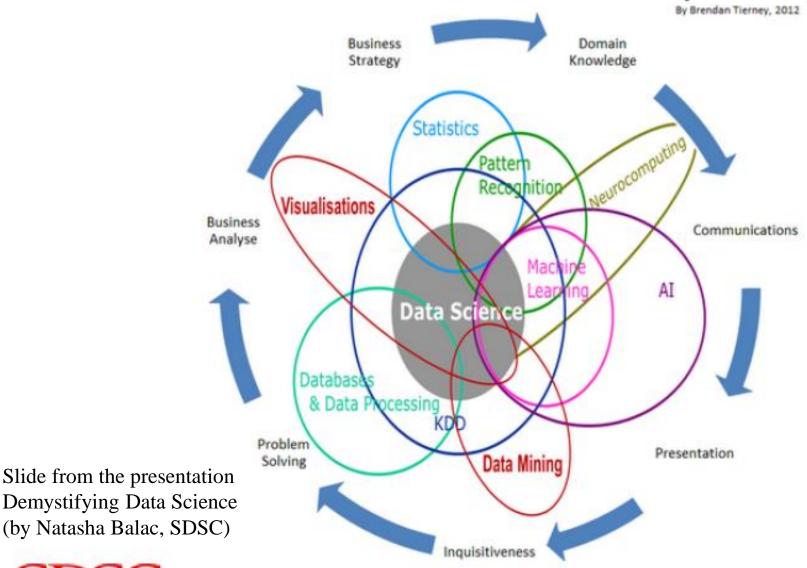
Data Scientist: New Profession and Opportunities



- There will be a shortage of talent necessary for organizations to take advantage of Big Data.
 - By 2018, the United States alone could face a shortage of 140,000 to 190,000 people with deep analytical skills as well as
 - 1.5 million managers and analysts with the know-how to use the analysis of big data to make effective decisions

SOURCE:US Bureau of Labor Statistics; US Census; Dun & Bradstreet; company interviews; McKinsey analysis

Data Science Is Multidisciplinary







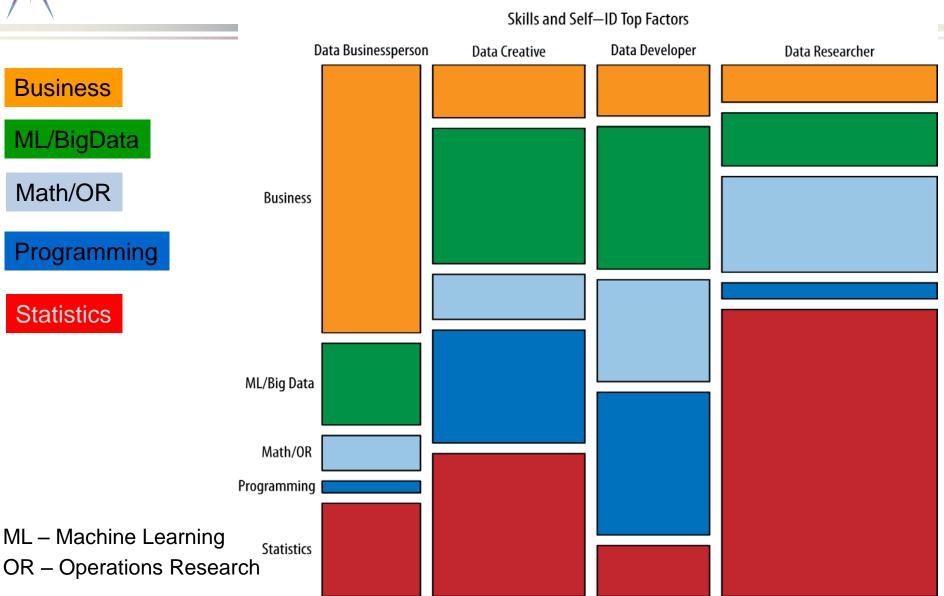


Strata Survey Skills and Data Scientist Self-ID

| | | Business | ML/Big | g Data | Math/OI | R | Programming | Statistics |
|--|-----------------|-------------------------------|--|----------------------|---|---|---|---|
| Analyzaina the Analyzaera O'D | | Product Developement Business | Unstruct Dat Struct Dat Mach | ta ured ta | Optimization Math Graphical Models | | Systems Administration Back End Programming Front End | Visualization Temporal Statistics Surveys and Marketing |
| Analysing the Analysers. O'Reilly Strata Survey – Harris, Murphy&Vaisman, 2013 Based on how data scientists think about themselves and their work Identified four Data Scientist clusters | | | Learr Big a Distrib Dat | ning and outed | Monte Carlo Statistics | | Programming | Spatial Statistics Science Data Manipulation |
| Data Developer | Developer | Engineer | | | | | | Classical |
| Data Researcher | Researcher | Scientist | | Statist | ician | | | Statistics |
| Data Creative | Jack of All Tra | des Artist | Hacker | | r | | | |
| Data Businessperson | Leader | Businessp | Businessperson Entre | | eneur | | | |



