

AI and Analytics : Opportunities Landscape

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Data scientist – Technical Lead

Outline

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AI and Analytics: Introduction

Introduction

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AI trends

Overview of AI trends and economic opportunities

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AI/ML applications to different industries

The different kinds of AI/ML problems across industries

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Data analytics – Roles and careers

Career roles and career opportunities in a data driven company

5

Data scientist Role

Data scientist role and how to prepare a career in data science

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AI and analytics – Introduction



Advent of Big data analytics

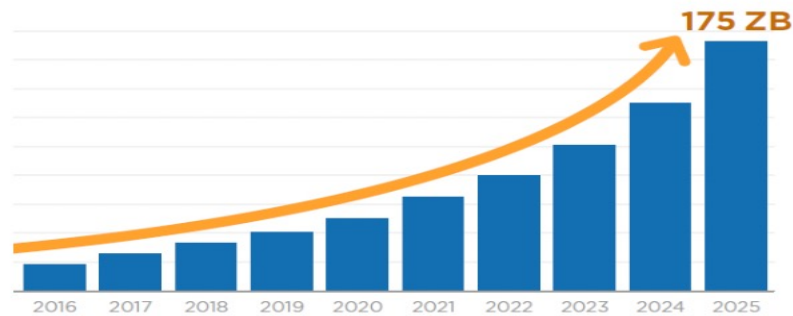
DRIVERS OF DATA GROWTH

- Digital transformation of businesses
- Proliferation of devices
- Logarithmic drop of storage costs



EXPLOSION OF DATA

- Data collected mobiles, sensors, and IoT devices
- Data generated in enterprise-hardened servers
- Applications embed data collection as part

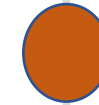


IDC says 175 ZB will be created by 2025 (Image courtesy IDC)

BIG DATA TIMELINE



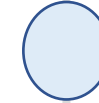
1997: Google launches search engine



1999: "Big data" term used first



2005: "Hadoop" paper



2010: Spark paper

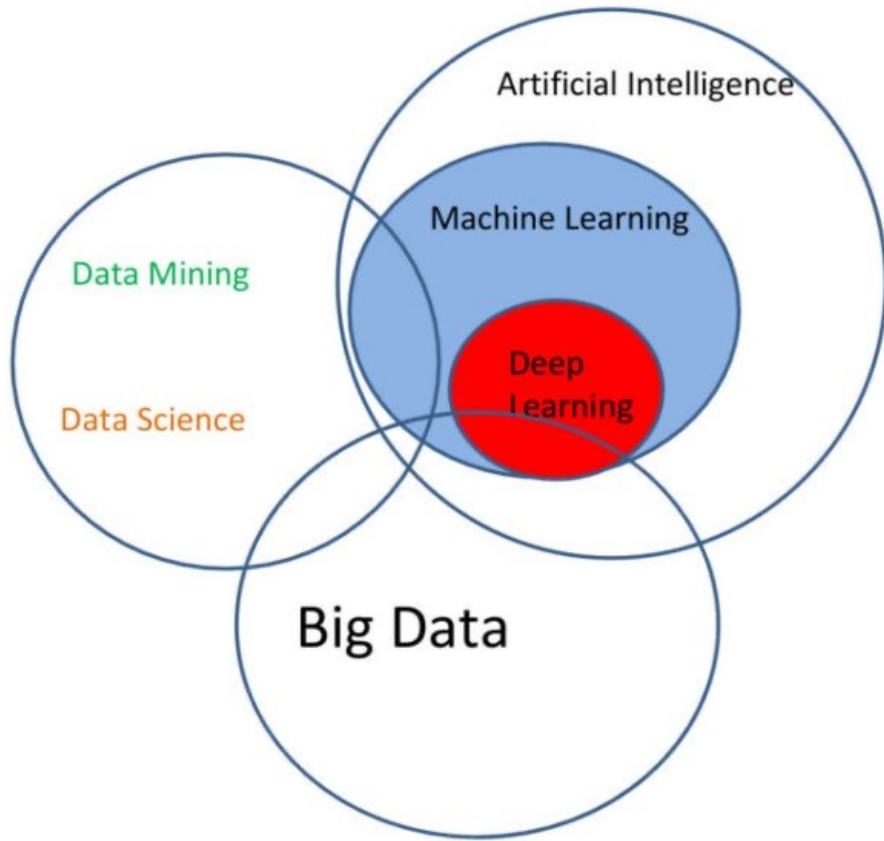


2012 : Role and JD for data scientist became popular



2015-2018: Bigdata on the cloud

AI, Machine learning and Big data analytics



Artificial intelligence: technology that mimics human intelligence

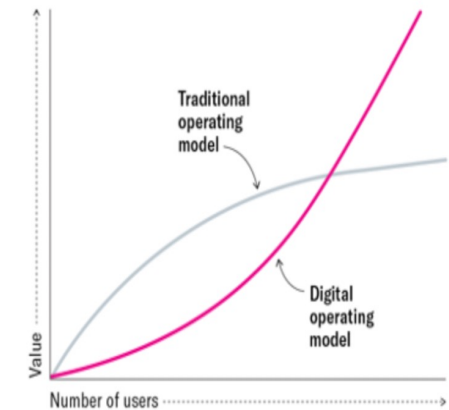
- visual perception
- speech recognition
- natural language understanding.
- automated decision-making

AI techniques include machine learning, deep learning, natural language processing, computer vision, and robotics.

