Machine learning and data pyramid

MACHINE LEARNING FOR BUSINESS



Karolis Urbonas

Head of Machine Learning & Science, Amazon

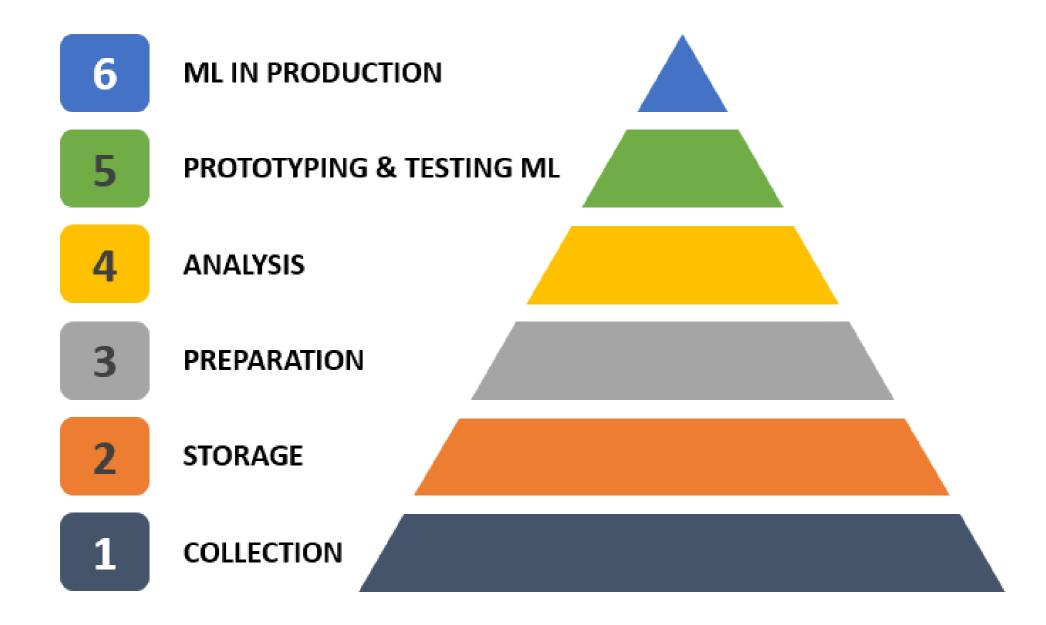


Machine Learning applications

ML is applying statistical or computer science methods on data to:

- 1. Draw causal insights
 - "What is causing our customers to cancel their subscription to our services?"*
- 2. Predict future events
 - "Which customers are likely to cancel their subscription next month?"*
- 3. Understand patterns in data
 - "Are there groups of customers who are similar and use our services in a similar way?"*

