

## DS-2002: Data Science Systems

A Survey of Data Management Systems

Prof. Jon Tupitza – Fall 2022



## Class Mechanics

A Survey of Data Management Systems



#### Who is... Teaching this Course?



- Jon Tupitza @ Microsoft: Sr. Cloud Solution Architect for Data, AI & Analytics
  - Lead Enterprise Data Science Program (EDSP) for Retail & Consumer Goods OU
  - Mentor & Courseware developer for EDSP Mentoring Program Certification Proctor
- MS Data Analytics 2018
- More Industry Certifications than I can remember see LinkedIn (MCP since 1999)
- Microsoft Guy: 25+ yrs. Focusing on Microsoft technologies
  - Software Engineering & Architecture
  - Data Platform Architect Data Engineer Business Intelligence Data Science
- PubSec/National Security SME: 22 yrs. Supporting DoD (Army, DARPA), IC (NGA, FBI)
- Past @ Buchanan-Edwards: CTO/Solution Architect Practice Lead: Predictive Analytics
- Past @ Microsoft Consulting: Sr. Consultant specializing in SQL Server & Business Intelligence
- Past @ Avanade: Principal Solution Developer (.NET)
- Past @ Microsoft Partners: Data Scientist, Data Warehouse Architect, Software Architect

#### Who am I?



- Love to Teach: Born to be a Professor... no kidding it's been my life's goal
  - Wanted to be interesting (and financially secure) so I've done lots of other professional stuff first
- Philadelphia area (Valley Forge) native... *Where they throw snowballs @ Santa*
- Recent Transplant from DC Metro area Alexandria, VA (*Mt. Vernon*)
- 2 Kids and 1 Wife... *of 32 yrs*.
  - Wife (Karen): CFP, MS Financial Planning VP of Wealth Mgmt. & Trust... *Plans my financial future*
  - Daughter (Jill): UVA Alum 2016 PhD Candidate @ LSU (Oceanography & Coastal Sciences)
  - Son (Jake): VaTech Alum 2019 Eagle Scout, Baseball, Musician Works in the Clean Energy sector
- Transitioned from an Earlier Career in Advertising... *Art Director & Agency Computer Wiz (Mac)*
- BFA Degree (Fine Art)... from a long line of artists and craftsmen
- 41+ yrs. studying & 28 yrs. teaching Okinawan & Japanese Martial Arts
  - 9th Dan Hanshi Shoshinkan-ha Shorinken Karate/Kobudo & Itto-Ryu Kenjutsu
  - 5<sup>th</sup> Dan Kyoshi Classical Judo

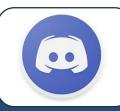
#### Class Mechanics: Course Content





#### **Lectures:**

- In-Person
- First Class of the Week (Monday)



#### **Discussion:**

Mondays as a part of Lecture Day



#### **Readings / Videos:**

- Articles and Posts
- Supporting tools and videos
- ~Hours per Week Prep for Mondays Class



#### Labs:

- Hands-On, Virtual & Wednesday.
- I may Provide a Zoom invite.

#### Class Mechanics: Course Content



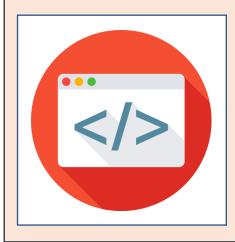


#### Labs

- Frequent Hands-On Labs
- AWS, Azure, OCI, GCP (Amazon, Microsoft, Oracle, Google)
- 1 Per Week (More Available)
- Begin on Wednesdays and Have Completed by Friday

## Class Mechanics: Grading





#### Data Projects

- Concrete Examples of Implementation (Two in total)
- Released long before they're due
- Detailed Instructions will be in Collab
- No Time Limits!

