

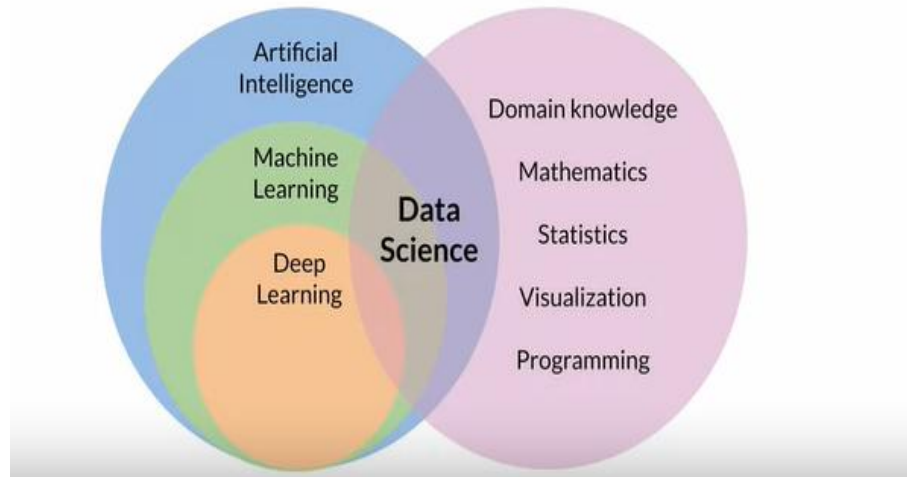
# Practical Data Science with AWS Cloud

*(Hand-written notes and Course Gist)*

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**Energy Data Scientist, Petroleum From Scratch**

AI, ML, DL, data science...?