Data hierarchy of needs



Collection

1 COLLECTION

Extract data from source systems

Storage

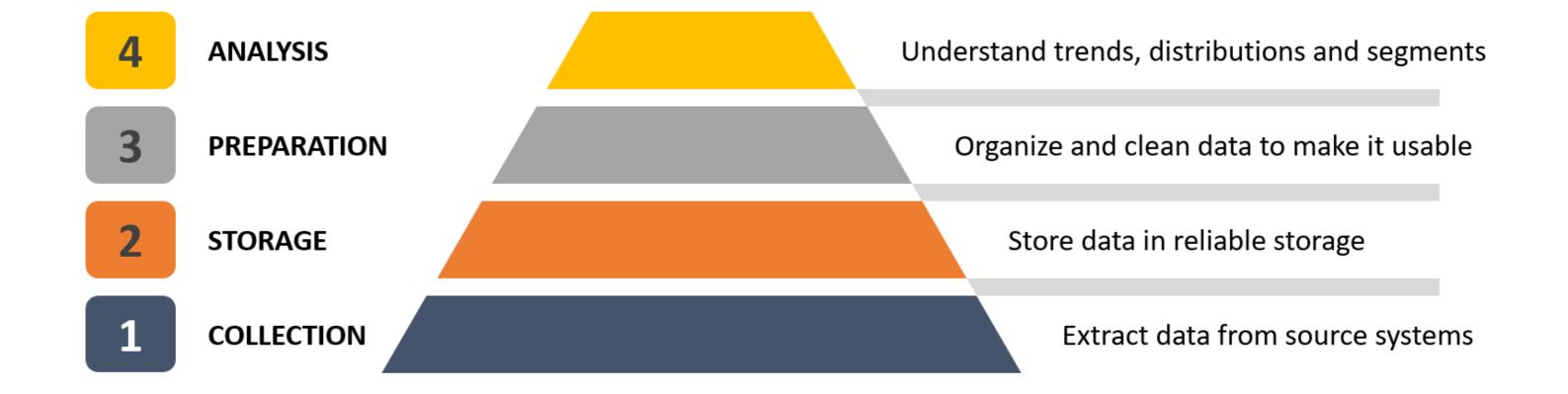


Preparation



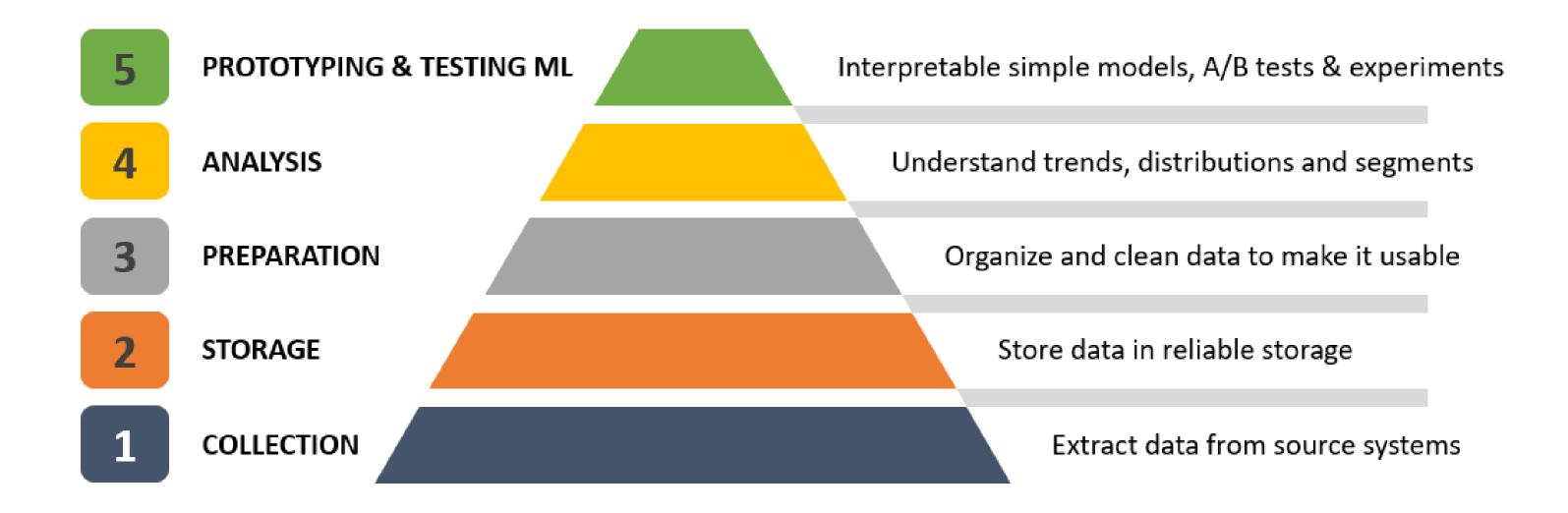


Analysis



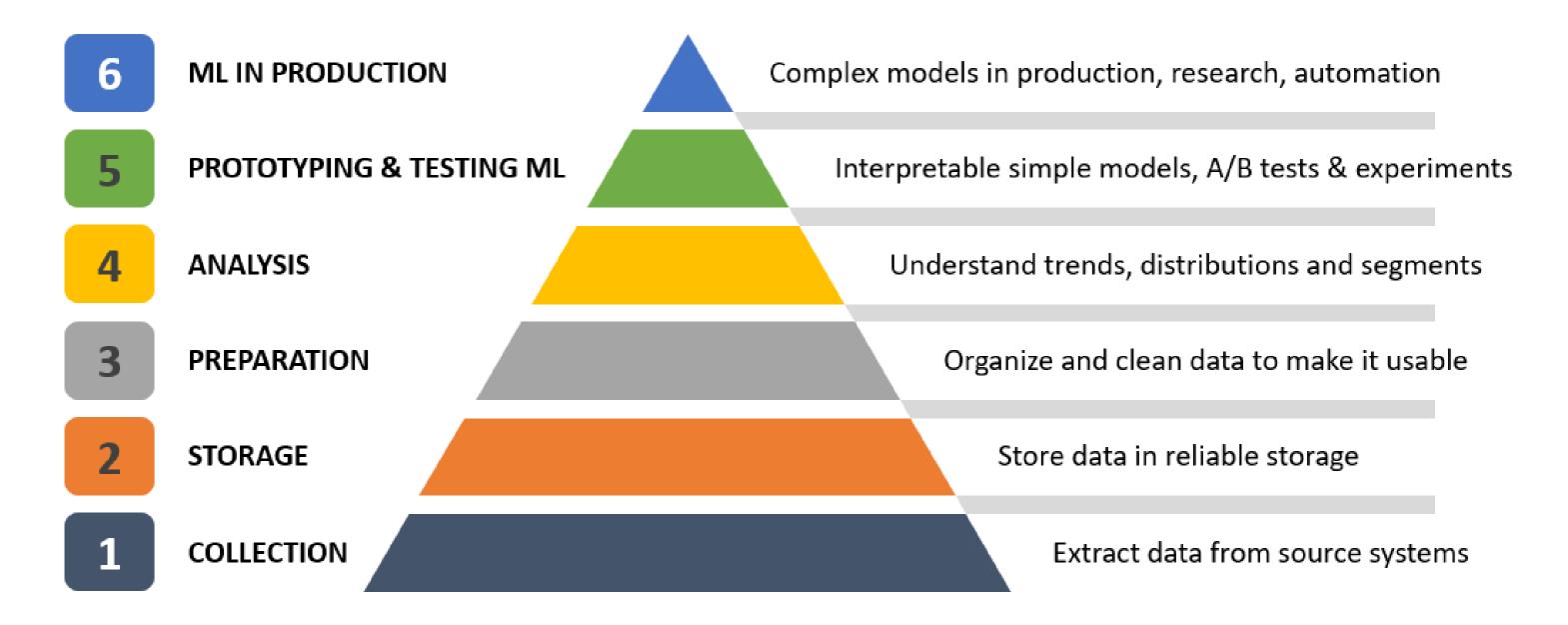


Model prototyping and testing

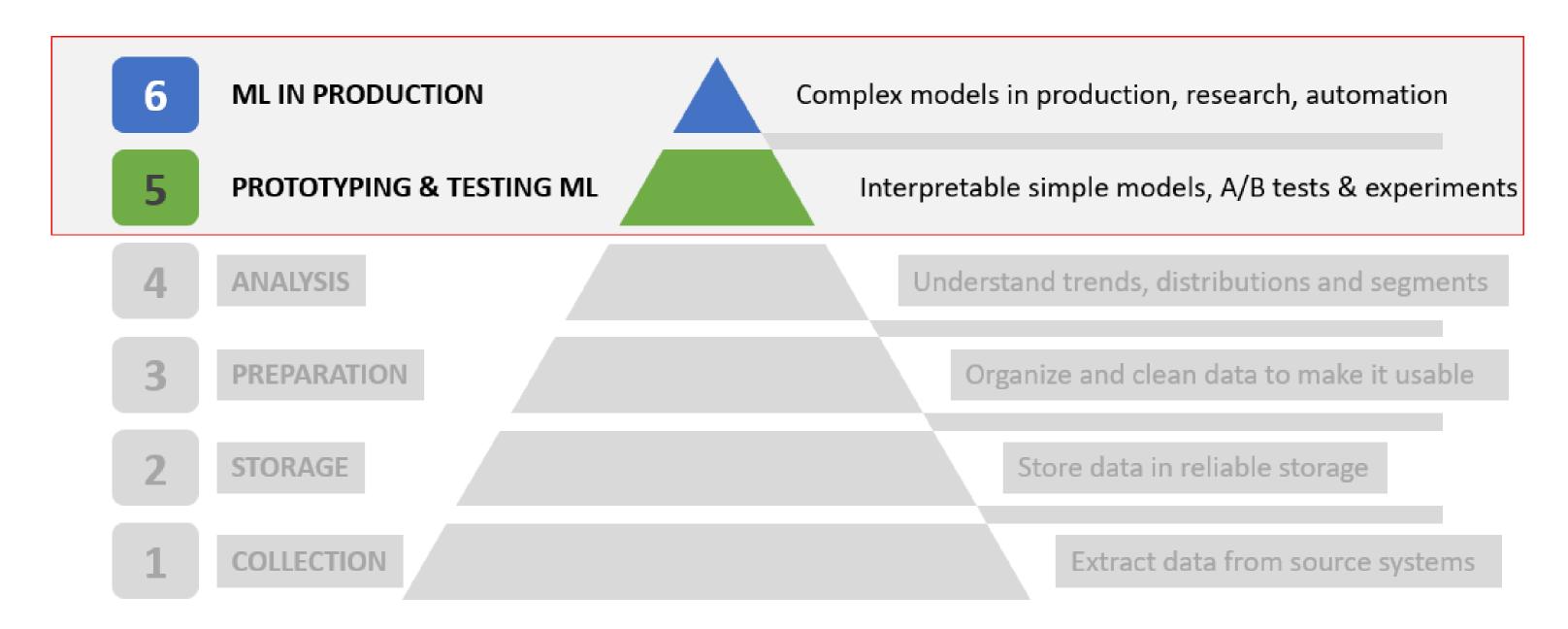




ML in production



Focus



Let's practice!

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Machine learning principles

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Machine learning types

Machine learning - applying statistical or computer science methods on data to:

- 1. Draw causal insights
 - "What is causing our customers to cancel their subscription to our services?"
- 2. Predict future events
 - "Which customers are likely to cancel their subscription next month?"