#### Class Mechanics: Grading



Component	Weight	Frequency
Student Survey		First Week
Lectures / Readings / Material		Weekly
Engaged Discussion	5%	Weekly
Quiz	15%	3 or 4
Labs	30%	Weekly
Data Projects	50%	2 projects (25% each)

#### Class Mechanics: Things You Need to Know



How this course will be conducted: Class Location, Course Materials, Communications

- We will be IN-PERSON on Tuesdays...
- Wednesdays we may be on Zoom for Hands-on Labs
  - Sometimes we'll do them in groups
  - ...And sometimes you'll do them alone
- The schedule is prone to updates: I'll be sure to keep everyone informed!
- Everything will be updated on GitHub:
  <a href="https://github.com/JTupitza-UVA/DS-2002-002">https://github.com/JTupitza-UVA/DS-2002-002</a>
- Quizzes, Projects and Grades will be released on Collab
- You can DM me on Discord with Questions or to Set-up Office Hours

## Learning Objectives

**Develop Robust Facility for Handing Data** 

Acquire a Strong Understanding of SQL and NoSQL Databases

**Understand Systems Options in the Public Cloud** 

Gain Fluency in Tools, Processes, and Services ...and How to Choose Among Them

**Data Retrieval** 

**Data Shipping** 

**Data Schemas** 

**Data Ingestion** 

**Data Processing** 

**Data Normalization** 

**Data Analysis** 

# Overview of the Semester Topics

A Survey of Data Management Systems



## In the Beginning... There Was Mainframe





- Data stored in ISAM files (indexed sequential access method)
  - Flat files having fixed-length fields
- Centralized computing and storage resources
  - "Dumb" terminal clients
- Only affordable to large, wealthy enterprises:
  - Governments
  - Large Banks

#### Then Came... Client/Server Networks





- Data stored in Relational Database Management Systems (RDBMS):
  - Oracle
  - SQL Server
- Enterprise Data Centers:
  - Company owned & managed
  - Based on commodity hardware
- Some computation and storage resources shared by clients

#### Which Gave Us... SMP Servers



Relational Database Management running on Symmetric Multi-Processing Servers



- Dedicated Database Servers:
  - RDBMS (Relational Database Management Server)
  - Oracle, SQL Server, DB2, etc.
- OLTP (Online Transaction Processing)
  - Characterized by a large volume of transactions each of which affect a small number of rows
  - Online Sales, Bank Deposits & Transfers
  - Highly Normalized Database Schema
- OLAP (Online Analytical Processing)
  - Characterized by a small volume of read transactions each of which affect a large number of rows
  - Periodic Post-hoc Analysis (What Happened?)
  - De-Normalized Multi-Dimensional Data Warehouse Schema

## But Then... An Explosion of Data – "BIG Data"



A Rapid, Exponential Proliferation of New Devices: The Internet of Things (IoT)



• Explosion in social media, mobile apps, digital sensors, RFID, GPS, and more have caused exponential data growth.



• Sources like social networking and sensor signals create data at a tremendous rate; making it a challenge to capture, store, and analyze that data in a timely or economical manner.



• Traditionally BI has sourced structured data, but now insight must be extracted from unstructured or poly-schematic data like large text blobs, digital media, sensor data, etc.