**DAY 1**



# 1. Basics about AWS cloud : Scaling Up & Scaling Out abilities

PRACTICAL DATA SCIENCE ON AWS - CLOUD

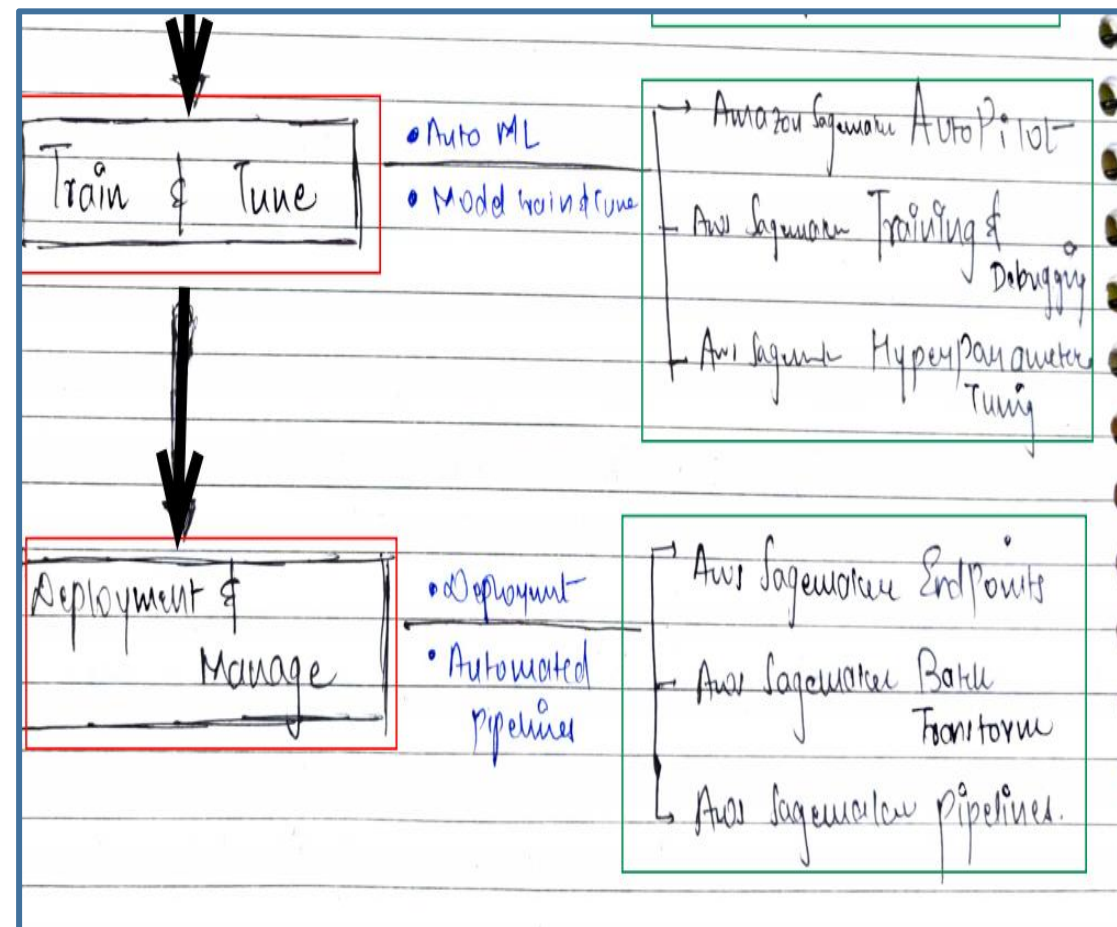
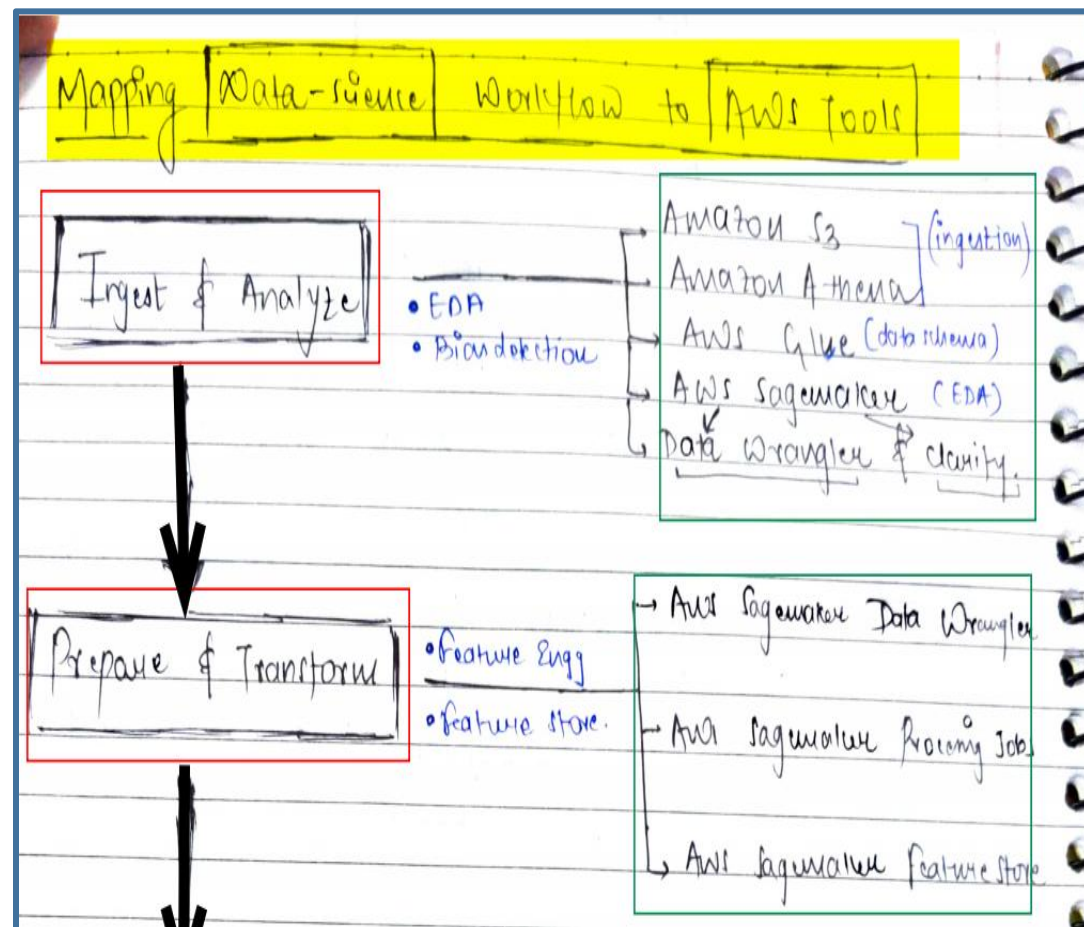
- Local Data Science → limited by Hardware & Memory  
↓  
[ Data Science on cloud → Agility | Security | scale up | scale out ]
- Scaling up → If Model training takes too long → means u've exhausted your CPU limit on the current compute instance, so you can  
↓  
↑ the size of compute instance (single CPU)  
↳ pick a compute instance with higher CPU resources.  
↳ pick up a GPU based compute instance
- Scaling out → Move from single CPU to PARALLEL computing CPU instances.  
↓  
(Parallel X CPUs)

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"Scaling Up & Scaling Out Are possible within seconds in cloud"

- Once model training is finished, the instance is killed.  
↓  
No extra payment. { Pay As you Use }
- Cloud has (AWS has) its own ML Toolkit that saves time & makes it Quick & easy.

