



dataflo

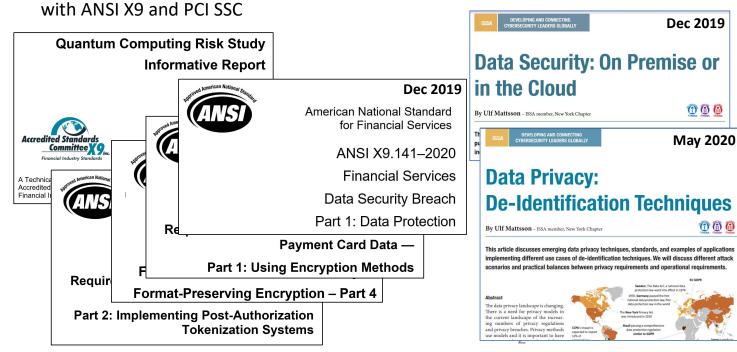
Ulf Mattsson

ISACA London Chapter

Standards Council

- Previously Head of Innovation at TokenEx and Chief Technology Officer at Atlantic BT, Compliance Engineering, and Protegrity, and IT Architect at IBM
- Products and Services:
 - Data Encryption, Tokenization, Data Discovery, Cloud Application Security Brokers (CASB), Web Application Firewalls (WAF), Robotics, and Applications
 - Security Operation Center (SOC), Managed Security Services (MSSP), and Security Benchmarking/Gap-analysis for Financial Industry

• Inventor of more than 70 issued US Patents and developed Industry Standards



Payment Card Industry (PCI) Security Standards Council (SSC):

- 1. Tokenization Task Force
- 2. Encryption Task Force, Point to Point Encryption Task
- Risk Assessment SIG
- 4. eCommerce SIG
- 5. Cloud SIG, Virtualization SIG
- 6. Pre-Authorization SIG, Scoping SIG Working Group

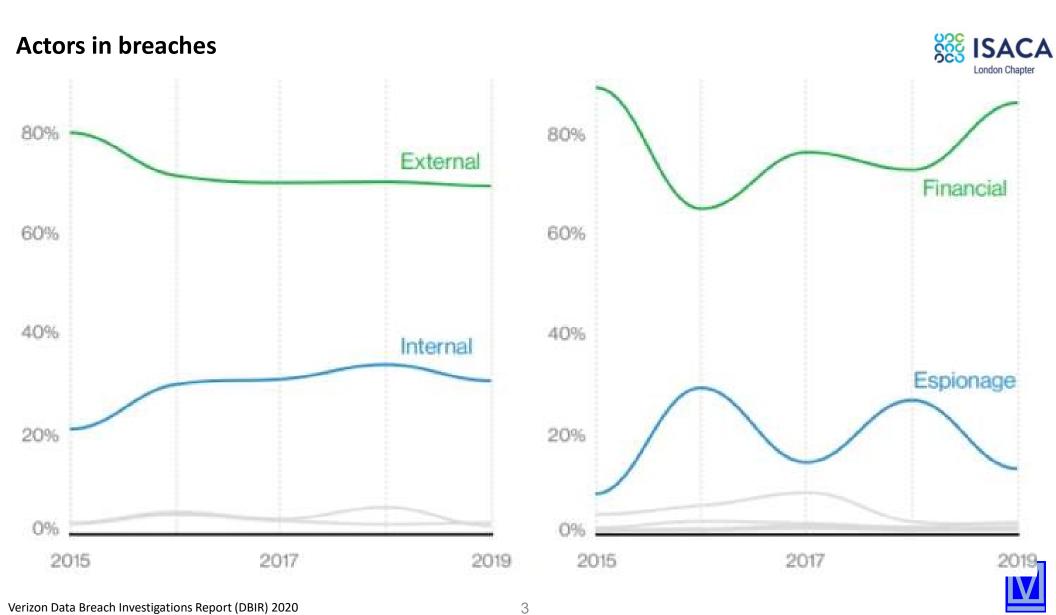
Practical Data Security and Privacy for GDPR and CCPA

Author: Ulf Mattson
Date Published: 13 May 2020

With sensitive data residing everywhere and the breach epidemic growing, the need for advanced data security solutions has become even more critical.
Compliance with regulations such as the EU General Data Protection Regulation (IOPRI). Payment Card Industry Data Security Standard (PCI DSS), US State

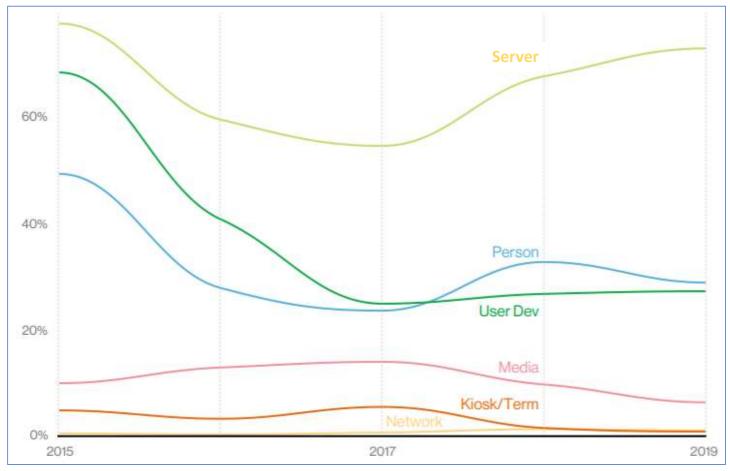


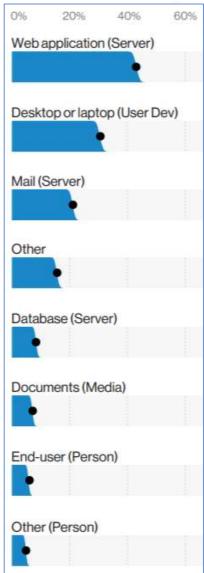




Assets in breaches

- On-premises assets are still 70% in our reported breaches dataset.
- Cloud assets were involved in about 24% of breaches.
 - Email or web application server 73% of the time.







