

# Removing Danger From Data

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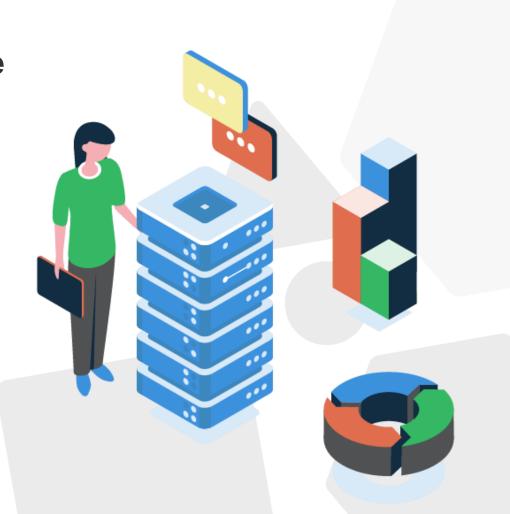


# Businesses face increasing risk around safety of data.

 Data is increasingly taking central role in many businesses.

... while at the same time

 Government regulations around safeguarding practices are rapidly emerging and are being actively enforced.





## **Rising Regulations**

### European Regulation – GDPR

Enforcement began in May 2018

Fines already levied against Google, British Airways, Marriot

### US Regulation status

No national regulation

California CCPA regulation takes effect January 2020

Microsoft will comply with CCPA nationwide

Other states following California model

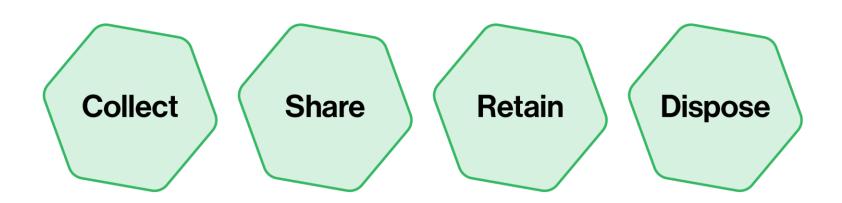


# It's not only Plls that needs safeguarding

- PII danger more obvious and includes potential legal costs name, birthdate, SSN, account number
- Non-PII data is also potentially dangerous.
   sales forecasts, product plans, KPIs



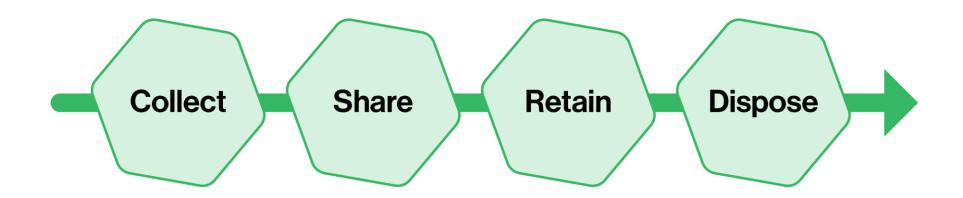
# Danger lurks in all phases of the Data Life Cycle





### Danger lurks in all phases of the Data Life Cycle

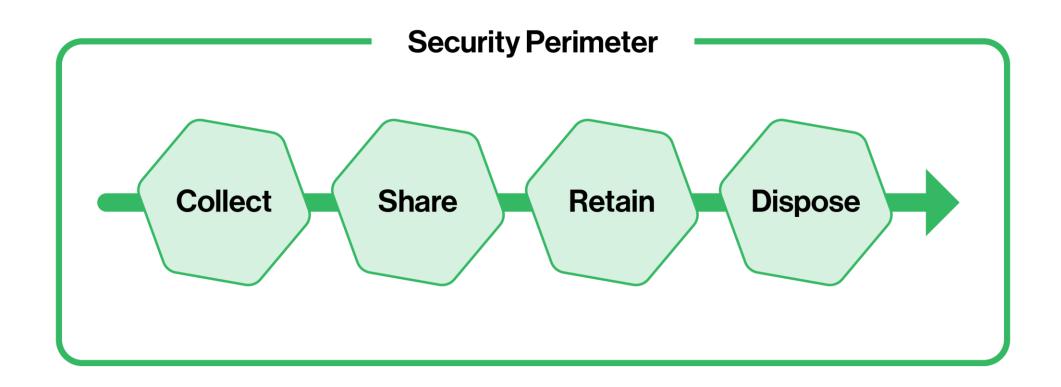
Typically there's **processing** required to move data between stages