- 3. Understand patterns in data
 - "Are there groups of customers who are similar and use our services in a similar way?"

Supervised vs. unsupervised ML

Machine learning - applying statistical or computer science methods on data to:

- 1. Draw causal insights
 - SUPERVISED Machine Learning
- 2. Predict future events
 - **SUPERVISED** Machine Learning
- 3. Understand patterns in data
 - **UNSUPERVISED** Machine Learning

Supervised ML data structure

Transaction 1

Transaction 2

Transaction 3

Transaction ...

Transaction data A	Transaction data B	Transaction data C	Transaction data D

Fraud probability	

Target variable

Tra	0000	tion	1
Tra	nsac	:TION	

Transaction 2

Transaction 3

Transaction ...

Transaction N

Transaction data A	Transaction data B	Transaction data C	Transaction data D

Target variable

Fraud probability



Input features

Data about transactions that the business collected (input features)

T	d
Iransaction	

Transaction 2

Transaction 3

Transaction ...

Transaction N

Transaction data A	Transaction data B	Transaction data C	Transaction data D

Target variable

Fraud probability	

Example input features

Transaction 1

Transaction 2

Transaction 3

Transaction ...

Past fraud count	Time of transaction	Declined in T-30 days	Amount
20	3 am	Yes	5.25 USD
1	9 pm	Yes	19.5 USD
0	9.30 am	No	500 USD

Fraud	
Yes	
Yes	
No	

Using input features

Use these data points

Transaction 1

Transaction 2

Transaction 3

Transaction ...

Past fraud count	Time of transaction	Declined in T-30 days	Amount
20	3 am	Yes	5.25 USD
1	9 pm	Yes	19.5 USD
0	9.30 am	No	500 USD