## 2. Mapping Data Science workflow with AWS Tools





### 3. Mapping Data Science workflow with AWS Tools

#### Typical data ingestion & Query / Explore Workflow.

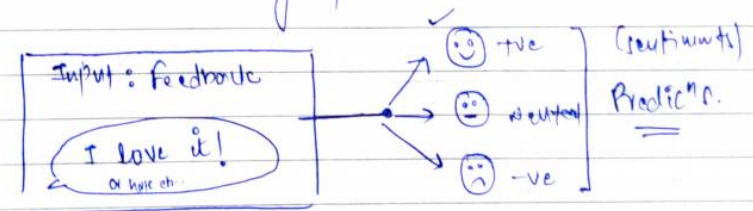
- ① Ingest : Ingest the data with Aws S3  
↓
- ② CATALOG : Catalog the data into desired schema by using Aws Glue  
↓
- ③ EXPLORE : Explore & run SQL queries on data using Aws Athena

Use-Case Description → Multi Class Classification for sentiment Analysis of Product reviews.

- You work at an e-commerce company like Amazon / Flipkart.
- Your customers leave feedbacks via all online channels.

↓  
Aim is to analyze these feedbacks quickly & alert if there are any product issues.

Example



Observation 1 → The feedbacks generated/reviewed can be millions every minute. So, we might need a data (ingestion) Repository that's

ELASTIC enough to expand as per data size & format.

## 4. Example Use case

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Aws Glue

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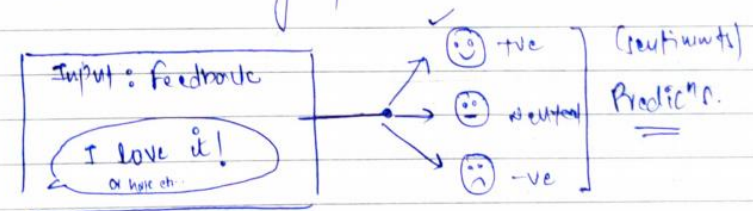
↳ Aws Athena

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# 5. Data Lake (AWS S3), Data Wrangler, AWS Athena & AWS Glue

Data LAKE : data is ingested into data lakes.

↳ centralized & secure repository

↳ store / discover / share data @ ANY SCALE

- Structured data (CSV...)
- Semi-structured data (XML, JSON...)
- Unstructured (Image, video...)
- Streaming data (LIVE).

→ • Need to be Governed ie managed to be used by further teams.

↳ Data lakes on AWS : S3

↳ Stores data chunks as objects → each object with unique identifier & metadata for easy extraction.

\* AWS Data Wrangler (pip install awswrangler)

- is a python library
- converts pandas + AWS power.

→ import awswrangler as wr  
→ import pandas as pd  
→ df = wr.s3.read\_csv(path = 'f/f')

\* AWS Glue Data Catalog

→ Catalogue is the metadata of your data (path, schema etc)  
↓  
info that you need to register before importing



# 5. Data Lake (AWS S3) , Data Wrangler, AWS Athena & AWS Glue

## \* AWS Glue Data Wrangler Catalogue

→ Catalogue is the metadata of your data (path, schema etc)  
↓  
info that you need for register before importing

DATE	
PAGE NO.	
YOUVA	
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### Catalog

→ import awswrangler as wr

→ wr.catalog.create\_database(name = ...)

→ wr.catalog.create\_mv\_table( ... )

name = ~  
database = ~  
classification = mv  
location = s3://< > / ...

## \* SQL Queries using AWS Athena → (SQL Queries in Python<sup>2w</sup>)

→ import awswrangler as wr

→ wr.athena.create\_athena\_bucket()

- df = wr.athena.read\_sql\_query(sql = 'select \* ...',  
database = ' ... ')

↓  
This queried df is first stored by Athena into S3  
↓  
then stored in variable (df).