Big data is the fuel on which artificial intelligence runs

How AI-Driven Companies Can Outstrip Traditional Firms

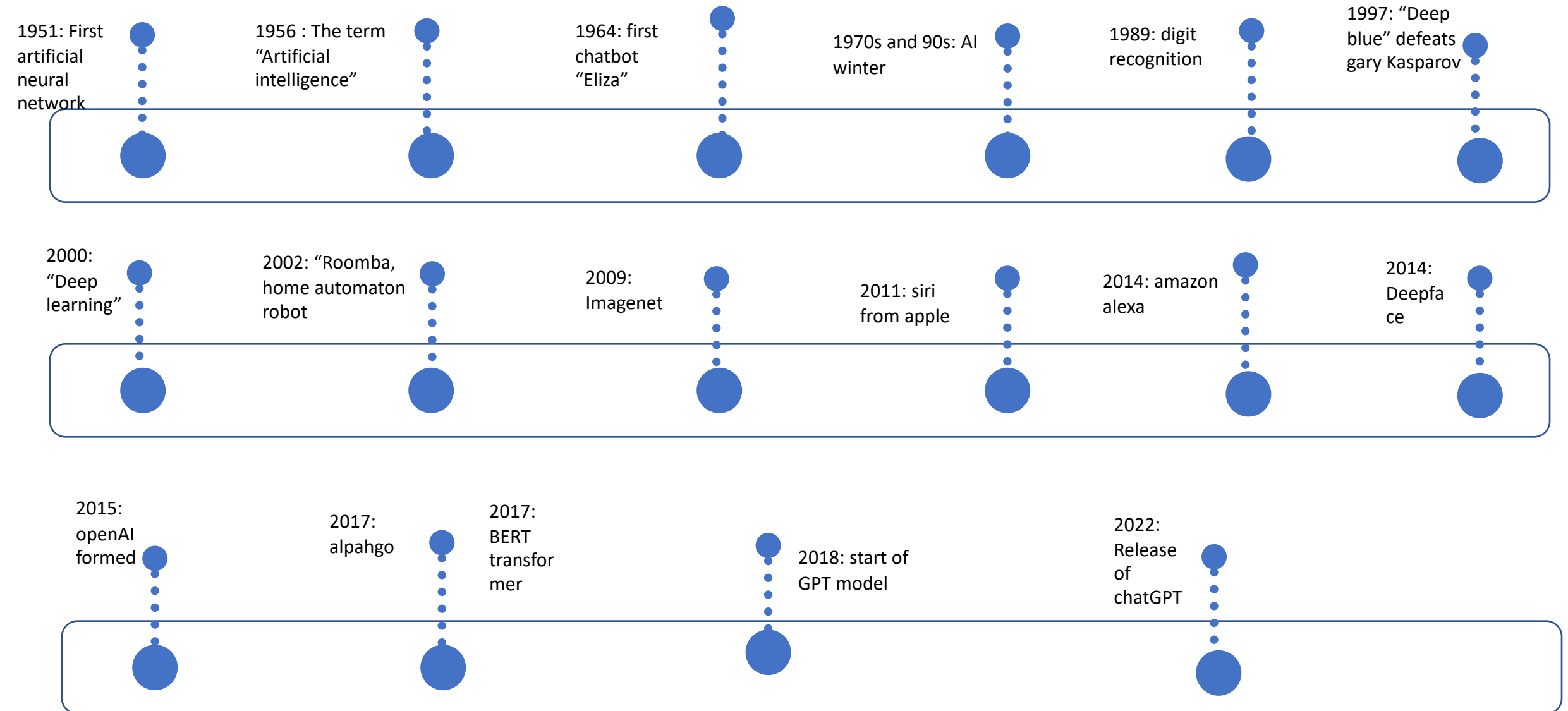
The value that scale delivers eventually tapers off in traditional operating models, but in digital operating models, it can climb much higher.



From: "Competing in the Age of AI," by Marco Iansiti and Karim R. Lakhani, January-February 2020