Fraud
Yes
Yes
No

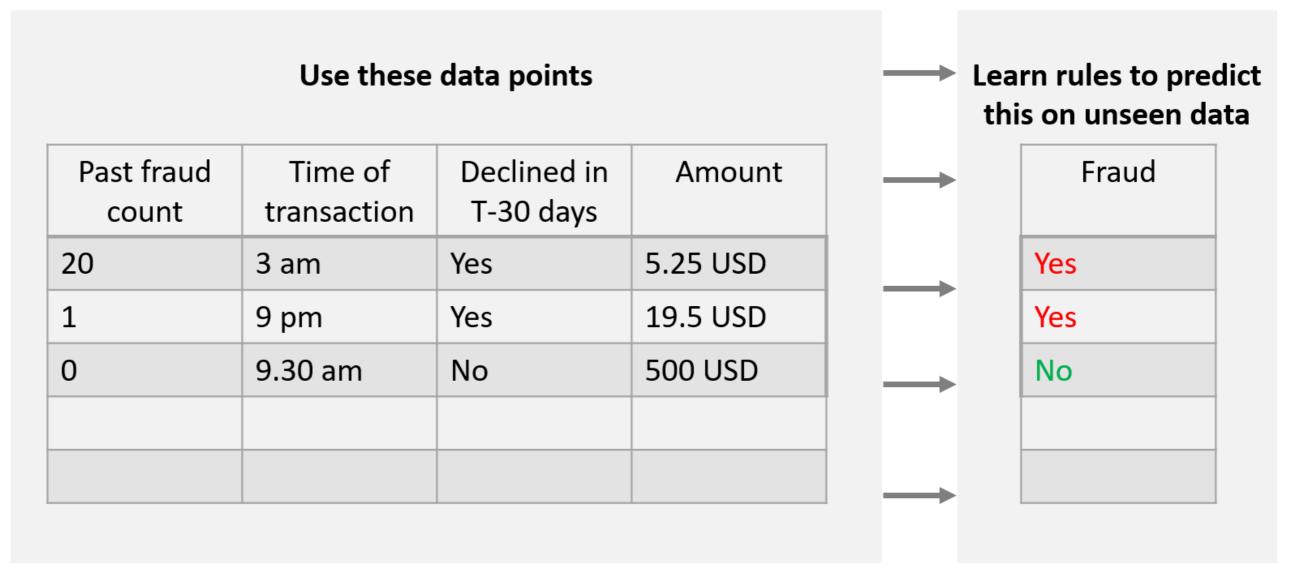
Predicting target variable

Transaction 1

Transaction 2

Transaction 3

Transaction ...





Unsupervised ML data structure

Transaction 1

Transaction 2

Transaction 3

Transaction ...

Transaction data A	Transaction data B	Transaction data C	Transaction data D



Unsupervised input features

Use these data points

Transaction 1

Transaction 2

Transaction 3

Transaction ...

Transaction data A	Transaction data B	Transaction data C	Transaction data D



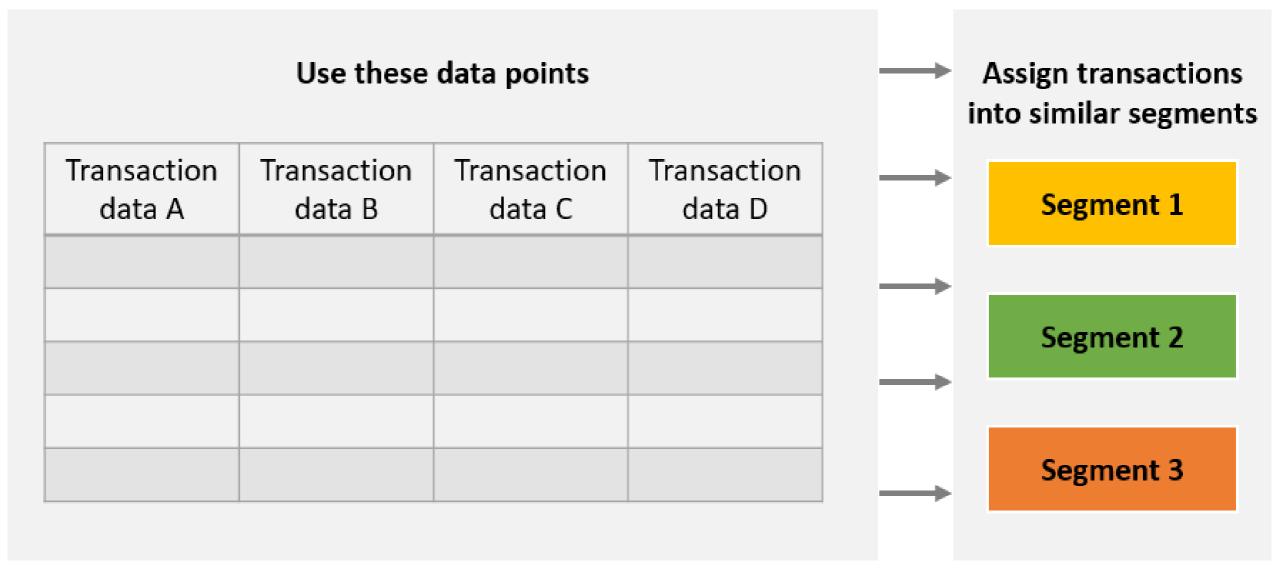
Unsupervised ML results

Transaction 1

Transaction 2

Transaction 3

Transaction ...