This processing can include actions to remove danger

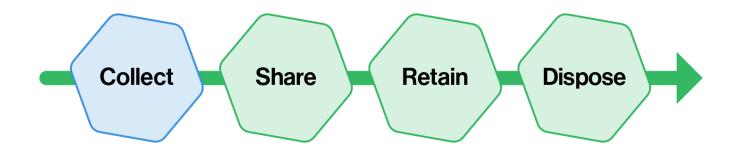


# Controlling danger from within





### **Collection Phase**



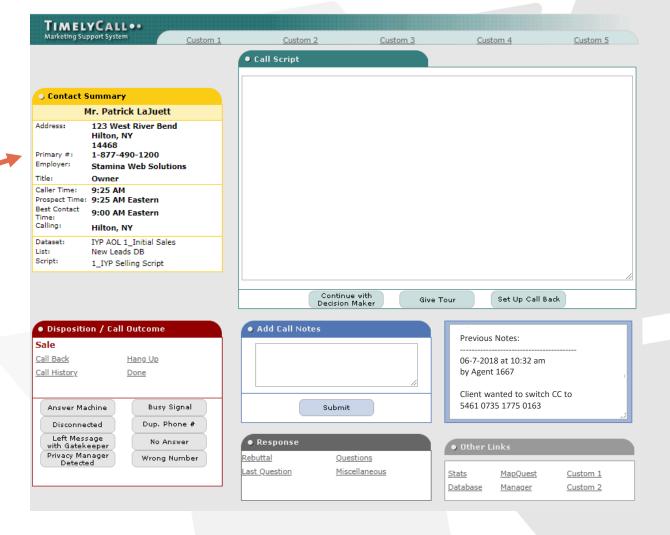
- Danger originates in collection phase
- Not knowing if PII data has been introduced into system



### **Collection Phase:**

# Look for dangerous data everywhere



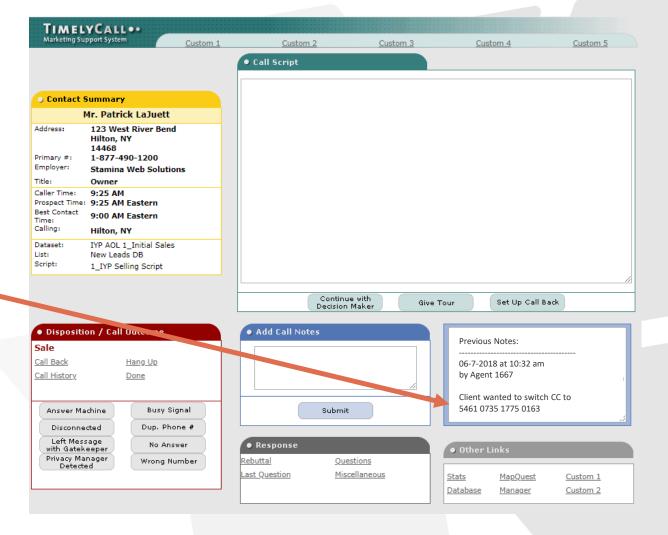




### **Collection Phase:**

# Look for dangerous data everywhere

unexpected





### **Collection Phase**

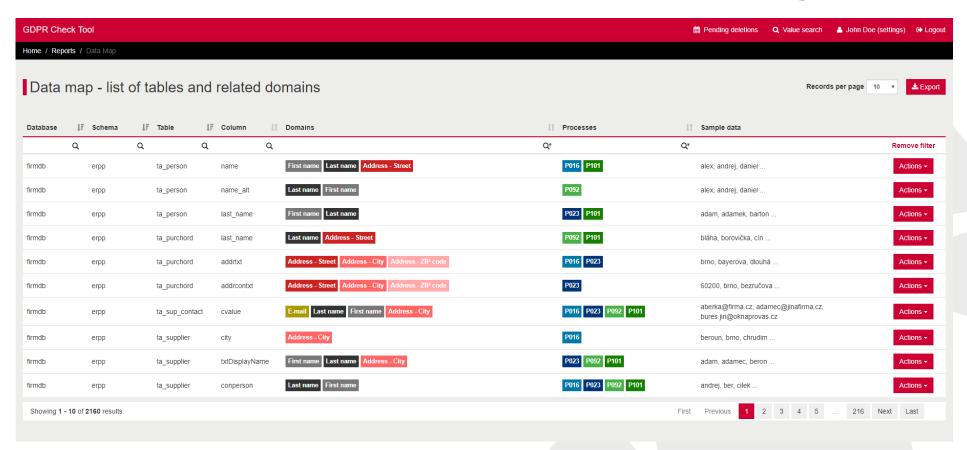
#### Remedies:

- Data minimization
- Choose ETL tools that provide input classification
- Consider Software to catalog and tag collected data



### **Collection Phase:**

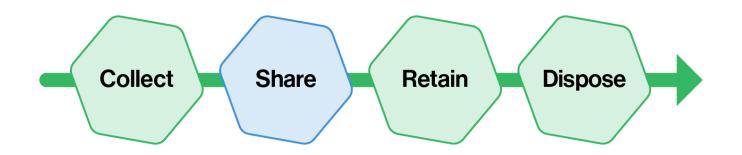
Data Classification tools can help track incoming sensitive data



Data Classification tags can be stored and accessed in a Data Catalog



### **Share phase:**



- Sharing usually requires transforming data assets so they can provide desired business value.
   searching, mapping, sorting, merging categorizing, analyzing, reshaping.
- Primary danger is oversharing
   Sharing too much data
   Sharing to an unnecessarily large audience
   Accidental sharing



### **Share phase:**

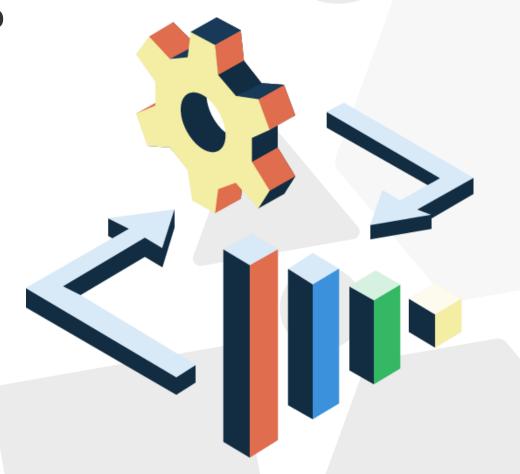
Processing techniques available to deliver business value without unnecessarily sharing sensitive data.

- Anonymization
- Privacy preserving analytics



# **Anonymization**

Process of obfuscating data, so they keep some of their original attributes but not to extent they could be used to infer relation to real people or entities.





# **Anonymization – Shuffle**

First Name	Surname	Age	First Nam	e Surname	Age
Alice	Smith 🛶	42	Alice	→ Doe	42
Bob	Johnson	21	Bob	Jackson	21
Dave	Doe 🚽	74	Dave	Chang	74
Eve	Jackson-	44	Eve	Smith	44
Grace	Chang -	32	Grace	Johnson	32



# **Anonymization – Jitter**

Birth Date	Gender	<b>Education Level</b>		Birth Date	Gender	<b>Education Level</b>	
7-Jul-82	Male	High School	+3 days	10-Jul-82	Male	High School	
18-May-85	Male	High School	-1 day	17-May-85	Male	High School	
5-Mar-87	Female	Bachelor	+3 days	8-Mar-87	Female	Bachelor	
17-Jun-97	Male	Associate	-1 day	16-Jun-97	Male	Associate	
27-Feb-82	Male	Graduate	-3 days	24-Feb-82	Male	Graduate	
9-Nov-58	Male	Graduate	+1 day	10-Nov-58	Male	Graduate	



## **Anonymization – Masking**

Naively masked 4024 XXX

4024 XXXX XXXX XXXX

Intelligently masked

4024 0071 4314 0399

Keeping Issuer code VISA Credit card Issued by Bank of America Randomized Account Number

Valid Luhn checksum

Preserves card types, issuers, preserves validity



# Anonymization – Aggregation/Generalization

Birth Date	Gender	Education Level	Age	Gender	Education Level
7-Jul-82	Male	High School	30's	Male	High School
18-May-85	Male	High School	30's	Male	High School
5-Mar-87	Female	Bachelor	30's	Female	Bachelor
17-Jun-97	Male	Associate	20's	Male	Associate
27-Feb-82	Male	Graduate	30's	Male	Graduate
9-Nov-58	Male	Graduate	60's	Male	Graduate



## **Privacy Preserving Analytics**

- Methods developed at Linked-In for data mining on very large (web-scale) result sets
- Provide accurate analytics without inadvertently identifying individual users.