Skills and Self-ID Top Factors



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Key to a Great Data Scientist

Technical skills (Coding, Statistics, Math)

- + Commitment + Creativity
 - + Intuition
 - + Presentation Skills
 - + Business Savvy
- = Great Data Scientist!
- How Long Does It Take For a Beginner to Become a Good Data Scientist?
 - 3-5 years according to KDnuggets survey [278 votes total]



Summary Big Data

- Researching, learning mastering Big Data domain is a Big Data problem itself
- Cloud Computing as a native platform for Big Data
 - Acceptance of clouds will grow, so demand for specialists
- Demand for advanced high performance network will remain and grow
- New generically data centric models are required
- New distributed data processing and analytics computing models to be developed/re-factored
- Data Scientist is a new focus for talents search by companies



Теми презентації та питання для дискусії

- 1. Характеристики і можливості хмарних технологій, тенденції розвитку та стандартизація.
- 2. Приклади використання та типи впровадження комп'ютерних хмар: корпоративні, публічні, комунальні; міграція корпоративної ІТ інфраструктури на хмарну платформу, необхідні передумови і рівень "зрілості", переваги віртуалізації сервісів і ресурсів.
- 3. Законодавча та регуляторна база в Європі, програми підтримки впровадження хмар в Європі.
- 4. Глобальні провайдери хмарних послуг і ресурсів: Amazon AWS, Microsoft Azure, GoogleCloud: можливості, послуги, засоби розробки.
- 5. Великі Дані: Об'єм, Швидкість, Номенклатура, Мінливість, Цінність , Достовірність (Volume, Velocity, Variety, Variability, Value, Veracity).
- 6. Великі Дані та бізнес- аналітика: приклади використання і нові можливості.
- 7. Проблеми Великих Даних: зберігання, передача, обробка, контроль доступу, захист даних і персональної інформації.
- 8. Нові спеціальності для Хмарних технологій та Великих Даних: підготовка фахівців, тренінг та освіта.

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Questions and Discussion



- SNE Technical Report on Architecture Framework and Components for the Big Data Ecosystem. Draft Version 0.2, 12 September 2013
 http://www.uazone.org/demch/worksinprogress/sne-2013-02-techreport-bdaf-draft02.pdf
- Addressing Big Data Issues in Scientific Data Infrastructure. By Yuri Demchenko, et al. Paper at IEEE CTS 2013 Conf., May 20-24, 2013, San Diego, USA http://www.uazone.org/demch/papers/bddac2013-bigdata-infrastructure-v06.pdf
- BoF on Big Data Challenges for NREN's organised at TNC2013 https://tnc2013.terena.org/core/event/15
- NIST Big Data Working Group (NBD-WG) http://bigdatawg.nist.gov/home.php
 - Input documents http://bigdatawg.nist.gov/show_InputDoc2.php
 - Big Data Reference Architecture
 http://bigdatawg.nist.gov/_uploadfiles/M0226_v7_2611176301.docx
 - Big Data Architectures Survey
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 - Requirements for 51 usecases
 http://bigdatawg.nist.gov/uploadfiles/M0224_v1_1076079077.xlsx



NIST Documents on Cloud Computing (1)

- [NIST CC] NIST SP 800-145, "A NIST definition of cloud computing", [online] Available: http://csrc.nist.gov/publications/nistpubs/800-145/SP800-145.pdf
- [NIST CCRA] NIST SP 500-292, Cloud Computing Reference Architecture, v1.0.
 [Online] http://www.nist.gov/customcf/get_pdf.cfm?pub_id=909505
- [NIST Synopsis] DRAFT NIST SP 800-146, Cloud Computing Synopsis and Recommendations. [online] Available: http://csrc.nist.gov/publications/drafts/800-146/Draft-NIST-SP800-146.pdf
- Draft SP 800-144 Guidelines on Security and Privacy in Public Cloud Computing.
 [online] Available: http://csrc.nist.gov/publications/nistpubs/800-144/SP800-144.pdf
- NIST SP 500-299: NIST Cloud Computing Security Reference Architecture.
 [online] http://collaborate.nist.gov/twiki-cloud-computing/pub/CloudComputing/CloudSecurity/NIST_Security_Reference_Architecture.

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NIST Documents on Cloud Computing (2)

- NIST SP 500-293: US Government Cloud Computing Technology Roadmap Volume 1, High-Priority requirements to Further USG Agency Cloud Computing Adoption. http://www.nist.gov/itl/cloud/upload/SP_500_293_volumel-2.pdf
- NIST SP 500-293: US Government Cloud Computing Technology Roadmap Volume II, Useful Information for Cloud Adopters (Draft) http://www.nist.gov/itl/cloud/upload/SP_500_293_volumeII.pdf
- NIST SP 500-293: US Government Cloud Computing Technology Roadmap Volume III, Technical Considerations for USG Cloud Computing Deployment Decisions (Draft). http://collaborate.nist.gov/twiki-cloud-computing/pub/CloudComputing/RoadmapVolumeIIIWorkingDraft/NIST_cloud_roadmap_VIII_draft_110311.pdf

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