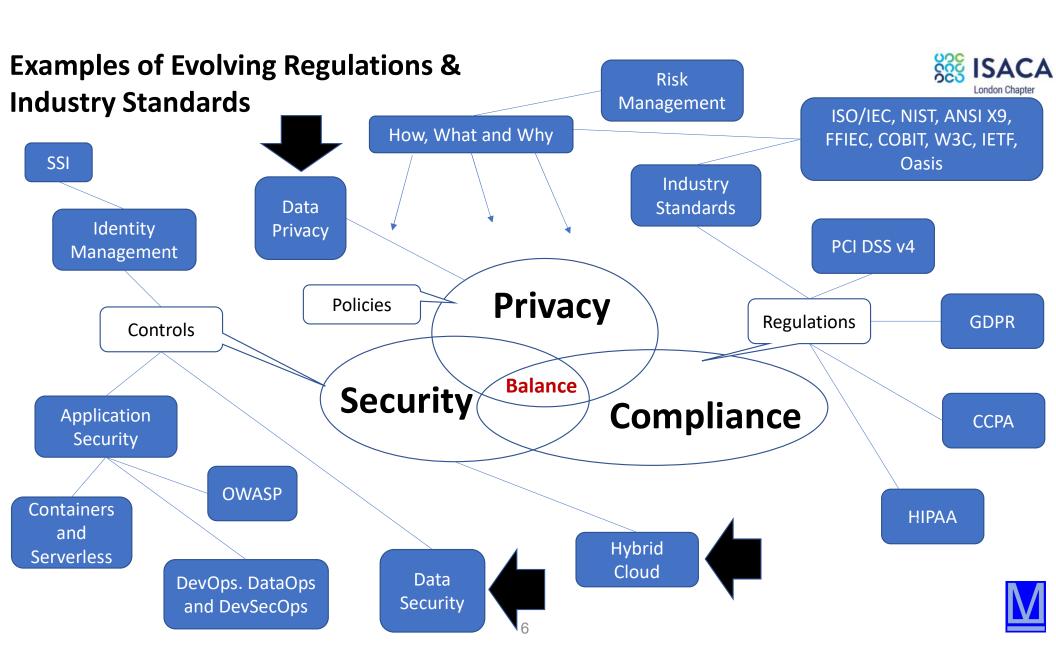


Action varieties in breaches



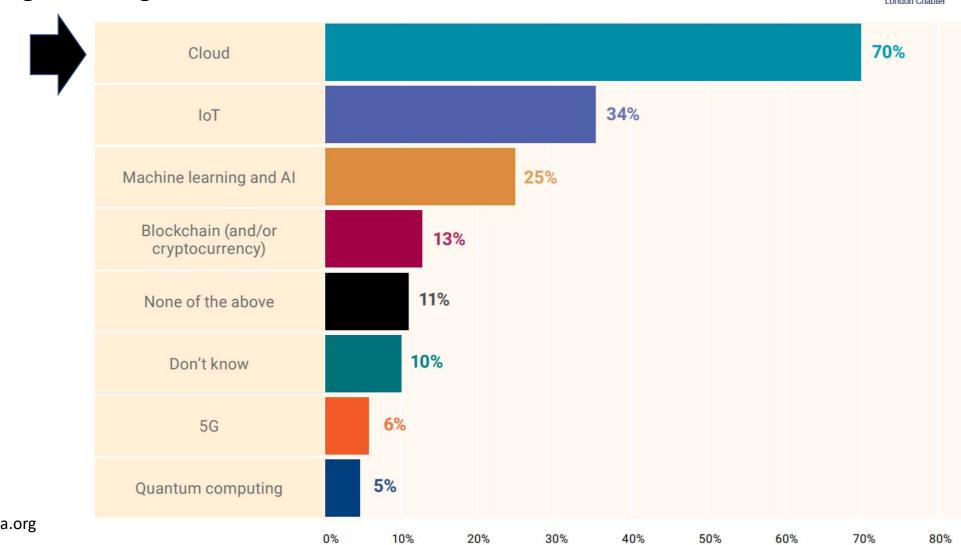






Emerging Technologies that Increase Risk

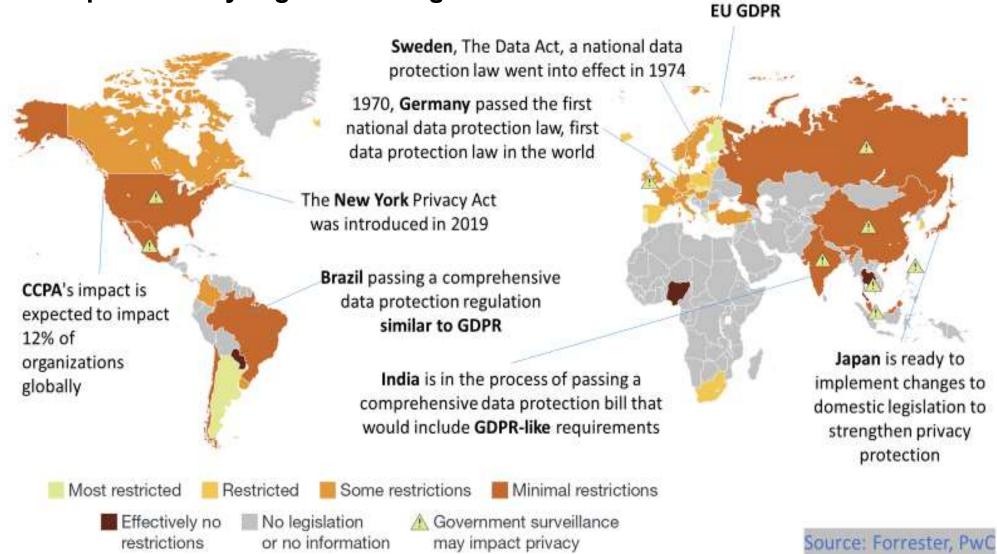




Source: www.isaca.org

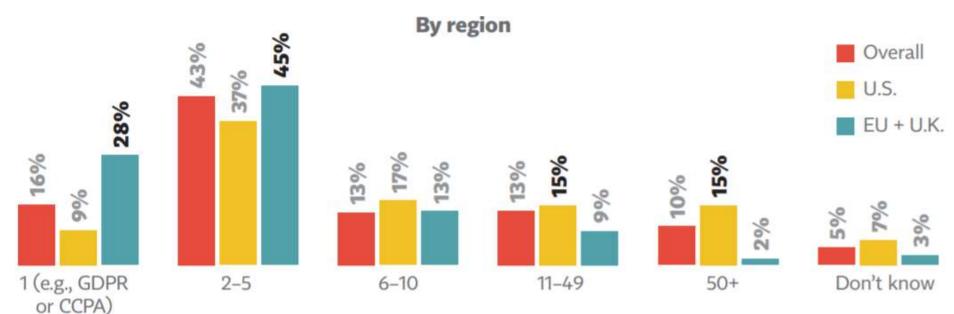


SS ISACA



How many privacy laws are you complying with?





General Data Protection Regulation (EU) 2016/679 (**GDPR**) is a regulation in EU law on data protection and privacy in the European Union (EU) and the European Economic Area (EEA). It also addresses the transfer of personal data outside the EU and EEA areas.

California Consumer Privacy Act (CCPA) is a bill that enhances privacy rights and consumer protection for residents of California, United States.





Legal and regulatory risks are exploding







80 New Laws in 2019



More Complex Rules

Category	North America	Latin America	EMEA	Asia Pacific	Totals
Comprehensive Data Protection		1	4	2	7
GDPR Implementation			9		9
Information Security	9		1	3	13
Health Privacy	4		2		6
Financial Privacy	5				5
Education Privacy	3				3
Breach	11	1			12
Privacy Rights	3	1		1	5
Other	8	1	9	2	20
Totals for 2019	43	4	25	8	80



TrustArc

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GDPR Security Requirements – Encryption and Tokenization

Design **Transform** Conform **Operate** Assess PREPARE: ROADMAP: PROTECT: MANAGE SECURITY **DEMONSTRATE:** Security requirements · Assess security current Create security Implement PROGRAM: · Demonstrate technical and state, identify gaps. remediation and privacy-enhancing organizational measures to · Manage and implement benchmark maturity. implementation plan controls (for example, ensure security appropriate security program establish conformance **SECURITY BY** encryption, tokenization, I to processing risk practices such as risk roadmaps dynamic masking) · Document security program: **DESIGN:** assessment, roles and · Identify vulnerabilities, · Implement security ongoing monitoring, responsibilities, program · Create security reference supporting Security by controls; mitigate access effectiveness assessment, evaluation and architecture Design risks and security reporting of security controls Design Technical and RUN SERVICES: vulnerabilities and activities DISCOVER: Organizational Measures · Monitor security Discover and classify RESPOND: (TOMs) appropriate to operations and personal data assets and risk (such as encryption. · Respond to and manage intelligence: monitor. affected systems to breaches pseudonimization, access detect, respond to and design security controls control, monitoring) mitigate threats · Govern data incident response and forensics practices **Security Encryption and Discover Tokenization** by Design **Data Assets**

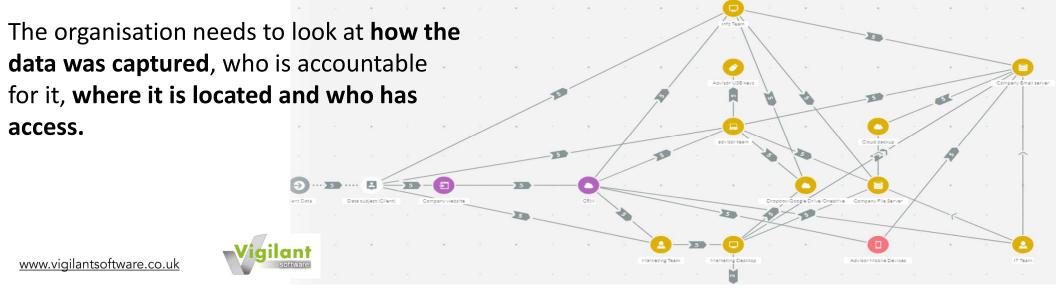


Data flow mapping under GDPR



- If there is not already a **documented workflow** in place in your organisation, it can be worthwhile for a team to be sent out to identify how the data is being gathered.
- This will enable you to see how your data flow is different from reality and what needs
 to be done to amend this.

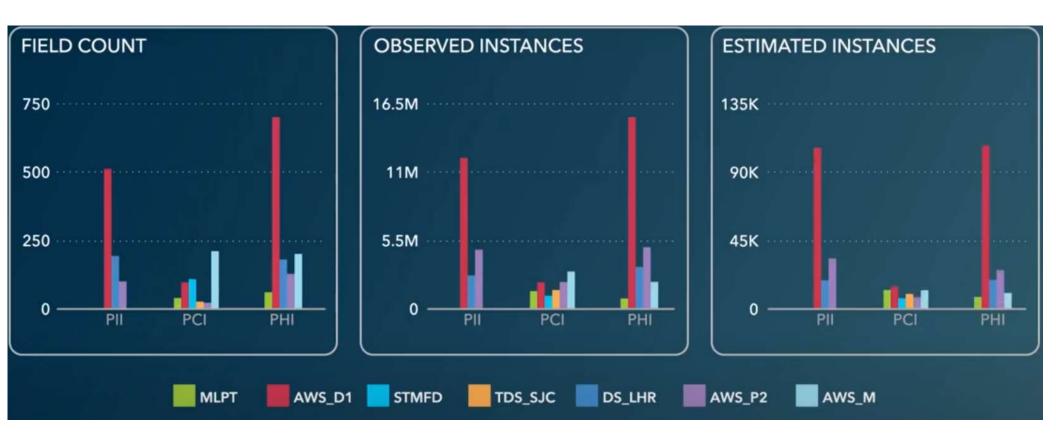
If an organisation's theory about how its data is flowing is different from the reality, you have a breach and could be fined.





See security improvement and track sensitive data







www.protegrity.com 14

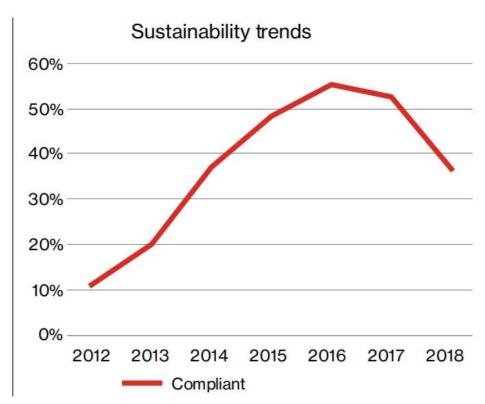


Protecting Sensitive Data



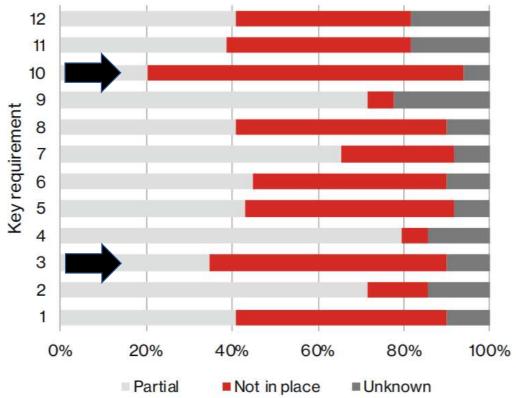
PCI DSS Compliance Issues with breached organizations and PCI DSS v4







- Meeting the security intent of PCI DSS by using security approaches that may be different than traditional PCI DSS requirements.
- Compensating controls will be removed

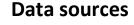


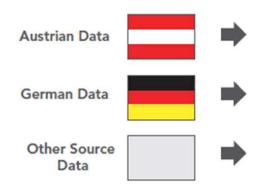
- PCI DSS Requirement 3 is addressing protecting cardholder data.
- PCI DSS Requirement 10 is addressing network security and access.