Inject small amounts of random noise to query/search

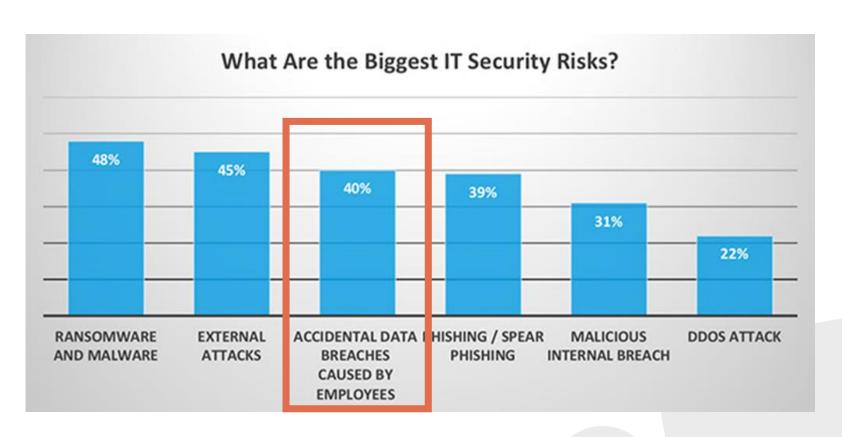
Offset noise with more processing to maintain data consistency

See Privacy-Preserving Analytics and Reporting at LinkedIn



### **Share Phase**

Employees simply doing their job can be dangerous.



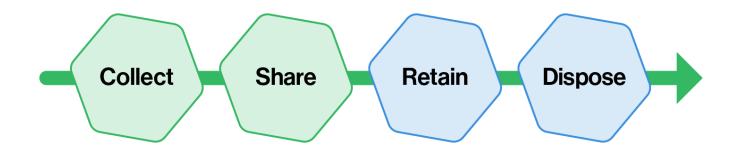


### **Share Phase:**

- 5 most common technologies that lead to accidental data breaches by employees:
- 1. External email services (Gmail, Yahoo!, etc.) (51 percent)
- 2. Corporate email (46 percent)
- 3. File sharing services (FTP sites, etc.) (40 percent)
- 4. Collaboration tools (Slack, Dropbox, etc.) (38 percent)
- 5. SMS / messaging apps (G-Chat, WhatsApp, etc.) (35 percent)



### **Retain and Dispose Phases**

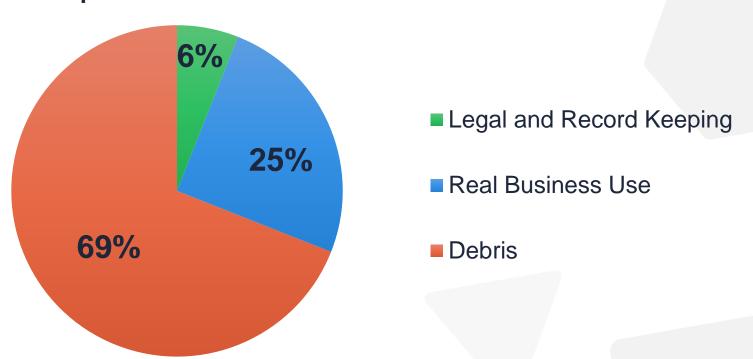


- Minimize time data is kept
- A Data Catalog can also house retention and disposal status



### Most enterprise data is retained indiscriminately

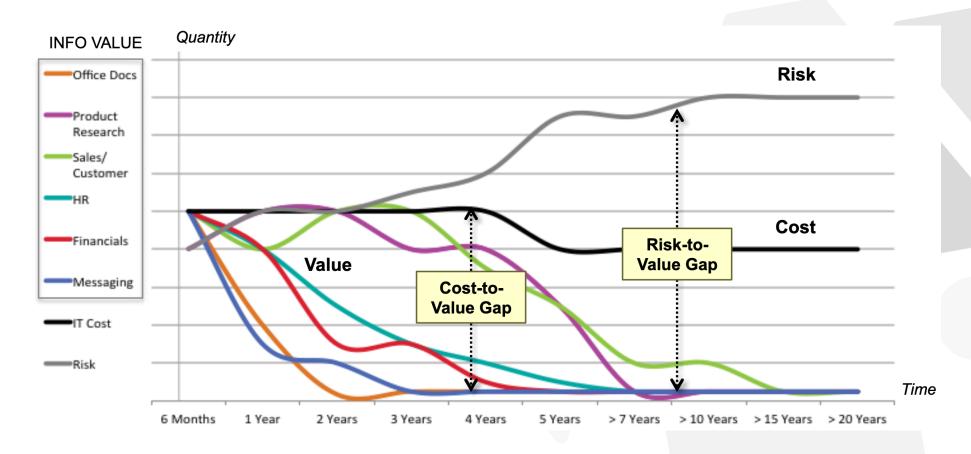
#### **Enterprise Data Retention**



https://cedar.princeton.edu/sites/cedar/files/media/information\_lifecycle\_governance.pdf