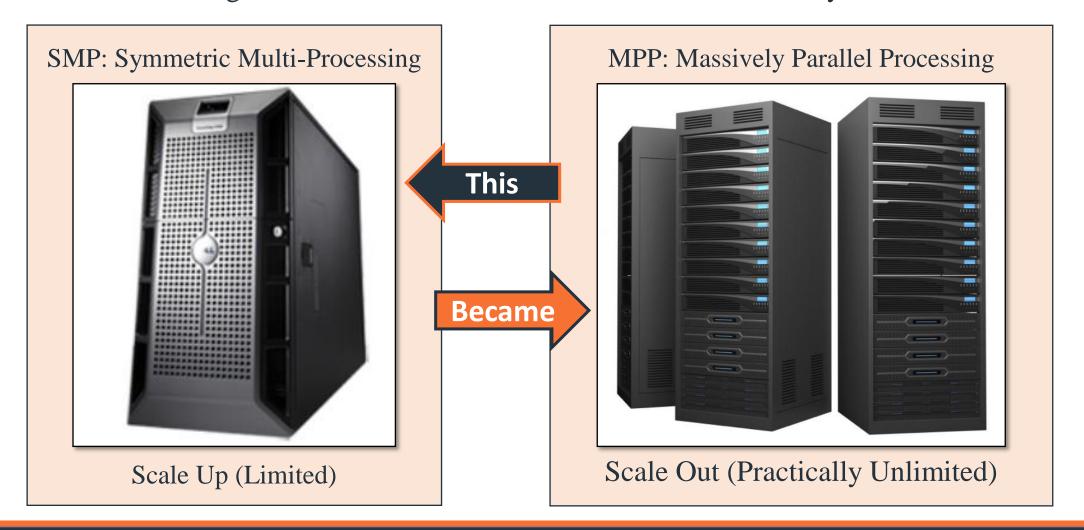


• The anonymity of the WWW, incredible sources like social networking and duplicate systems bring into question the authenticity of the information being generated and collected.

## But... How Can We Manage All That Data: Scalability



The Advent of Big Data Drove a Need to Increase Scale to the Petabyte Level: \$\$\$\$\$\$



## And Today... Cloud-hosted Databases & Services



Re-centralization of Servers • Based on Commodity HW • Resources Shared by Client

**Essential Characteristics** 

**On-Demand Self-Service** 

**Broad Network Access** 

**Resource Pooling** 

**Rapid Elasticity** 

**Measured Service** 

**Service Models** 

Infrastructure as a Service (IaaS)

Platform as a Service (PaaS)

Software as a Service (SaaS)

**Deployment Models** 

**Public Cloud** 

**Private Cloud** 

**Community Cloud** 

**Hybrid Cloud** 

#### Service Models: On-Premises vs Cloud-Hosted



Managed by Owner	<b>Managed by Service Provider</b>
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On Premises	IaaS	PaaS	SaaS
Applications	Applications	Applications	Applications
Data	Data	Data	Data
Runtime	Runtime	Runtime	Runtime
Middleware	Middleware	Middleware	Middleware
OS	OS	OS	OS
Virtualization	Virtualization	Virtualization	Virtualization
Servers	Servers	Servers	Servers
Storage	Storage	Storage	Storage
Networking	Networking	Networking	Networking

#### Data Management... In the Cloud

Economies of Scale Afforded by the Cloud Have Enabled Massive Storage & Compute

- File-based Storage: Blobs, Queues, Tables, Data Lakes
- Relational Database Management Systems:
  - Online Transaction Processing (OLTP) systems:
    - IaaS: Cloud-hosted VMs Running RDBMS Software & Databases
    - PaaS: Managed RDBMS Instance
    - SaaS: (Database as a Service) Single Databases or Multi-Database Pools
  - Online Analytical Processing (OLAP) systems:
    - Massively Parallel Processing (MPP) Data Warehouse
- NoSQL Database Management Systems:
  - Semi-Structured or Poly-Schematic data
  - Massively Parallel Processing
    - Data partitioned and replicated across many server machines (nodes)
    - Data typically stored in file-based format (e.g., JavaScript Object Notation (JSON)

#### Paradigms: Data Storage and Retrieval



Schema on Write versus Schema on Read



Schema on Read: Applies schema only when read, data stored in its original format





#### NoSQL: Not Only SQL



## Key-value stores

- The simplest NoSQL database; based on "dictionaries" or "maps".
- Items are stored in associative arrays; pairing a name (or "key"), with a value.
- Riak, FoundationDB, and Redis

## **Column** stores

- Combines a key, value and timestamp for each item.
- Optimized for large datasets by storing columns of data together, rather than in rows.
- Cassandra, BigTable and HBase

## **Document** databases

- Pairs keys with complex data structures (documents) using XML, JSON or BSON.
- Documents may contain key-value pairs, key-array pairs, and nested documents.
- MongoDB, MarkLogic, and Apache CouchDB

#### **Graph stores**

- Stores interrelated networks of data such as social connections, or network topologies.
- Optimized for interconnected data elements with an undetermined number of relations.
- AllegroGraph, Neo4J and HyperGraphDB.