Source: Verizon 2019 Payment Security Report

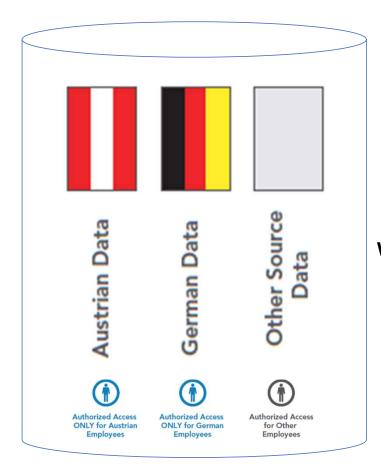
Example of Cross Border Data-centric Security







- Protecting Personally Identifiable Information (PII), including names, addresses, phone, email, policy and account numbers
- Compliance with EU Cross Border Data Protection Laws
- Utilizing Data **Tokenization**, and centralized policy, key management, auditing, and reporting



Complete policyenforced deidentification of sensitive data across all bank entities

Data Warehouse In Italy



How Coronavirus Is Eroding Privacy Technology to track and monitor individuals aims to slow

pandemic, but raises concerns about government overreach

European nations monitor citizen movement by tapping telecommunications data that they say conceals individuals' identities.

American officials are drawing cellphone location data from mobile advertising firms to track the presence of crowds—but not individuals. Apple Inc. and Alphabet Inc.'s Google recently announced plans to launch a voluntary app that health officials can use to reverse-engineer sickened patients' recent whereabouts—provided they agree to provide such information.

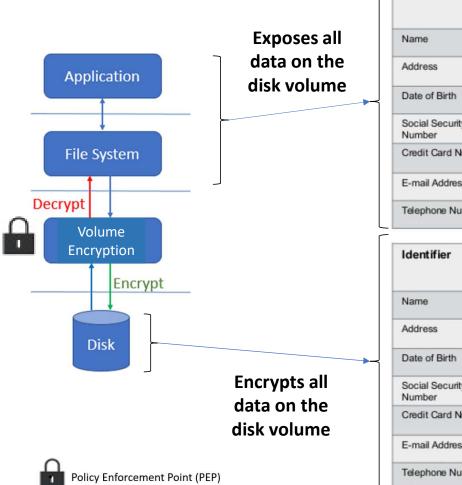
The extent of tracking hinges on a series of tough choices: **Make it voluntary or mandatory?** Collect personal or **anonymized data? Disclose information** publicly or privately?

In Western Australia, lawmakers approved a bill last month to install surveillance gadgets in people's homes to monitor those placed under quarantine. Authorities in Hong Kong and India are using geofencing that draws virtual fences around quarantine zones. They monitor digital signals from smartphone or wristbands to deter rule breakers and nab offenders, who can be sent to jail. Japan's most popular messaging app beams health-status questions to its users on behalf of the government.

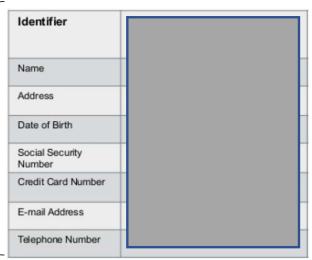


Example of disk level encryption





Identifier	Clear
Name	Joe Smith
Address	100 Main Street, Pleasantville, CA
Date of Birth	12/25/1966
Social Security Number	076-39-2778
Credit Card Number	3678 2289 3907 3378
E-mail Address	joe.smith@surferdude.org
Telephone Number	760-278-3389



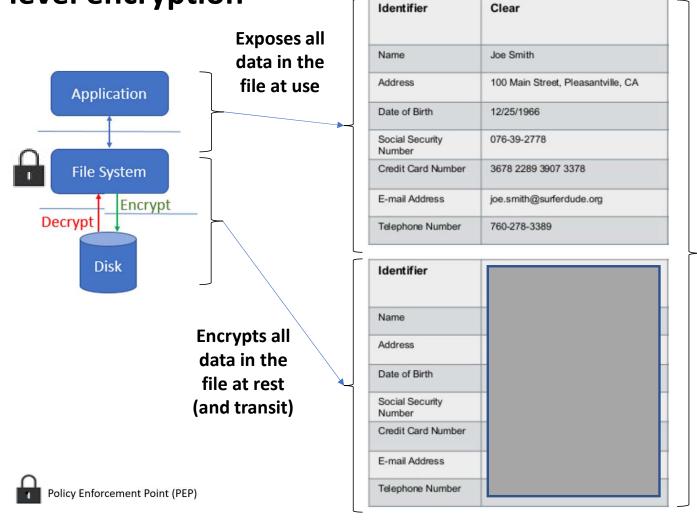


- "All or nothing" approach
- · Does NOT secure file contents in use



Example of file level encryption



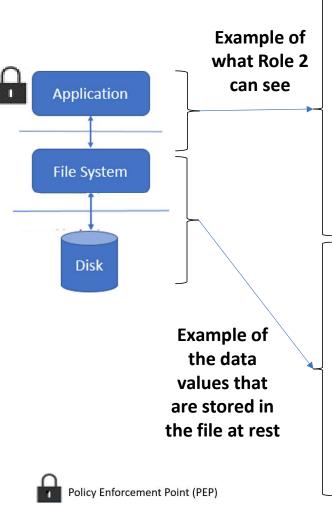




- · "All or nothing" approach
- · Does NOT secure file contents in use
- · OS File System Encryption
- · HDFS Encryption
- · Secures data at rest and in transit



Reduce risk by field level protection



Identifier	Authorized Role 2 * Can see limited data in the clear
Name	Joe Smith
Address	"No Access"
Date of Birth	01/02/1966
Social Security Number	xxxxx2278
Credit Card Number	3846 2290 3371 3890
E-mail Address	joe.smith@surferdude.org
Telephone Number	998-389-2289

Identifier	Protected		
Name	csu wusoj		
Address	476 srta coetse, cysieondusbak, HA		
Date of Birth	01/02/1966		
Social Security Number	478-39-8920		
Credit Card Number	3846 2290 3371 3890		
E-mail Address	eoe.nwuer@beusorpdqo.aku		
Telephone Number	998-389-2289		

Reduce risk by **not exposing the full data value** to applications and users that only need to operate on a limited representation of the data

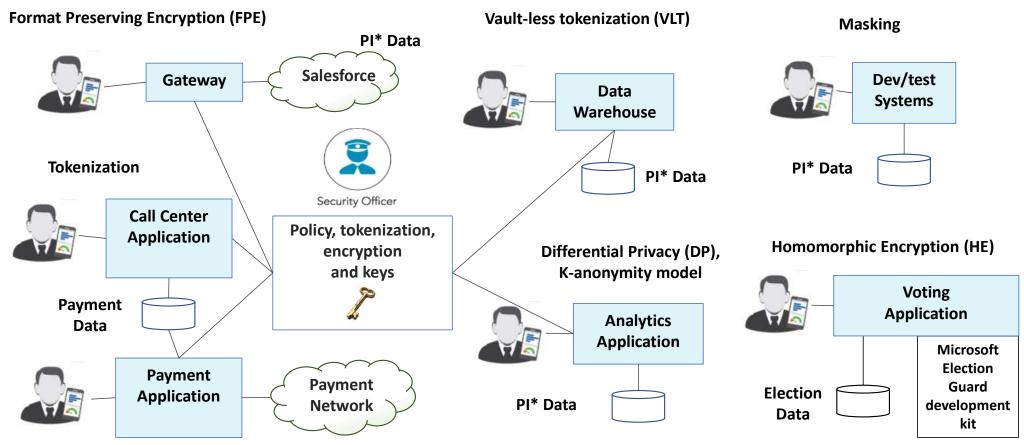


- · At the individual field level
- · Fine Grained Protection Methods:
 - Vaultless Tokenization
 - Encryption
 - · Format Preserving Encryption
 - · Masking/Data Obfuscation
- · Data is protected wherever it goes (even In-Use)
- Business intelligence analytics capability retained (80%-90% of analytics performed on data in protected form)



Use-cases of some de-identification techniques





^{*:} PI Data (Personal information) means information that identifies, relates to, describes, is capable of being associated with, or could reasonably be linked, directly or indirectly, with a consumer or household according to CCPA



Privacy Standards

11 Published International Privacy Standards (ISO)



Techniques	20889 IS Privacy enhancing de-identification terminology and classification of techniques	
Management	27701 IS Security techniques - Extension to ISO/IEC 27001 and ISO/IEC 27002 for privacy information management - Requirements and guidelines	
	27018 IS Code of practice for protection of PII in public clouds acting as PII processors	
Cloud	29100 IS Privacy framework	
	29101 IS Privacy architecture framework	IS: International Standard
Framework	29134 IS Guidelines for Privacy impact assessment	
	29190 IS Privacy capability assessment model	
Impact	29191 IS Requirements for partially anonymous, partially unlinkable authentication	
Requirements	29151 IS Code of Practice for PII Protection	Guidelines to help comply with ethical standards
	19608 TS Guidance for developing security and privacy functional requirements based on 15408	TS: Technical Specification
Process	27550 TR Privacy engineering for system lifecycle processes	TR: Technical Report

Privacy Standards

16 International Privacy Standards in development (ISO)



Big Data	
Dig Dutu	

27046 IS Big data security and privacy - Implementation guidelines

20547 IS Big data reference architecture - Part 4 - Security and privacy

27045 IS Big data security and privacy - processes

Framework

27559 IS Privacy-enhancing data de-identification framework

Design

27556 IS User-centric framework for the handling of personally identifiable information (PII) based on privacy preferences

31700 IS Consumer Protection - Privacy-by-design for consumer goods and services

Risk

27557 IS Organizational privacy risk management

Consent and Deletion

27555 IS Guidelines on Personally Identifiable Information Deletion

29184 IS Online privacy notices and consent

IoT

27560 TS Privacy technologies - Consent record information structure

27402 IS IoT security and privacy - Device baseline requirements

Authentication

23491 IS Security techniques - IoT security and privacy - Guidelines for IoT domotics

Smart Cities

27030 IS Security and Privacy for the Internet of Things

27551 IS Requirements for attribute-based unlinkable entity authentication

27570 TS Privacy Guidelines for Smart Cities

Audit

27006-2 (formerly 27558 IS) TS Information security, cybersecurity and privacy protection - Requirements audit and certification of privacy information management systems