HBR

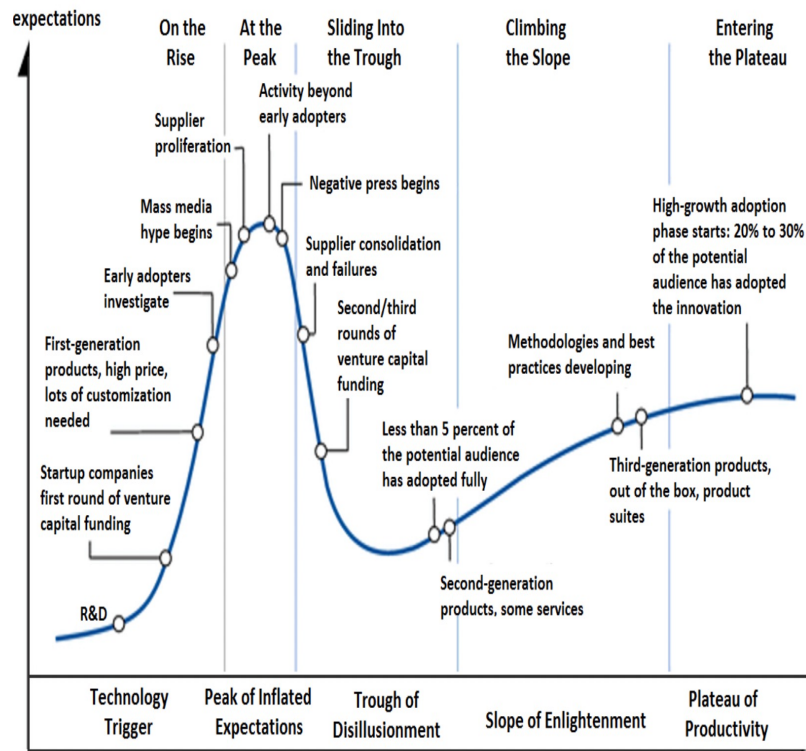
AI machine learning - Timeline



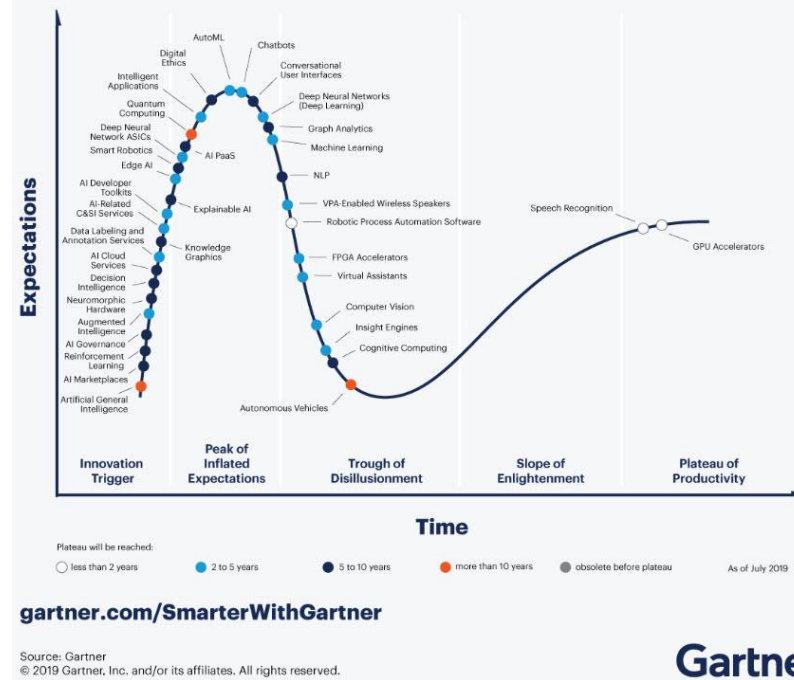


AI trends

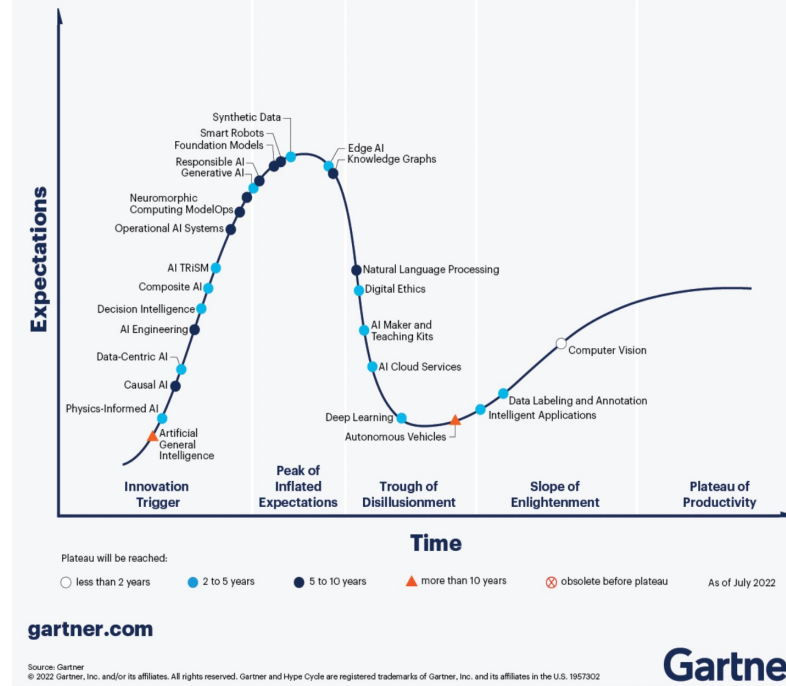
AI Trends: Gartner hype cycle for AI



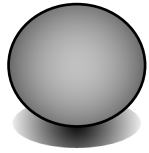
Gartner Hype Cycle for Artificial Intelligence, 2019



Hype Cycle for Artificial Intelligence, 2022

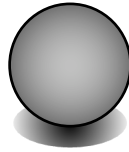


AI Trends



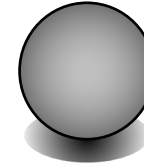
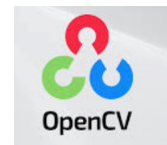
Deep Learning

- relies on the technique of neural networks usually with many layers
- Neural nets can model any function
- With increasing data, the performance keeps increasing, uncovers complex patterns
- Many applications : robotics, Cybersecurity, Smart agriculture



Computer vision

- Significant progress in Computer vision problems: object detection, tracking, face recognition, image segmentation etc
- Many applications: agri-tech, autonomous cars, contactless retail etc



Chatbots/NLP

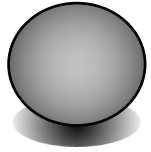
- Significant progress in generating next best response
- Improvements needed in terms of contextual awareness, multilingual, nuances in human language
- Applications in customer service, teaching, marketing



Hugging Face

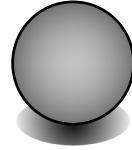


AI Trends



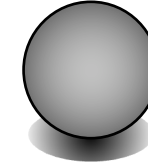
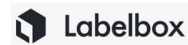
Synthetic data

- lower-quality substitute and used when real data is inconvenient
- banks and financial services institutions use synthetic data by setting up multiagent simulations to explore market behaviours



Data labelling/annotation

- process of attributing, tagging, or **labeling data** to help machine learning algorithms understand and classify the information
- -image, text and audio annotation
- - Semi supervised learning, unsupervised learning



Generative AI and LLMs

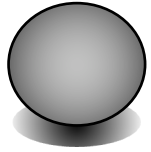
- A broad label description of AI that uses unsupervised learning algorithms to create images, video, audio, text or code
- LLM, or Language Model with Latent Variables, is a type of probabilistic model used in Natural Language Processing (NLP) that extends traditional language models by incorporating hidden, or latent, variables into the model.
- many applications in a variety of fields. In the entertainment industry, gaming industry, healthcare industry, marketing etc



Hugging Face

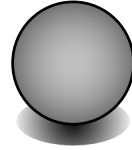
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AI Trends