#### SQL versus NoSQL Databases



	SQL Databases	NoSQL Databases
Types	One: Relational Database	Many: Key-value, document, column, and graph databases
Data Storage	Records are stored as rows in tables that represent entities. Entity relationships are modeled by joining tables.	Records may be stored as "documents" using XML or JSON. Entity relationships are modeled by nesting them hierarchically.
Schemas	<u>Static:</u> Structure & data types are fixed at design time. Adding new elements requires schema design changes.	<u>Dynamic:</u> New elements can be added at runtime. <u>Poly-schematic:</u> Dissimilar data can be stored together as necessary.
Scaling	<u>Vertically</u> : A single server must be made increasingly powerful to cope with increasing demand.	<u>Horizontally</u> : New commodity servers and storage are added to an array. Data is automatically distributed across all servers.
Transactions	Full transactional consistency (ACID)	Single element (document) only.
Manipulation	Using platform-specific languages (SQL)	Using OSS API's, low-level lang., JavaScript
Consistency	Supports strong consistency	Per-product: Most are eventually consistent.

## Onward to Data Science

A Brief Overview of How Data Systems are Involved

#### The Data Science: Hierarchy of Needs



**LEARN / OPTIMIZE** 

AGGREGATE / LABEL

**EXPLORE / TRANSFORM** 

**MOVE / STORE** 

**COLLECT** 

AI, Deep Learning

A/B Testing, Experimentation, Simple ML Algorithms

Analytics, Metrics, Segments, Aggregates, Features, Training Data

Cleaning, Anomaly Detection, Prep

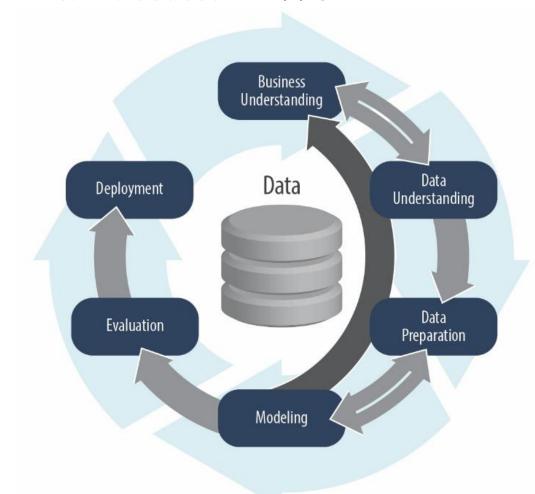
Reliable Data Flow, Infrastructure, Pipelines, ETL, Structured & Unstructured Data Storage

Instrumentation, Logging, Sensors, External Data, User-Generated Content

## CRISP-DM: Cross-Industry Standard Process-Data Mining



First Introduced in 1996!



Business Understanding

- Identify the Problem Domain
- Identify the Solution Scenario

Data Understanding

- Load and Explore Data
- Identify Influential Features

**Data Preparation** 

- Remove Duplicates & Nulls
- Impute Missing Values
- Select & Engineer Features

Modeling

- Train Models Using a Variety of Algorithms
- Tune Hyper-parameters

**Evaluation** 

- Test Models' Performance & Predictive Power
- Cross-Validate to Appraise Goodness-of-Fit
- Select Most Effective Model for Deployment

Deployment

- Publish Models On-premises or in the Cloud
- Consume Models Visually & Programmatically

#### Data Engineering... for Data Science



Frequently, Data Must Be Moved from Sources to a Database and/or Data Lake



Extract

■ This is the step where sensors wait for upstream data sources to land. Once available, we transport the data from their source locations to further transformations.



Transform

■ The heart of any ETL job: apply business logic, perform actions such as filtering, grouping, and aggregation to translate raw data into analysis-ready datasets.