IS: International Standard

TS: Technical Specification



Data protection techniques: Deployment on-premises, and clouds



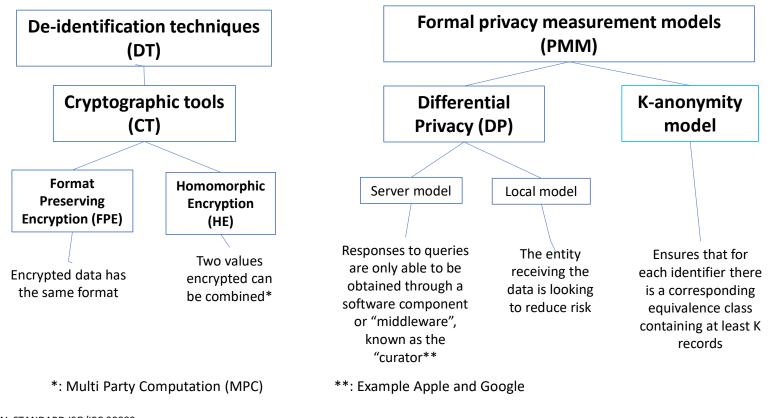
_	•	de-identification Ition of techniques	Data Warehouse	Centralized	Distributed	On- premises	Public Cloud	Private Cloud
		Vault-based tokenization		у				у
	Tokenization	Vault-less tokenization	у	у	у	у	у	у
De- identification	Cryptographic	Format preserving encryption		у	у	у	у	у
techniques	tools	Homomorphic encryption			у		у	
	Suppression	Masking	у	у	у	у	у	у
	techniques	Hashing	у	у	у	у	у	у
Formal	Differential	Server model	у	у	у	у	у	у
privacy	Privacy	Local model	у	у	у	у	у	у
measurement models	K-anonymity	L-diversity	у	у	у	у	у	у
models	model	T-closeness	у	у	у	у	у	у



ISO Standard for Encryption and Privacy Models



Privacy enhancing data de-identification terminology and classification of techniques





Examples of data de-identification

Source data:

				SSS ISACA
Patient	Age	Gender	Region	Disease on Chapter
173965429	57	Female	Hamburg	Gastric ulcer

Source data:

Last name	Balance	Age	Gender
Folds	93791	23	m
	•••	•••	•••



Generalization

Output data:

Patient	Age	Gender	Region	Disease
173965429	>50	Female	Germany	Gastric ulcer





Output data:



Rounding

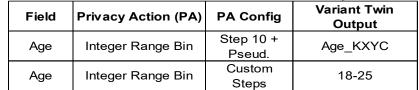
| | |

Pseudonymization

Aggregation/Binning

Field	Privacy Action (PA)	PA Config	Variant Twin Output
Gender	Pseudonymise		AD-lks75HF9aLKSa

Field	Privacy Action (PA)	PA Config	Variant Twin Output	
Balance	Nearest Unit Value	Thousand	94000	









Source: INTERNATIONAL STANDARD ISO/IEC 20889, Privitar, Anonos



Privacy Measurement Models

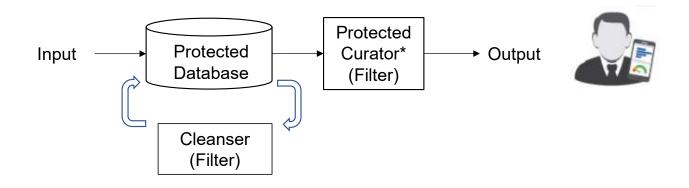


Privacy measurement models



Differential Privacy

Differential privacy is a model that provides mathematical guarantees that the probability distribution of the output of this analysis differs by a factor no greater than a specified parameter regardless of whether any data principal is included in the input dataset.



*: Example: Apple



Differential privacy model



Differential privacy is a formal privacy measurement model that, if incorporated in the **design of a particular statistical analysis**, provides mathematical **guarantees that the probability distribution of the output** of this analysis **differs by a factor no greater than a specified parameter** regardless of whether **any** particular data principal **is included** in the input dataset.

- a **mathematical definition of privacy** which posits that, for the outcome of any statistical analysis distribution independent of whether any given data principal is added to or removed from the dataset; and
- a measure of privacy that enables monitoring of cumulative privacy loss and setting of a "budget" for loss limit.

When adequately implemented and used, provide a mathematically proven guarantee of privacy.

The design and construction of a differentially private algorithm requires appropriate **expertise** in the field of **probability and statistics**, and of the **theory of differential privacy**.

Differentially private algorithms are built by **adding** a certain amount of "random **noise**" that is generated **from** a **carefully** selected probability distribution, such that the desired **usefulness** of data is **preserved**.

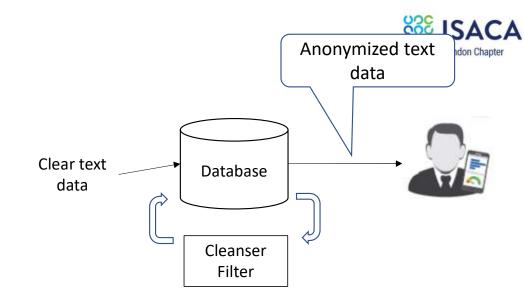


Privacy measurement models

K-anonymity model

The k-anonymity model that ensures that **groups** smaller than k individuals **cannot be identified**.

 Queries will return at least k number of records. Kanonymity is a formal privacy measurement model that ensures that for each identifier there is a corresponding equivalence class containing at least K records.



Some of the **de-identification techniques** can be used either independently or **in combination** with each other to **satisfy the K-anonymity model**.

Suppression techniques, generalization techniques, and **microaggregation*** can be applied to different types of attributes in a dataset to achieve the desired results.

*: Microaggregation replaces all values of continuous attributes with their averages computed in a certain algorithmic way.



Privacy measurement models

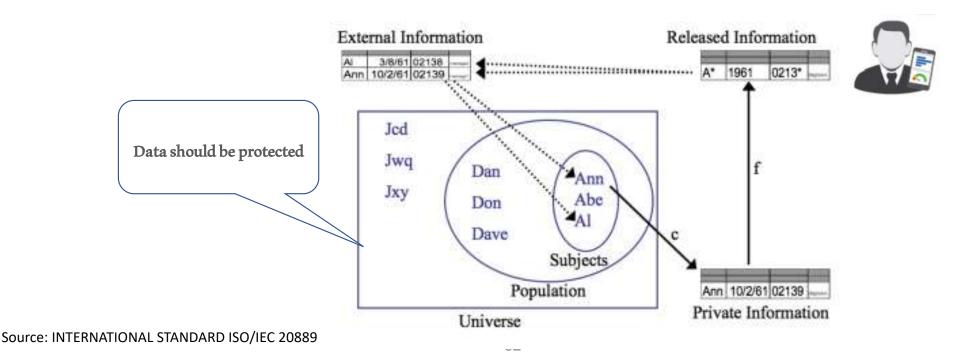
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K-anonymity model

The k-anonymity can thwart the ability to link field-structured databases

Given person-specific field-structured data, produce a release of the data with scientific guarantees that the individuals who are the subjects of the data cannot be reidentified while the data remain practically useful.

A release provides k-anonymity if the data for **each person cannot be distinguished from at least k-1 individuals** whose data also appears in the release





Risk reduction and truthfulness of standardized de-identification techniques and models

Technique name		Use Case / User Story	Data protected in			Data truthfulness at record	Applicable to types of attributes	Reduces the fisk CA		
			Transit	Use	Storage	level	attributes	Singling out	Linking	Inference
Pseudonymization	Tokenization	Protects the data flow from attacks	Yes	Yes	Yes	Yes	Direct identifiers	No	Partially	No
Cryptographic tools	Deterministic encryption	Protects the data when not used in processing operations	Yes	No	Yes	Yes	All attributes	No	Partially	No
	Order-preserving encryption	Protects the data from attacks	Partially	Partially	Partially	Yes	All attributes	No	Partially	No
	Homomorphic encryption	Protects the data also when used in processing operations	Yes	Yes	Yes	Yes	All attributes	No	No	No
Suppression	Masking	Protects the data in dev/test and analytical applications	Yes	Yes	Yes	Yes	Local identifiers	Yes	Partially	No
	Local suppression	Protects the data in analytical applications	Yes	Yes	Yes	Yes	ldentifying attributes	Partially	Partially	Partially
	Record suppression	Removes the data from the data set	Yes	Yes	Yes	Yes	All attributes	Yes	Yes	Yes
	Sampling	Exposes only a subset of the data for analytical applications	Partially	Partially	Partially	Yes	All attributes	Partially	Partially	Partially
Generalization	Generalization	Protects the data in dev/test and analytical applications	Yes	Yes	Yes	Yes	ldentifying attributes	Partially	Partially	Partially
	Rounding	Protects the data in dev/test and analytical applications	Yes	Yes	Yes	Yes	ldentifying attributes	No	Partially	Partially
	Top/bottom coding	Protects the data in dev/test and analytical applications	Yes	Yes	Yes	Yes	ldentifying attributes	No	Partially	Partially
Randomization	Noise addition	Protects the data in dev/test and analytical applications	Yes	Yes	Yes	No	ldentifying attributes	Partially	Partially	Partially
	Permutation	Protects the data in dev/test and analytical applications	Yes	Yes	Yes	No	ldentifying attributes	Partially	Partially	Partially
	Micro aggregation	Protects the data in dev/test and analytical applications	Yes	Yes	Yes	No	All attributes	No	Partially	Partially
Privacy models	Differential privacy	Protects the data in analytical applications	No	Yes	Yes	No	ldentifying attributes	Yes	Yes	Partially
	K-anonymity	Protects the data in analytical applications	No	Yes	Yes	Yes	Quai identifiers	Yes	Partially	No