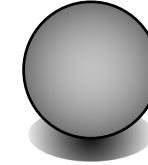
Knowledge graphs

- KG is a kind of semantic network with added constraints
- encode human knowledge leveraging a graph-based structure
- machine learning models will have better explainability and trustworthiness.
- Applications: Finance domain for fraud detection, social media, new product analytics



Smart robots

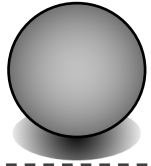
- learning about their environment and experience.
- better customer experience for doing repetitive tasks
- Applications: Home assistant robots,
- Service robots in hospitality, restaurants etc



Causal Machine learning

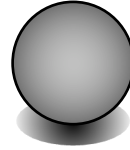
- AI is good at predicting “when X then Y”. However causal ML tries to understand why
- Some of the key challenges in causal machine learning include dealing with selection bias, identifying appropriate causal assumptions, and handling complex causal structures.
- Causal machine learning algorithms typically use observational or experimental data to infer causal relationships between variables

AI Trends



AI trism

- AI trust, risk and security management (AI TRiSM) ensures AI model governance, trustworthiness, fairness, reliability, robustness, efficacy and data protection. This includes solutions and techniques for model interpretability and explainability
- Manage AI risk are much more likely to experience negative AI outcomes and breaches. Prevent financial and reputational loss
- Avoid Models not performing as intended
- Compliance to AI regulations
- Manage security and privacy failures








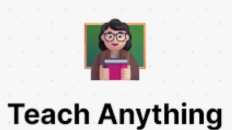








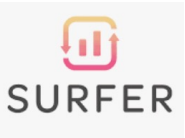


AI governance

1. Lifecycle management of AI models
2. MLOPS challenge: sensing and interpret a poorly working AI system. Designing alerts for problems
3. continuously learning AI models and monitor learning

Other ML topics

- Hardware accelerators:
- Neuromorphic computing
- Reinforcement learning
- Self supervised learning
- Physics informed neural nets

AI Tools

Writing assistants		Customer support	
Meeting assistants	 	Schedule and plan	
Learning & Research tools	 	Presentation	 
Note taking tool	 	Translation: Text-to-speech Text-to-music Text-to-image	  
Marketing	  		

A comprehensive collection of AI tools : <https://www.futuretools.io/>



AI/ML Applications in different industries



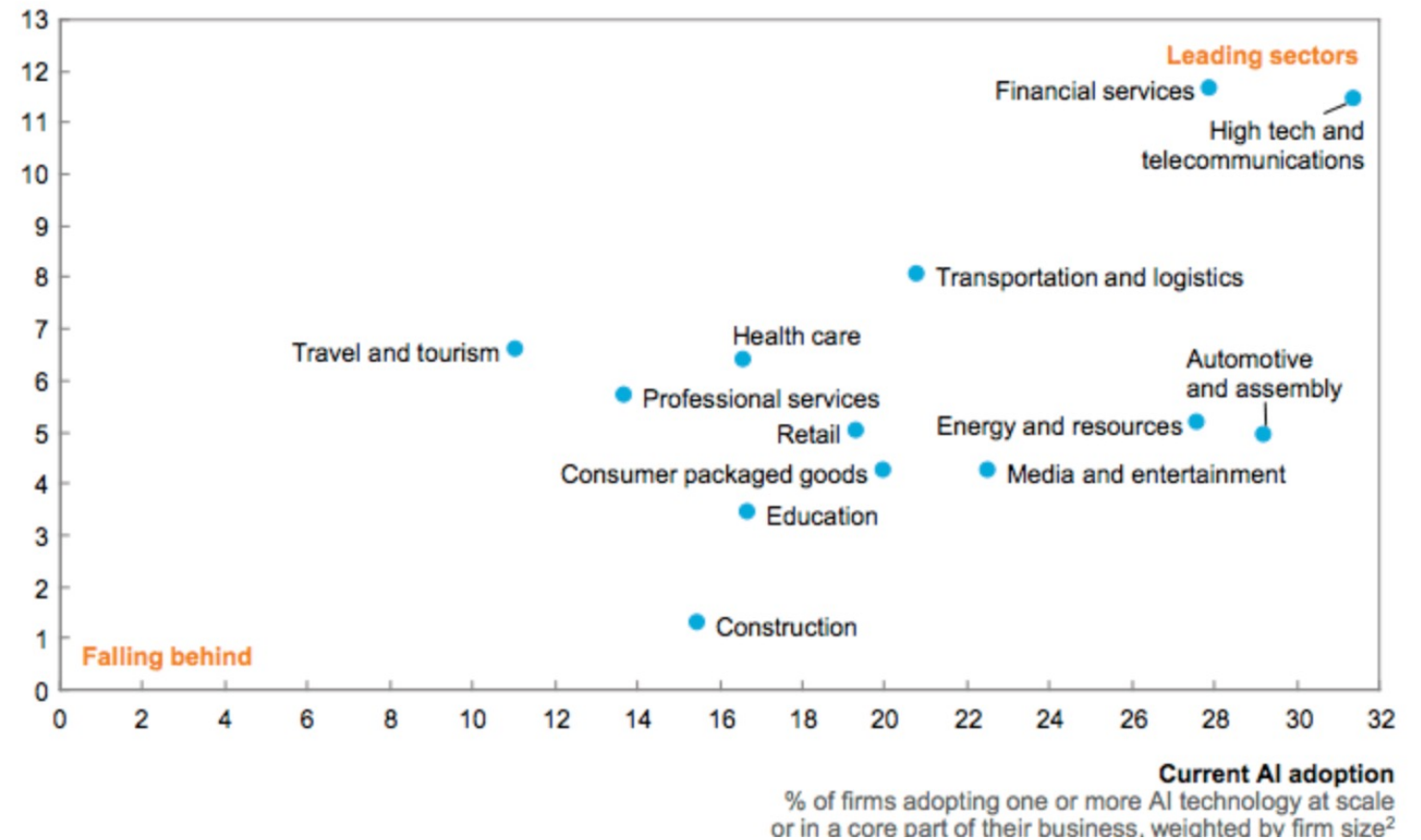
AI in different sectors

- All economic sectors adopt AI
- Driving value by
 - Decreasing expenses
 - Innovation: Help Launch new products or services
 - Improved decision making
 - Improved productivity
 - Enhancing Reliability security