ML examples - Marketing

SUPERVISED Machine Learning:

- Predict which customers are likely to purchase next month
- Predict each customer's expected lifetime value

UNSUPERVISED Machine Learning:

Group customers into segments based on their past purchases

ML examples - Finance

SUPERVISED Machine Learning:

- Identify key transaction attributes that indicate a potential fraud
- Predict which customers will default on their mortgage payments

UNSUPERVISED Machine Learning:

• Group transactions into segments based on their attributes to understand which segments are the most profitable

ML examples - Manufacturing

SUPERVISED Machine Learning:

- Predict which items in production are likely faulty and should be manually inspected
- Predict which machines are likely to break and need maintenance

UNSUPERVISED Machine Learning:

 Group readings from machine sensors and identify anomalies for potential manufacturing malfunctions

ML examples - Transportation

SUPERVISED Machine Learning:

- Predict the expected delivery of the parcel
- Identify the fastest route for driving
- Predict product demand to prepare enough stock, rent/buy vehicles and hire workers

Let's practice!

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Job roles, tools and technologies

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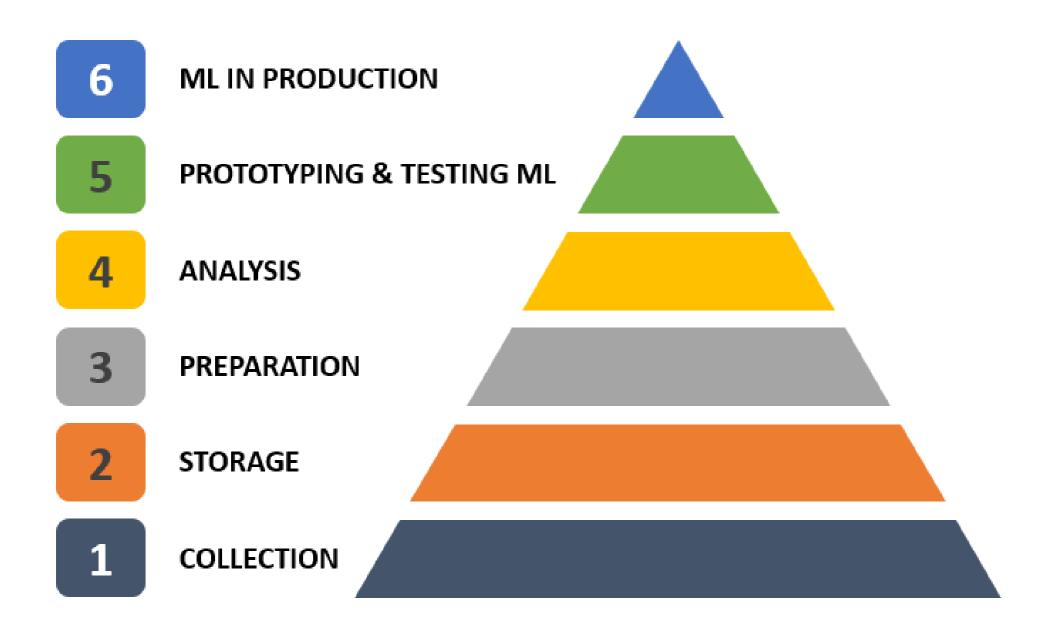