Source: INTERNATIONAL STANDARD ISO/IEC 20889

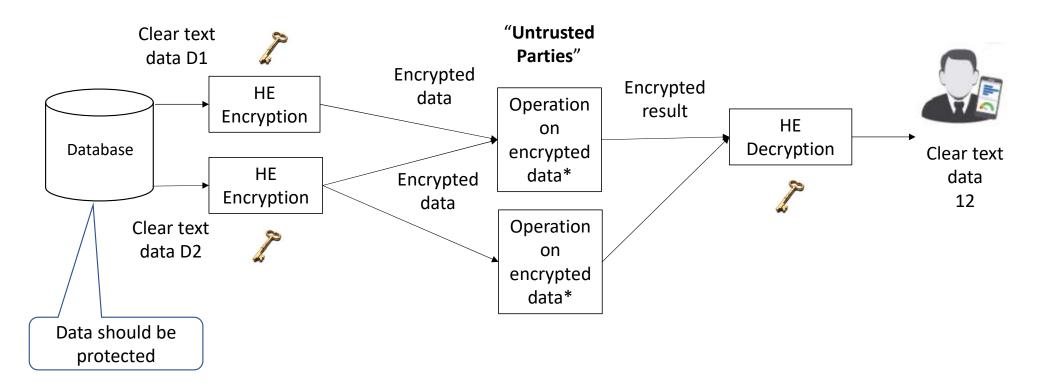


Encryption and Tokenization



Homomorphic Encryption (HE)







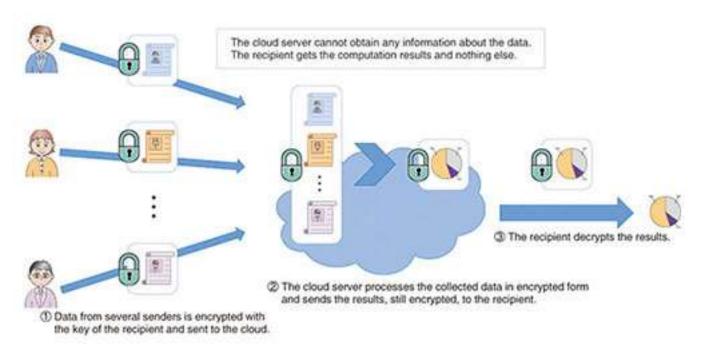
Homomorphic Encryption (HE)



Anonymous data processing with fully homomorphic encryption

Anonymous data processing involves **multiple users sending some sensitive data to a cloud server**, where it is aggregated, stripped of identifying information, and analyzed, typically to extract some statistical information, which is then delivered to the final recipient.

• The security requirement is that the cloud **server must learn nothing about the content of users' data**, and the **recipient must obtain only the anonymized results** of the statistical analysis, and, no **information on individual users**.

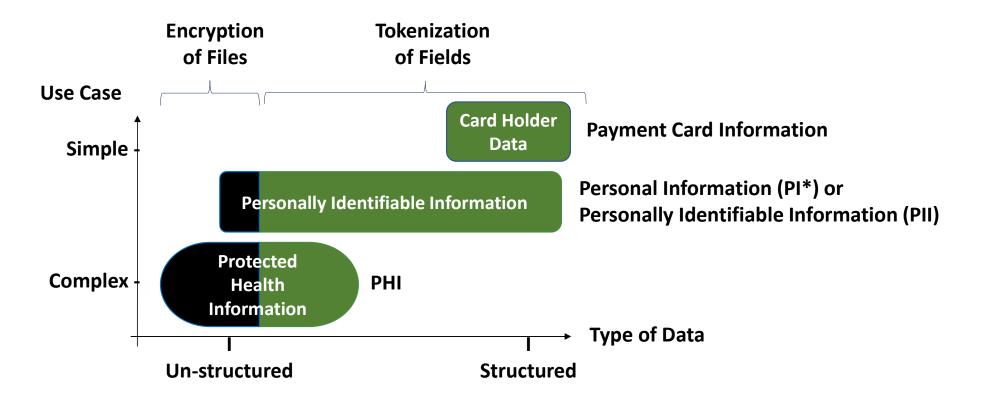




www.ntt-review.jp

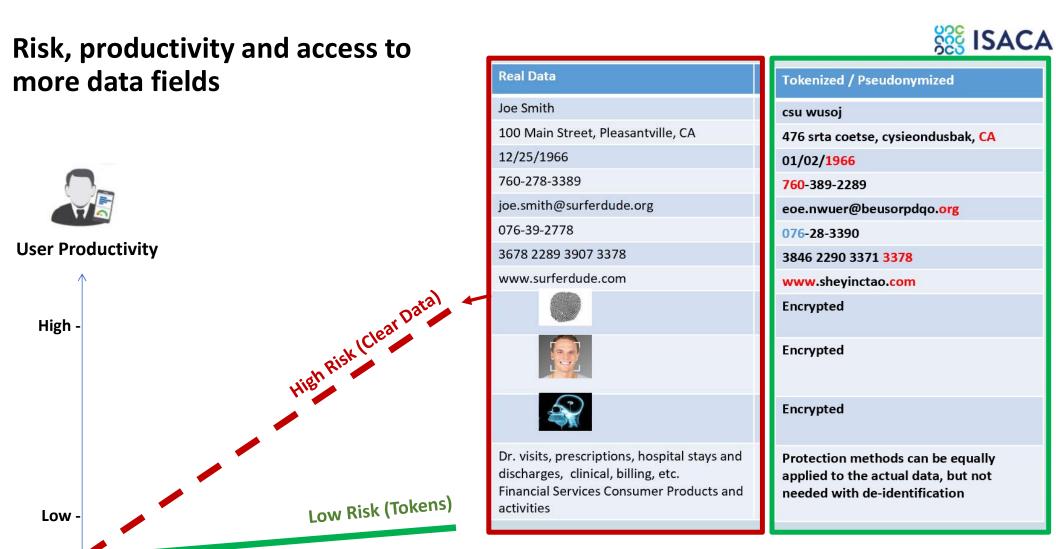
How to protect different types of data with encryption and tokenization





*: California CCPA







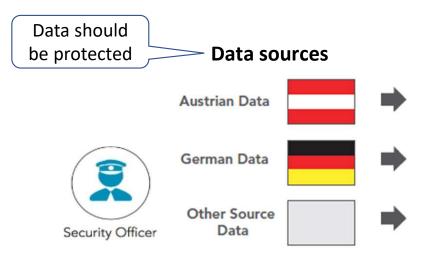
High

Low

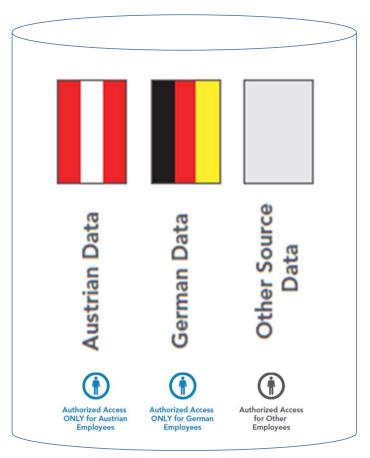
Access to Data Sources / Fields

Example of Cross Border Data-centric Security using Tokenization





- Protecting Personally Identifiable Information (PII), including names, addresses, phone, email, policy and account numbers
- Compliance with EU Cross Border Data Protection Laws
- Utilizing Data **Tokenization**, and centralized policy, key management, auditing, and reporting





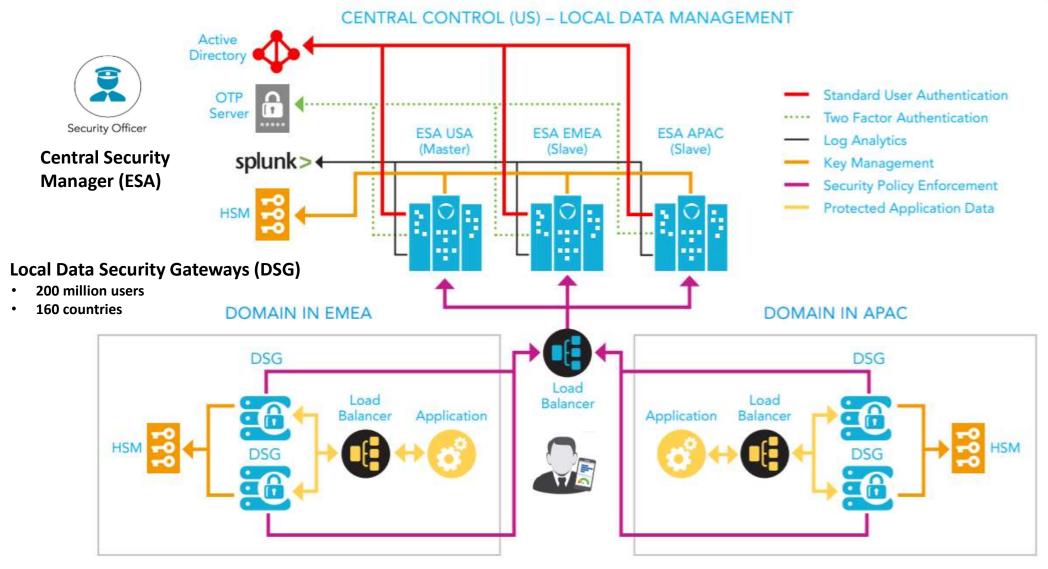
Data Warehouse

Complete policyenforced deidentification of sensitive data across all bank entities