Sectors leading in AI adoption today also intend to grow their investment the most




Future AI demand trajectory¹

Average estimated % change in AI spending, next 3 years, weighted by firm size²






SOURCE: McKinsey Global Institute AI adoption and use survey; McKinsey Global Institute analysis




AI/ML problems in different sectors

	SECTOR	AI/ML APPLICATION
	Telecom	<ul style="list-style-type: none">• How to detect churn ?• How to manage the network infrastructure effectively and proactively find problems?• How to segment customers based on value? How to predict customer lifetime value?• Dynamic pricing, promotion• How to optimize call center operations?
	Financial services	<ul style="list-style-type: none">• Risk modelling• Fraud detection• Algorithmic trading• Robo-advisors
	Retail	<ul style="list-style-type: none">• Inventory and merchandising management• Personalization




AI/ML problems in different sectors

SECTOR	AI/ML APPLICATION
	<p>Manufacturing</p> <ul style="list-style-type: none">• Predictive maintenance• Robotic process automaton• Industrial IOT applications
	<p>Agriculture</p> <p>Precision farming Crop monitoring Soil analysis and monitoring Dynamic pricing</p>
	<p>Energy</p> <p>Demand prediction and optimize energy grids Decisions for where to drill for oil from seismic data</p>

AI/ML problems in different sectors

SECTOR	AI/ML APPLICATION
 <p>Travel tech & Transportation</p>	<ul style="list-style-type: none">• Travel assistant• UX personalization• Flight fare/hotel pricing• Matching riders to drivers• Travel route optimization• Dynamic pricing
 <p>Health care</p>	<ul style="list-style-type: none">• Disease diagnosis from medical reports and images• Drug discovery• Gene analysis and editing
 <p>Education</p>	<ul style="list-style-type: none">• Customized learning journeys• Personalization• Content creation

AI/ML problems in different sectors

SECTOR	AI/ML APPLICATION
	<p>Construction/Real estate</p> <ul style="list-style-type: none">• Building information modelling• Predict housing prices
	<p>Consumer Tech</p> <ul style="list-style-type: none">• Recommendation for cross selling• Customer purchase funnel optimization
	<p>Media and entertainment</p> <ul style="list-style-type: none">• Content generation and personalized content• AR/VR games