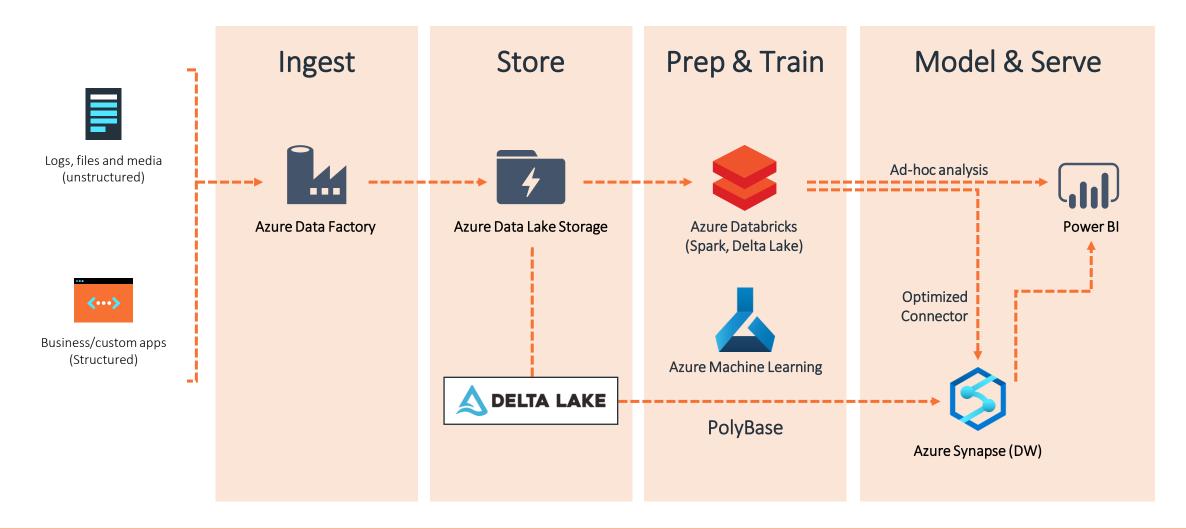


Load

Load the processed data and transport to a final destination.
 Can now be consumed directly by end-users or treated as yet another upstream dependency.

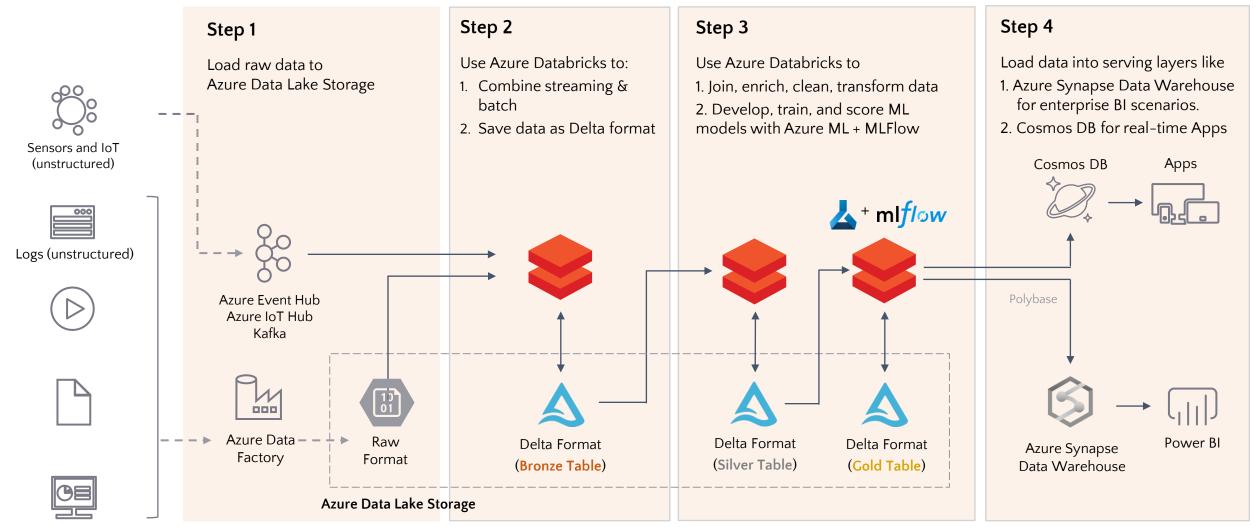
#### Data Engineering... for Data Science





#### Design Pattern: Modern Data Warehousing





Q & A

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