# Information Value Declines Over Time, Costs and Risks of retention do not.



https://cedar.princeton.edu/sites/cedar/files/media/information\_lifecycle\_governance.pdf



### **Retain and Dispose Phases**

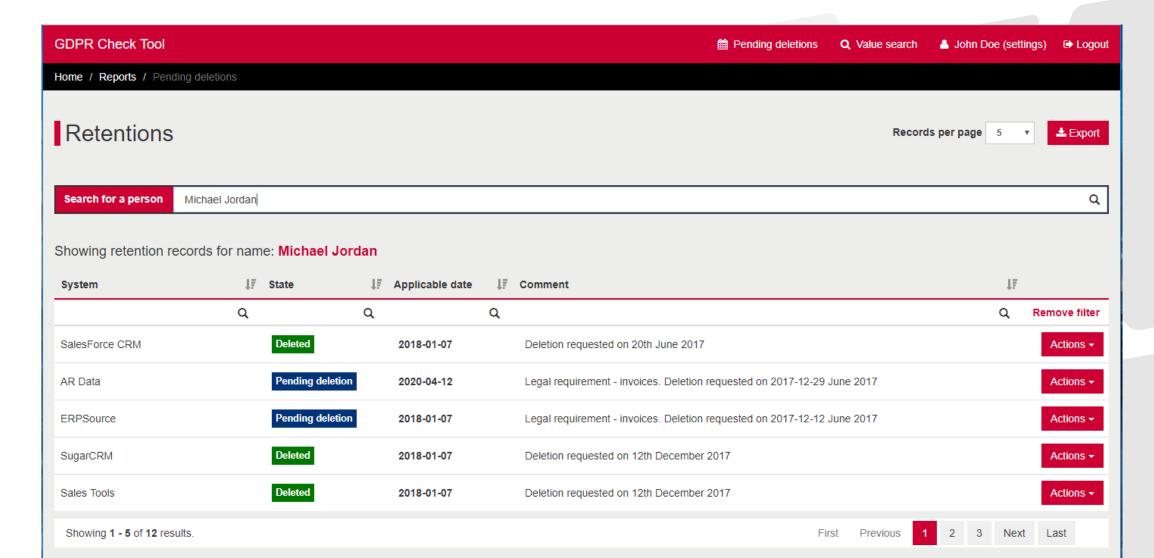
In 2019 CapitalOne Breach exposed sensitive data that included <u>rejected</u> credit card applications from as far back as <u>2005</u>.

Remediation expected to cost \$200-300 million without any explicit regulatory fines.





### **Retain and Dispose Phases**





# **Summary**

Emerging Regulations are increasing danger in data

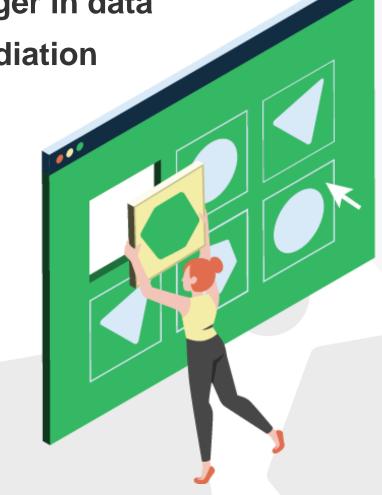
Removing Danger = Identification + Remediation

Realizing the dangers exist Software tools exist to help

Develop Risk Intelligence

Size of risk vs cost of remedy

Implement reasonable precautions





# Thank you

#### **About CloverDX Data Integration Platform**

CloverDX is a data integration platform for designing, automating and operating data jobs at scale. We've engineered CloverDX to solve complex data movement and transformation scenarios with a combination of visual IDE for data jobs, flexibility of coding and extensible automation and orchestration features.

www.cloverdx.com



### References

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