Indian AI startups across domains

AI PLATFORM

opslyft

FUTURE CARS



DRONES



INTELLIGENT ROBOTS



MEDICAL



HEALTH AND WELFARE



SMART MANUFACTURING



SMART CITY



AGRICULTURE



FINTECH



EDUTECH



REAL ESTATE



LAW/PATENT



TRAVEL



RETAIL/
E-COMMERCE



SECURITY



INSURANCE



FOOD/RESTAURANT



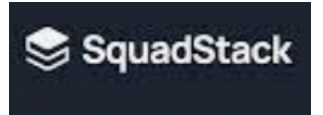
MEDIA/ENTERTAINMENT



DEVICES/IOT



CANTACT CENTER



MARKETING/SALES



TELECOM



GAMING



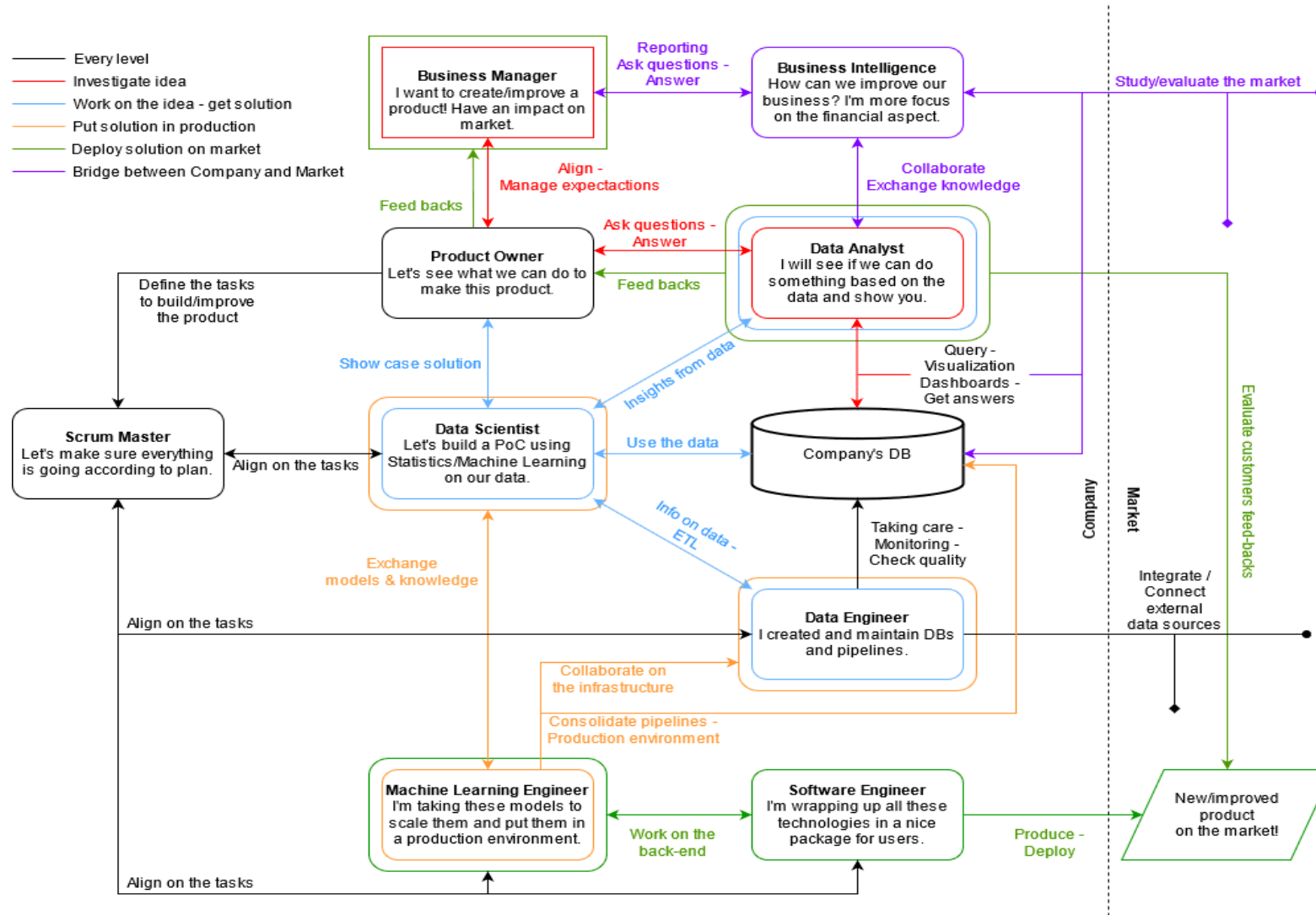
FASHION





Data driven companies and Data careers and roles

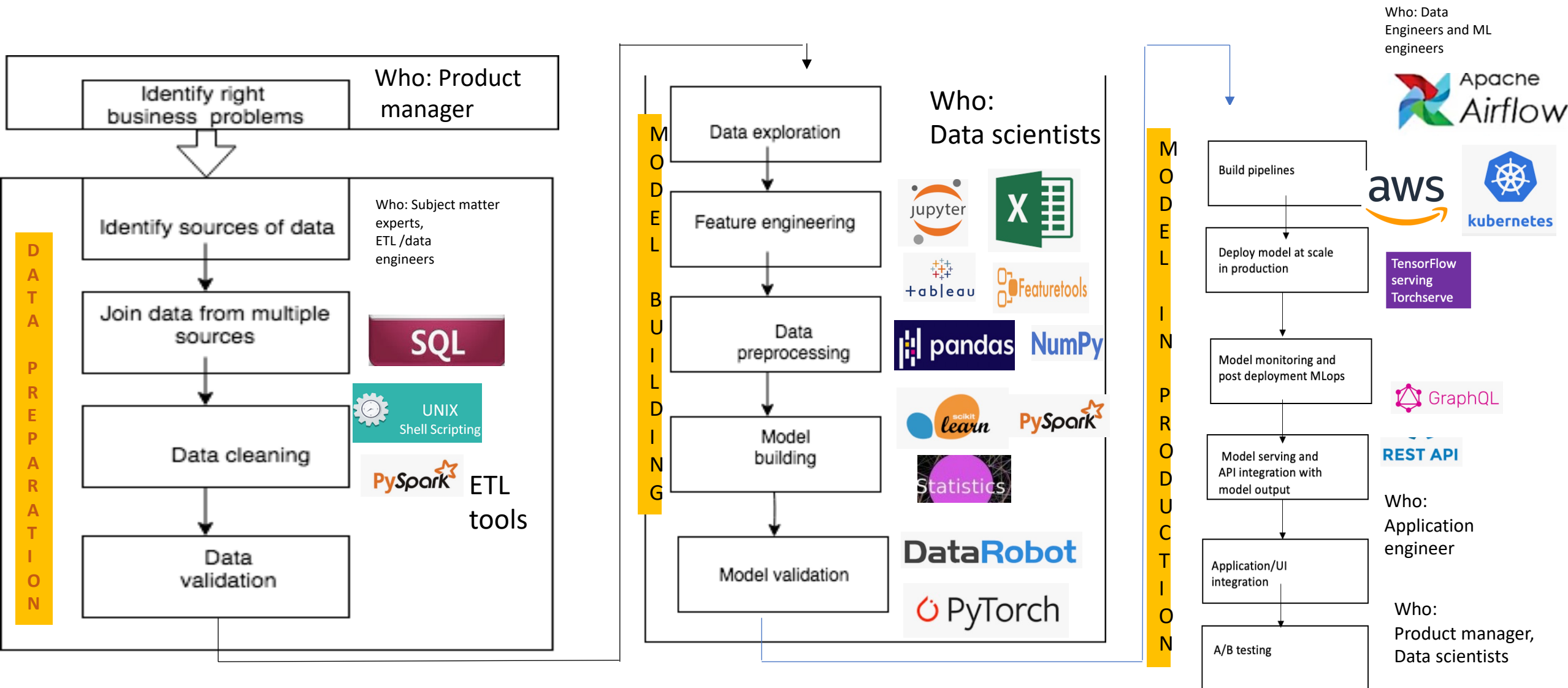
Different roles in creating a data product