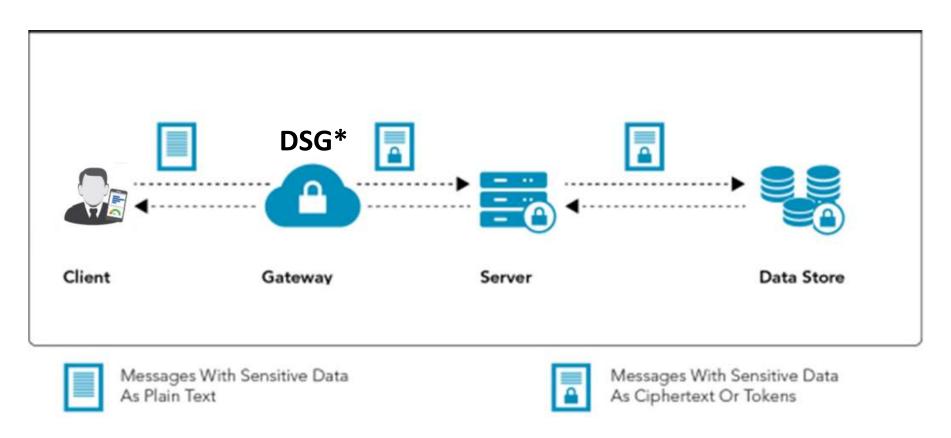
Use Case - Compliance with cross-border and other privacy restrictions







A Data Security Gateway (DSG) can turn sensitive data to Ciphertext or Tokens



^{*:} Example of supported protocols include HTTP, HTTPS, SFTP, SMTP and API utilizing web services or REST



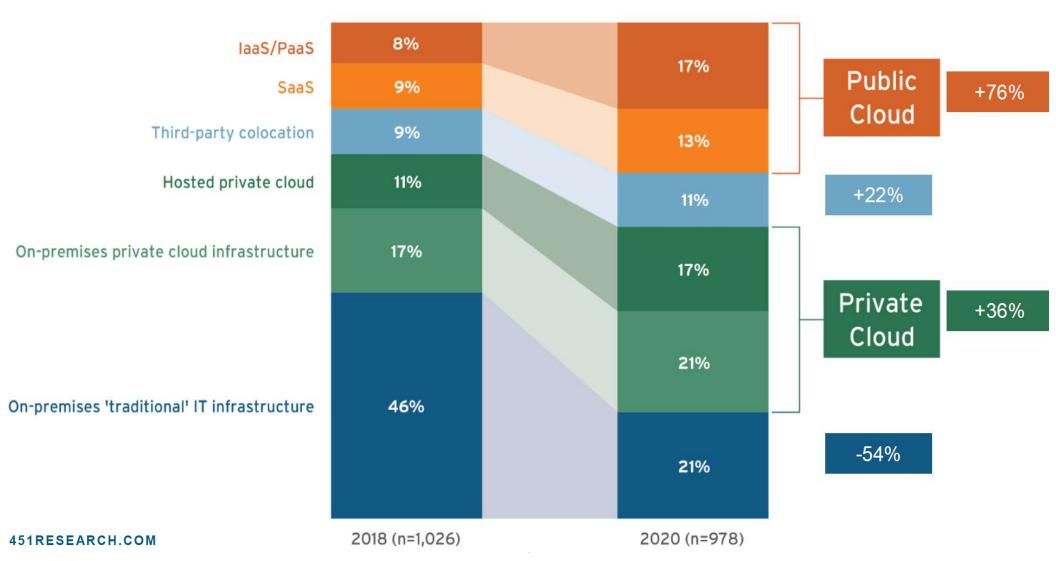


Trends in Cloud Security



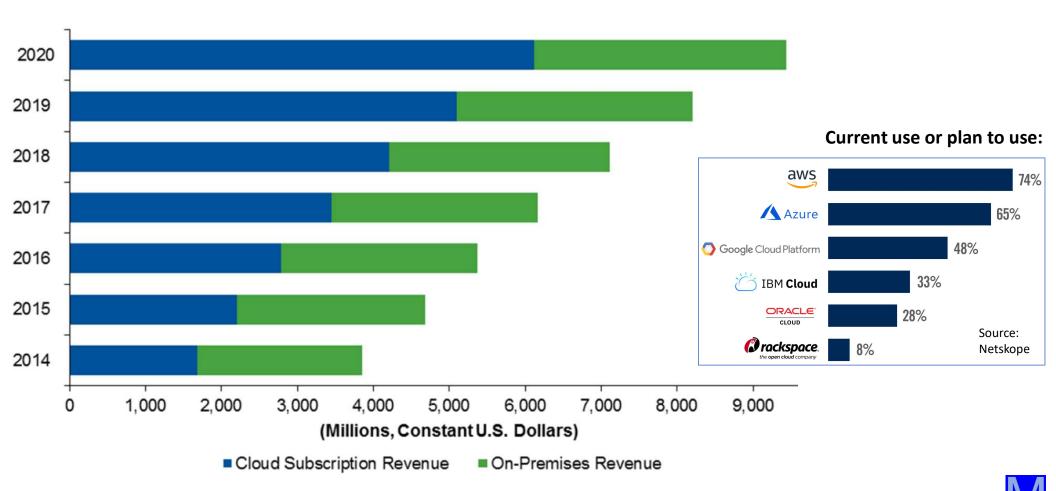
Cloud transformations are accelerating





Spending by Deployment Model, Digital Commerce Platforms, Worldwides ISACA





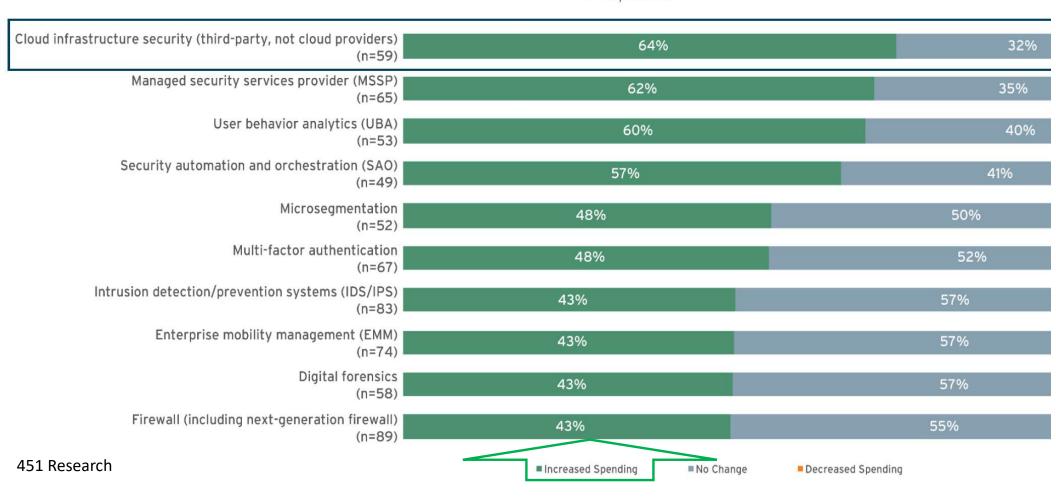
Source: Gartner

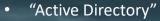
Securing Cloud Workloads – Greatest Increase in Spending



ANTICIPATED CHANGES IN SPENDING ON INFORMATION SECURITY TECHNOLOGIES IN NEXT 12 MONTHS

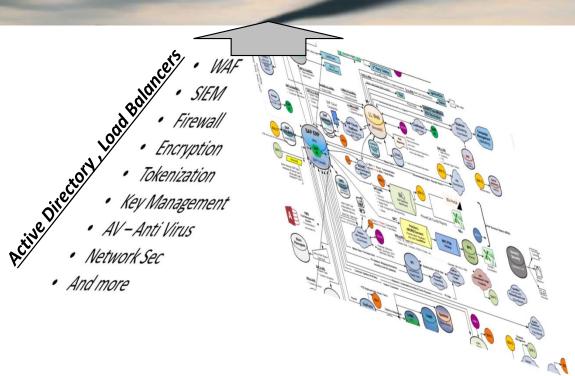
% of respondents





- WAF
- SIEM
- Firewall
- Encryption
- Tokenization
- Key Management
- AV Anti Virus
- Network Sec

Security controls can be applied On-premises or Cloud



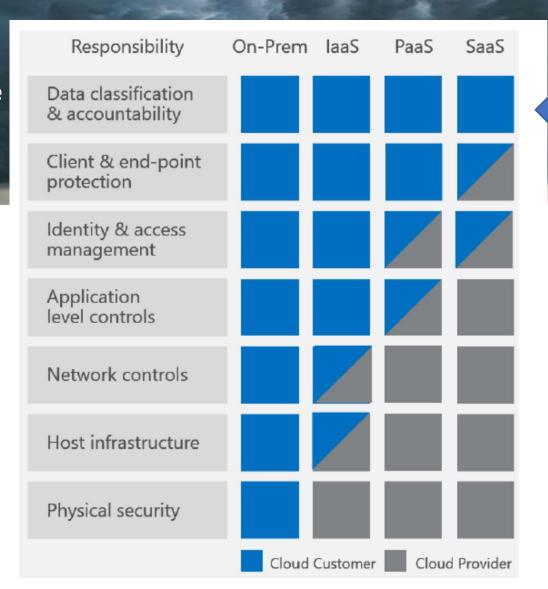


Microsoft Azure aWS

Google Cloud

Public Cloud / Multi-cloud

Shared responsibilities across cloud service models





The Customer is
Responsible for the
Data across all
Cloud Service
Models



Source: Microsoft

Multi-Cloud Key Management considerations

Legal Compliance and Nation-State Attacks



- Many companies have information that is attractive to governments and intelligence services.
- Others worry that litigation may result in a subpoena for all their data.

Jurisdiction

- Cloud service providers, especially laaS vendors, offer services in multiple countries, often in more than one region, with redundant data centers.
 - This redundancy is great for resilience, but regulatory concerns arises when moving data across regions which may have different laws and jurisdictions.



Securosis, 2019

Multi-Cloud Key Management considerations



Consistency

Most firms are quite **familiar with their on-premises encryption and key management** systems, so they often prefer to leverage the same tool and skills across multiple clouds.