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Data pyramid and roles



Infrastructure owner





Data Engineer

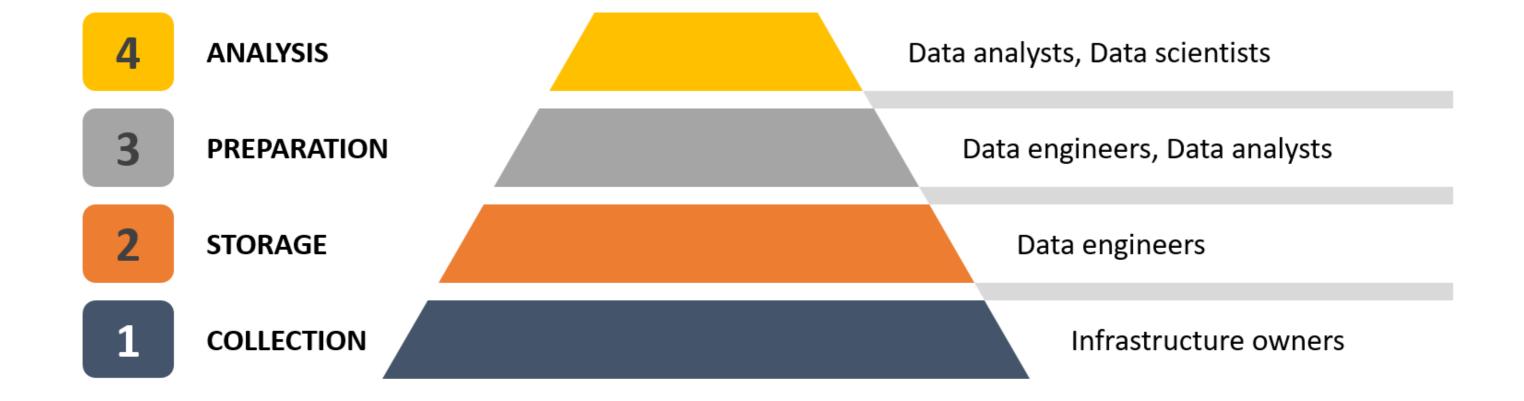


Data Analyst

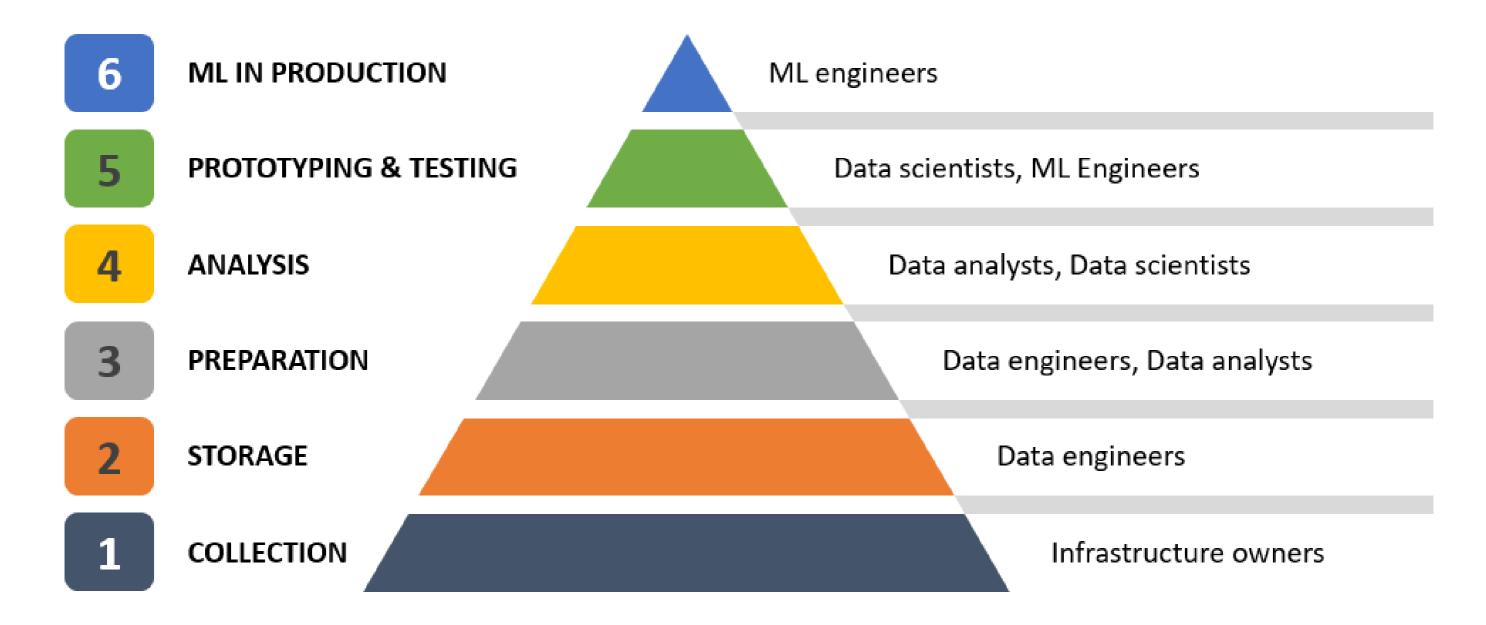




Data Scientist



Machine Learning Engineer



Team structure

- 1. Centralized
- 2. Decentralized (or embedded)
- 3. Hybrid

Team structure comparison

- 1. **Centralized** all data functions in one central team. Works well for small companies, startups, new organizations. Gets slow once business matures and requires focus
- 2. **Decentralized** each business unit, geography or department have their own data functions. Works well for larger companies. Introduces issues with data governance, differences in definitions, redundancies, and added complexity
- 3. **Hybrid** infrastructure, definitions, methods and tooling are **centralized**, while application and prototyping **decentralized**

Let's practice!

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