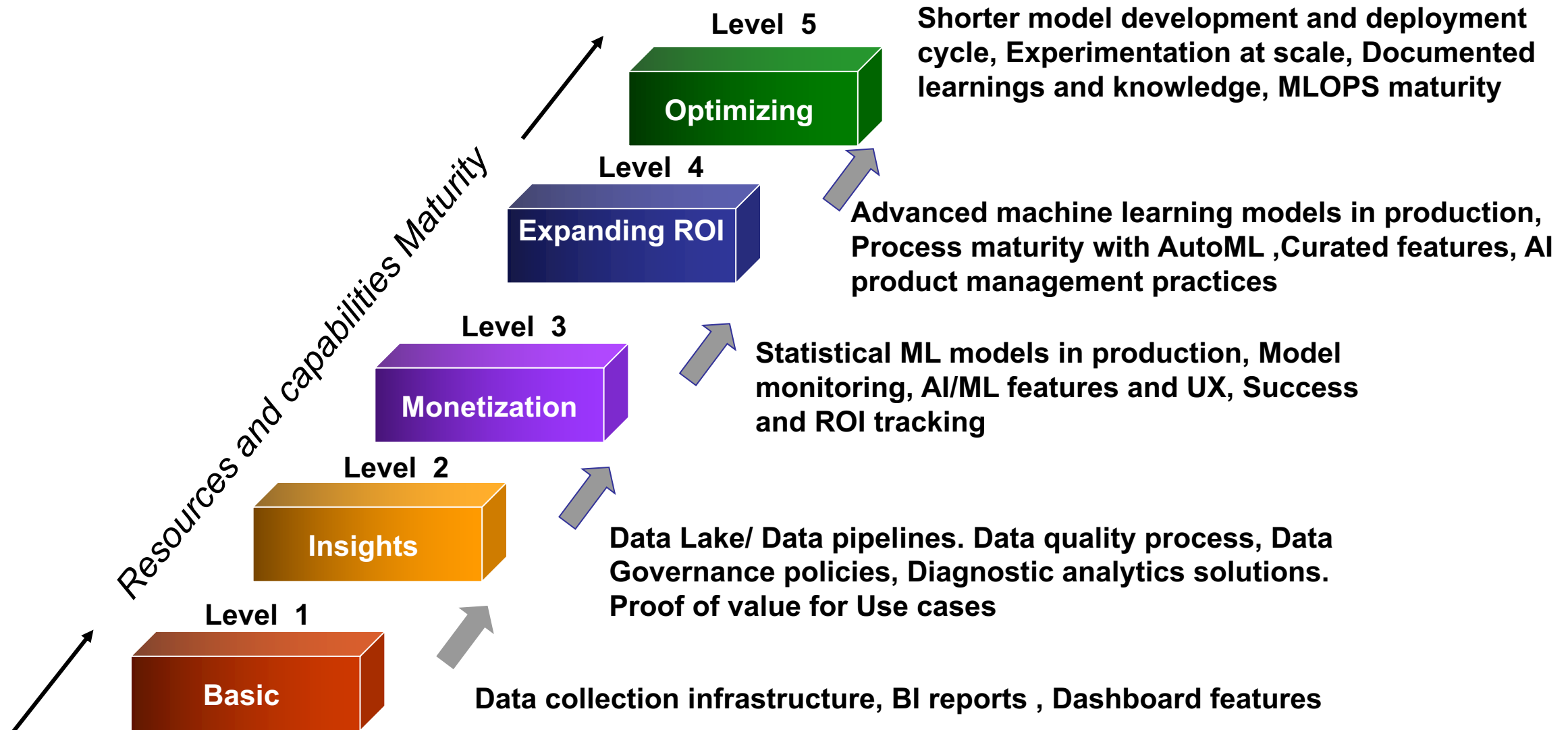
ROLES:

- Business manager
- Business Intelligence
- Product owner
- Scrum master
- Data scientist
- Data analyst
- Data engineer
- ML engineer
- Software engineer

ML model lifecycle



AI/ML org maturity levels



Data and analytics maturity index

Questions that assess maturity:

Data & Analytics Business Maturity Index – Stage Maturation

Version 2.0

	Business Monitoring	Business Insights	Business Optimization	Insights Monetization	Digital Transformation
Data Management	<ul style="list-style-type: none"> Data Warehouses Structured Data Aggregated Data Batch ETL Data Governance / Quality 	<ul style="list-style-type: none"> Data Lakes Unstructured data Granular data External / 3P data Data Pipelines 	<ul style="list-style-type: none"> Data Mesh IOT / Digital Twins / Sensors Intelligent Data Pipelines Real-time (Streaming) Data Feeds Published APIs 	<ul style="list-style-type: none"> Analytic Profiles Asset Hierarchies 	<ul style="list-style-type: none"> Contextual Knowledge Center Partner Data Ecosystem
Analytics Capabilities	<ul style="list-style-type: none"> Statistical Analysis Descriptive Analytics BI Reports / Dashboards Comparative Analysis Time-series Forecasting 	<ul style="list-style-type: none"> Collaborative Data Science Methodology ML / Predictive Analytics "Right time" Analytics Personalization Decision Engine 	<ul style="list-style-type: none"> Deep Learning / Neural Networks Prescriptive Analytics Propensity Models AutoML RPA / Automation RT Streaming Analytics Edge Analytics 	<ul style="list-style-type: none"> Human-AI Data Science Method Analytic Scores Reinforcement Learning Transformers AI Simulations AI-based Learning Assistants 	<ul style="list-style-type: none"> Autonomous Analytics AI Meta-Learning Techniques Economics of Data & Analytics Data & Analytic Asset Mgmt
Business Alignment	<ul style="list-style-type: none"> Management Reports Operational Dashboards Report Library 	<ul style="list-style-type: none"> Initial Use Cases ROI Data Monetization Pilots Use Case Priority Process Propensity Scores Next Best Offer / Recommendations 	<ul style="list-style-type: none"> Expanding Use Cases ROI Data Monetization Expansion Intelligent Applications Data Products / Data Apps Reusable Propensity Scores 	<ul style="list-style-type: none"> "Smart" Spaces "Smart" Things Use Case Library Insights-as-a-service Data Monetization Factory 	<ul style="list-style-type: none"> Transform Business Models Reengineer Value Chains Reinvent Operational Processes Hyper-personalized Services
Culture	<ul style="list-style-type: none"> BI Center of Excellence 	<ul style="list-style-type: none"> Chief Data & Analytics Officer Data Literacy Training Design Thinking Training 	<ul style="list-style-type: none"> Responsible AI Critical Thinking Training Design Lab Data Science 2.0 	<ul style="list-style-type: none"> Hub-and-spoke Analytics Org AI Innovation Center Design School (d.school) Ethical AI / Enterprise Automated Governance Human-AI Partnership Monetization Governance 	<ul style="list-style-type: none"> Human-AI Learning Interface Culture of Experimentation Organizational Improv Catalyst Network Comp/Reward Sharing Abundance Mentality Individual Empowerment / Innovation