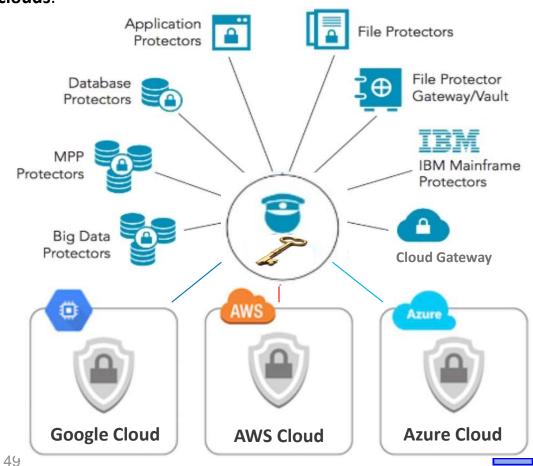
Firms often adopt a "best of breed" cloud approach.

Trust

Some customers simply **do not trust their vendors**.

Vendor Lock-in and Migration

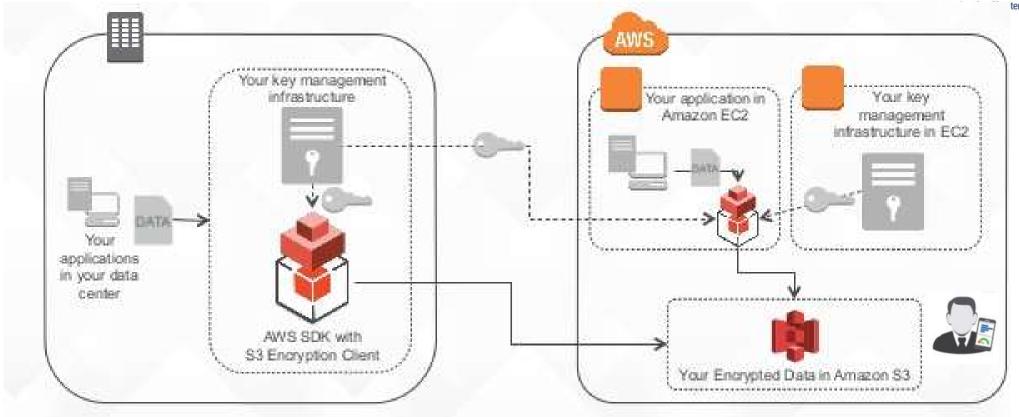
- A common concern is **vendor lock-in**, and an inability to migrate to another cloud service provider.
 - Some native cloud encryption systems do not allow customer keys to move outside the system, and cloud encryption systems are based on proprietary interfaces.
 - The goal is to maintain **protection** regardless of where data resides, moving between cloud vendors.



Securosis, 2019

Amazon S3 client-side encryption



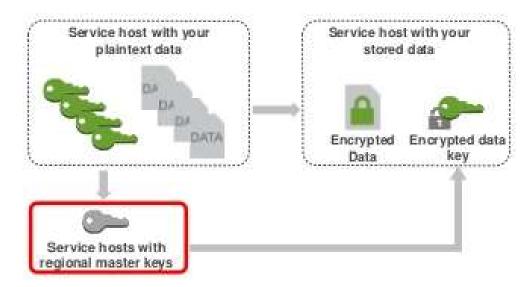


- Amazon S3 encryption and decryption takes place in the **EMRFS client** on your cluster.
- Objects are encrypted before being uploaded to Amazon S3 and decrypted after they are downloaded.
- The EMR (Elastic MapReduce) File System (EMRFS) is an implementation of HDFS that all Amazon EMR clusters use for reading and writing regular files from Amazon EMR directly to Amazon S3.

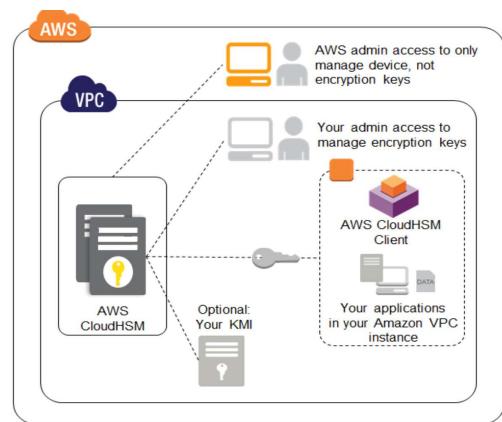


AWS encryption key management

- AWS service generates unique 256-bit
 AES data key per object, archive, cluster or database
- Service uses regularly rotated, regional 256-bit AES master keys to encrypt data keys
- Your encrypted data key is stored with your encrypted data



- Amazon Virtual Private Cloud (Amazon VPC) lets you provision a logically isolated section of the AWS Cloud where you can launch AWS resources
- Your virtual networking environment allows selection of your own IP address range, creation of subnets, and configuration



Multi-Cloud Key Management Encryption Key Management Computing Cost Encryption/decryption Point Risk Security Officer Protected data Security Officer - High Low -**On-premises On-premises Private Cloud** Hosted **Private Cloud** - Low High -**Public Cloud**



Elasticity

Out-sourced

In-house

A Cloud Security Gateway (CASB) can protect sensitive data in Cloud (SaaS)

Tokenization

or Encryption





Salesforce.com

- Example of supported protocols include HTTP, HTTPS, SFTP, and SMTP
- · Based on configuration instead of programming
- Secures existing web services or REST API calls
- See and control where sensitive data travels

- Install the Cloud Security Gateway in your trusted domain
- 2. Select the fields to be protected
- 3. Start using Salesforce with enhanced security



Encryption Key Management



Policy Enforcement Point (PEP)



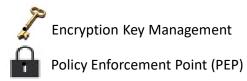
Protected data fields

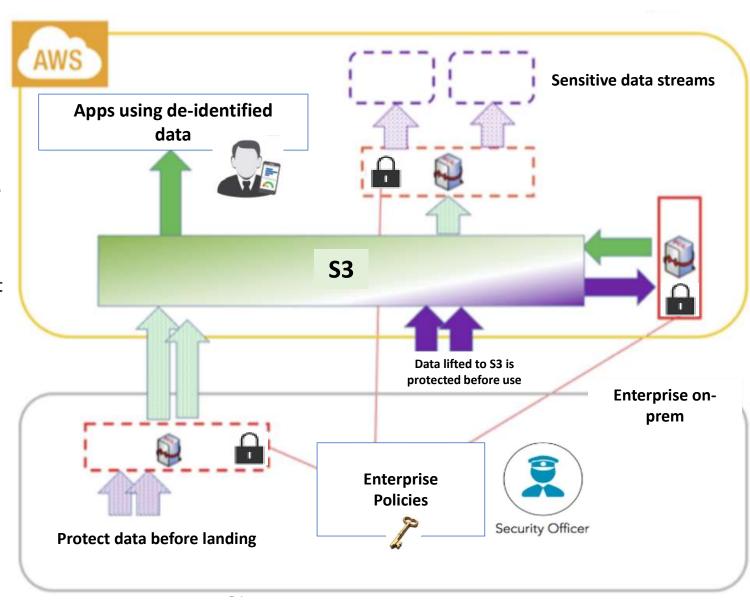


Protection of data in AWS S3 with Separation of Duties

- Applications can use deidentified data or data in the clear based on policies
- Protection of data in AWS S3 before landing in a S3 bucket

Separation of Duties

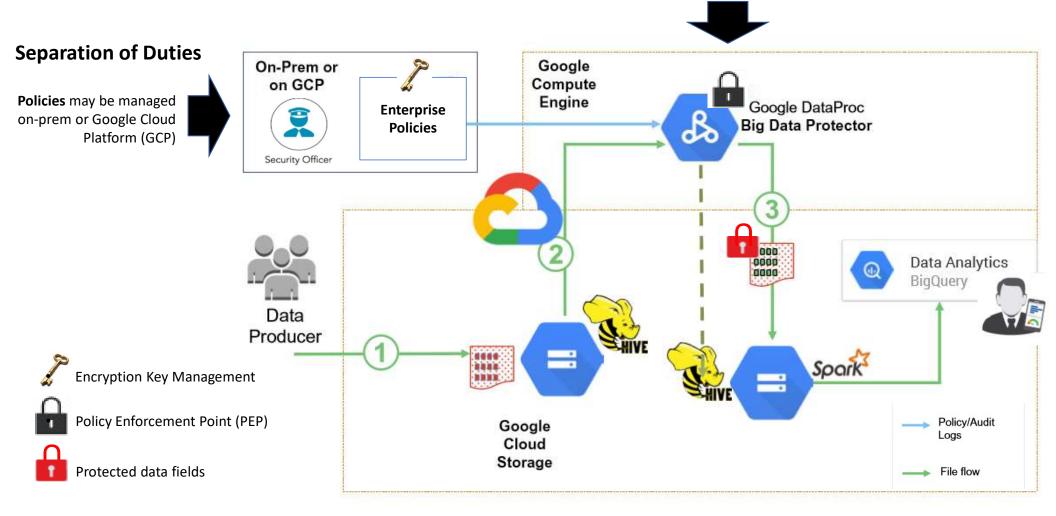




Big Data Protection with Granular Field Level Protection for Google Cloud

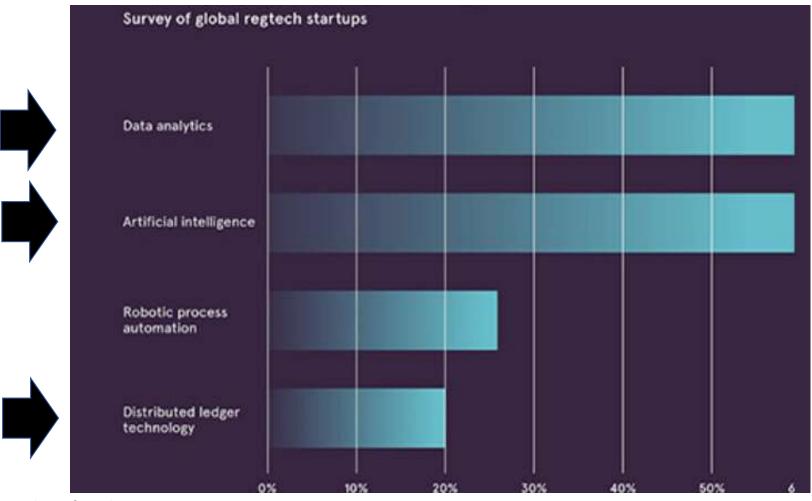
Big Data Protector tokenizes or encrypts sensitive data fields





Underlying Technology for RegTech Solutions





Source: medium, fintechnews



Shared platforms are most useful when built with digital technologies - Examples



Privacy Enhancing Technologies (PETs)

The market recently began exploring the use of PETs in KYC (e.g., FCA's July 2019 AML and Financial Crime TechSprint in London focused on applying examples of PETs to AML/KYC).²³

Homomorphic encryption (HE): Enables the processing of machine-to-machine encrypted data without the need to decrypt the data. Basically, HE allows data to remain encrypted while it is analyzed and processed.

Zero-knowledge proof (ZKP): Enables data to be verified without revealing the data itself. The technology can transform the way data is collected, used and transacted. ZKP uses the concept of a verifier and a prover. In each transaction, the prover can use the data without revealing the input or the computational process to the verifier.

Data Sharing Technologies

Digital Identity

Some shared platforms use biometrics and machine learning tools to establish, maintain and share digital identities of individuals or entities, while reducing friction for end users (e.g., India's Aadhaar).¹

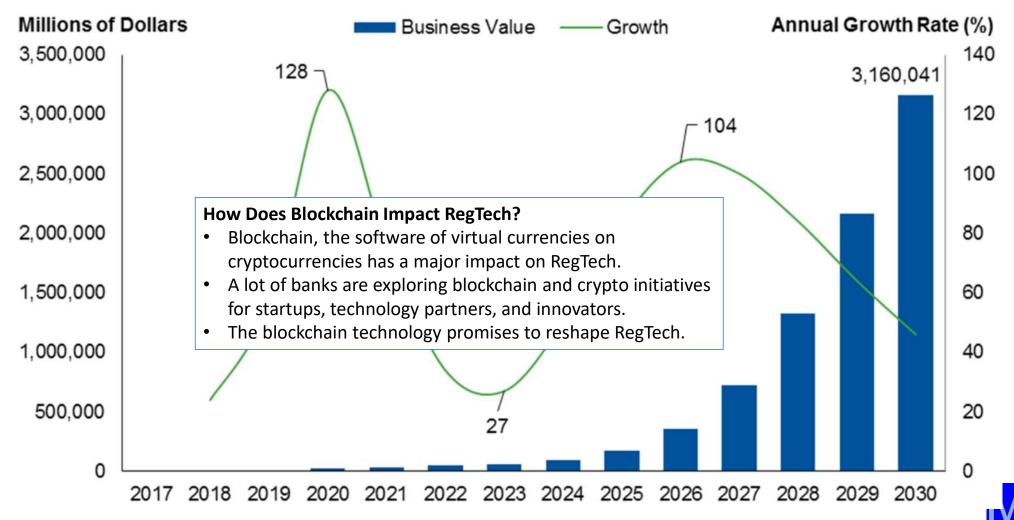
DLT, such as blockchain, underpins a secure ledger of digital events that is shared among all the parties participating in the events. Blockchain is bonded in nature, as each block can contain several transactions and has a unique proof of work attached. Together with the unique proof of work from the previous block, a chain effect is created, making it impossible to alter the information.⁴

DLT allows a high degree of data privacy and security while maintaining transparency through an audit trail of data changes. DLT uses smart contracts to help streamline roles and responsibilities in a shared platform.

Source: https://regtechassociation.org/wp-content/uploads/2019/11/Protiviti-IRTA-Urgent-Call-KYC-Optimization-Final-Report.pdf

Gartner Forecast: Blockchain Business Value, Worldwide

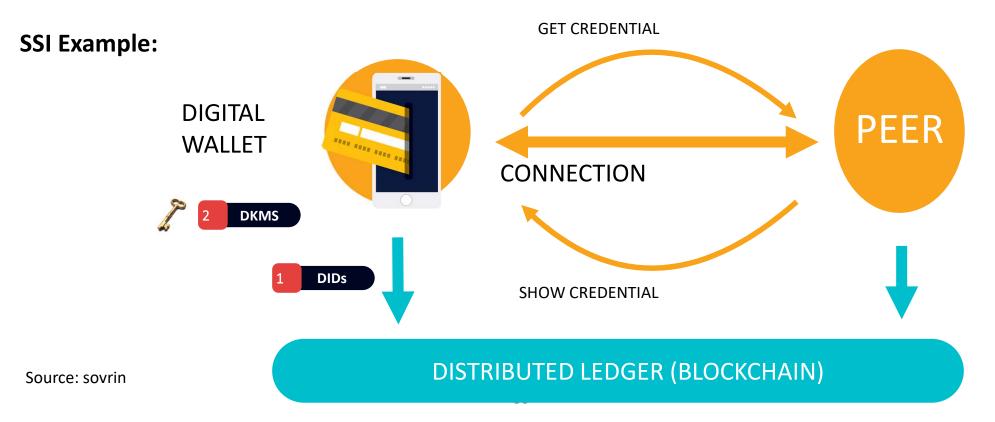




Emerging De Jure Standards for Self-Sovereign Identity (SSI)



- 1. Verifiable Credentials, (W3C)
- 2. DID Auth, (IETF)
- 3. DKMS (Decentralized Key Management System), (OASIS)
- 4. DID (Decentralized Identifier), (W3C)

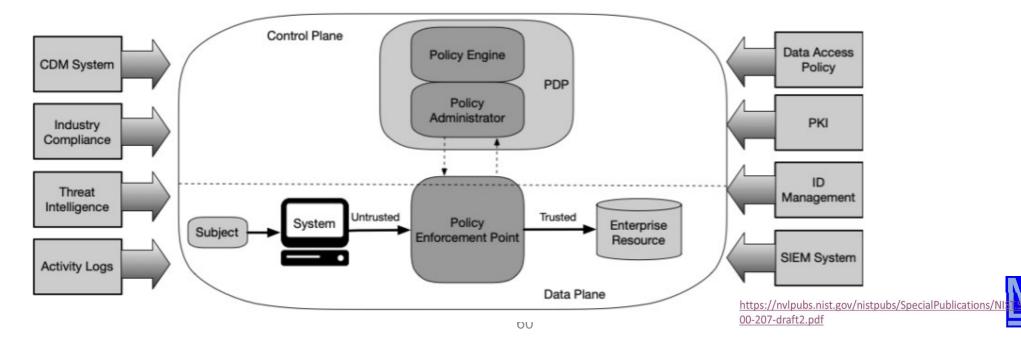




A zero trust architecture (US NIST, ANSI X9)



- 1. All data sources and computing services are considered **resources**.
- 2. All **communication is secured** regardless of network location.
- 3. Access to individual enterprise resources is granted on a **per-session** basis.
- 4. Access to resources is determined by **dynamic policy**—including the observable state of client identity, application, and other behavioral attributes.
- 5. The enterprise monitors assets to ensure that they remain in the most secure state possible.
- 6. All resource authentication and authorization are dynamic and strictly enforced before access is allowed.
- 7. The enterprise collects the current state of network infrastructure and communications.



References:



- 1. California Consumer Privacy Act, OCT 4, 2019, https://www.csoonline.com/article/3182578/california-consumer-privacy-act-whatendon-Chapter-you-need-to-know-to-be-compliant.html
- 2. CIS Controls V7.1 Mapping to NIST CSF, https://dataprivacylab.org/projects/identifiability/paper1.pdf
- 3. GDPR and Tokenizing Data, https://tdwi.org/articles/2018/06/06/biz-all-gdpr-and-tokenizing-data-3.aspx
- 4. GDPR VS CCPA, https://wirewheel.io/wp-content/uploads/2018/10/GDPR-vs-CCPA-Cheatsheet.pdf
- 5. General Data Protection Regulation, https://en.wikipedia.org/wiki/General Data Protection Regulation
- 6. IBM Framework Helps Clients Prepare for the EU's General Data Protection Regulation, https://ibmsystemsmag.com/IBM-Z/03/2018/ibm-framework-gdpr
- 7. INTERNATIONAL STANDARD ISO/IEC 20889, https://webstore.ansi.org/Standards/ISO/ISOIEC208892018?gclid=EAIaIQobChMlvI-k3sXd5gIVw56zCh0Y0QeeEAAYASAAEgLVKfD_BwE
- 8. INTERNATIONAL STANDARD ISO/IEC 27018, https://webstore.ansi.org/Standards/ISO/ ISOIEC270182019?gclid=EAIaIQobChMIleWM6MLd5gIVFKSzCh3k2AxKEAAYASAAEgKbHvD_BwE
- 9. New Enterprise Application and Data Security Challenges and Solutions https://www.brighttalk.com/webinar/new-enterprise-application-and-data-security-challenges-and-solutions/
- 10. Machine Learning and AI in a Brave New Cloud World https://www.brighttalk.com/webcast/14723/357660/machine-learning-and-ai-in-a-brave-new-cloud-world
- 11. Emerging Data Privacy and Security for Cloud https://www.brighttalk.com/webinar/emerging-data-privacy-and-security-for-cloud/
- 12. New Application and Data Protection Strategies https://www.brighttalk.com/webinar/new-application-and-data-protection-strategies-2/
- 13. The Day When 3rd Party Security Providers Disappear into Cloud https://www.brighttalk.com/webinar/the-day-when-3rd-party-security-providers-disappear-into-cloud/
- 14. Advanced PII/PI Data Discovery https://www.brighttalk.com/webinar/advanced-pii-pi-data-discovery/
- 15. Emerging Application and Data Protection for Cloud https://www.brighttalk.com/webinar/emerging-application-and-data-protection-for-cloud/
- 16. Data Security: On Premise or in the Cloud, ISSA Journal, December 2019, ulf@ulfmattsson.com
- 17. Webinars and slides, www.ulfmattsson.com