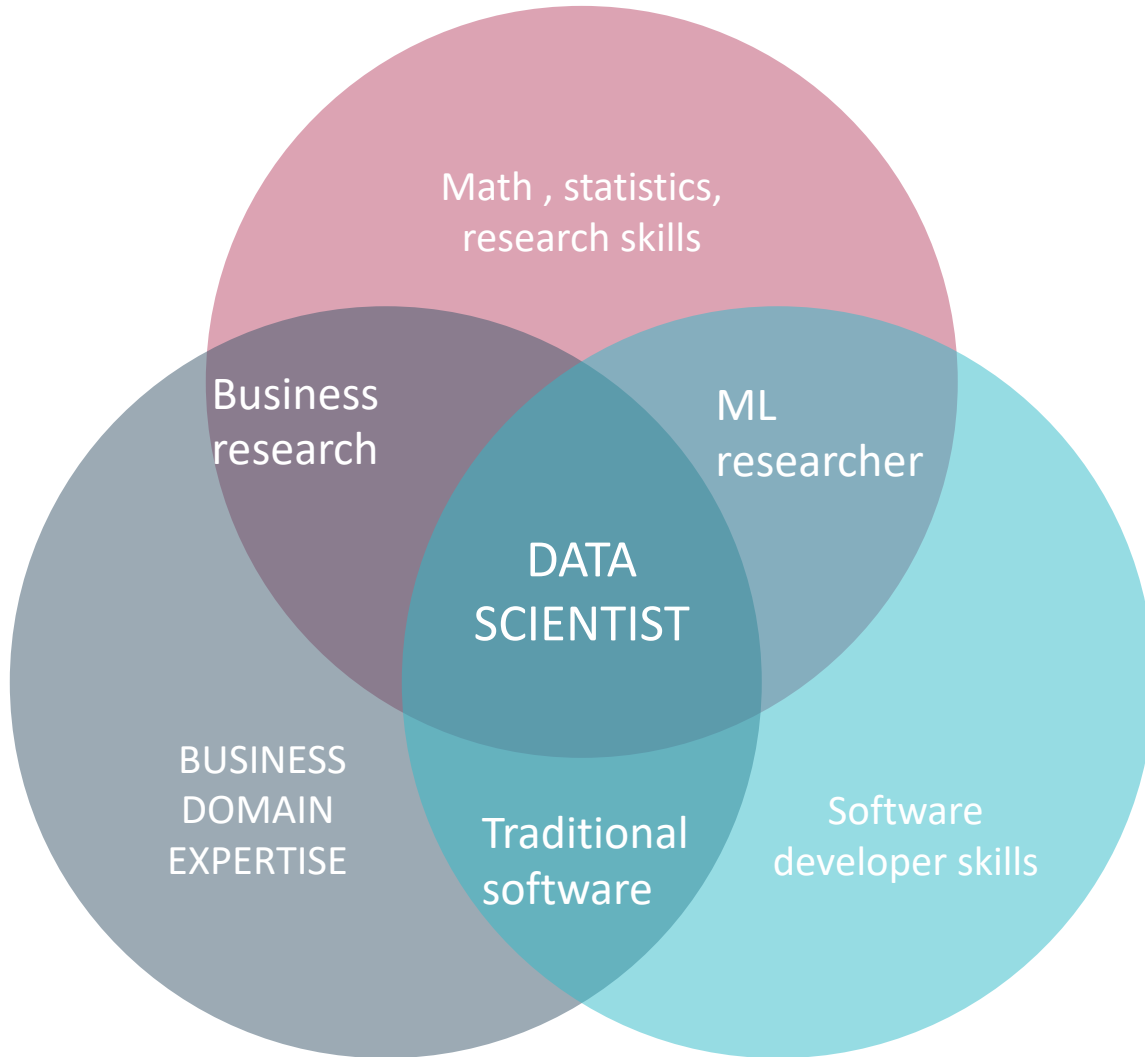
Source: Bill Schmarzo "Big Data MBA" Course Curriculum

1. how much data is generated and stored everyday?
2. Where do data reside? What kind of data platform? How many pipelines run everyday?
3. what is the model deployment platform? Do you use in-house infra/cloud ?
4. is there a feature store for curated features
5. how is data quality assessed within the org?
6. how is the data governance policies? how is access controlled to these data
7. what tools you use to discover data?
8. are there tools to find out lineage of a data source or table?
9. what are the tools for model development? what experiment tracking tools do data scientists use?
10. what processes are there to assess model bias?
11. how many A/B test/experiments are done an year?
12. what kind of post model MLops is done? how do you identify drift?
13. do you track the cost of model deployed? how are models retired?
14. how many real time models are deployed per year?
15. what is the ratio for data analysts: data scientists: data engineers: ML engineers?
16. how are the results of data analysis documented and propagated?
17. what is the process and reviews do models go through before deployment?
18. how are the model outputs integrated with UX?
19. how many deep learning models run in production? what are the use cases ?
20. Do you use any AutoML tools for feature engineering and modelling?



Data scientist role

Data scientist role



THREE MAIN EXPERTISE FOR DATA SCIENTIST

1. Software developer skills:
2. Business domain expertise:
3. Math, statistics and research skills

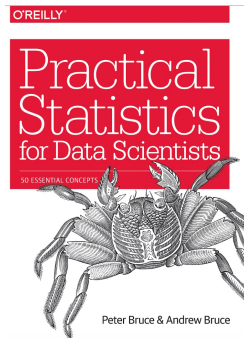
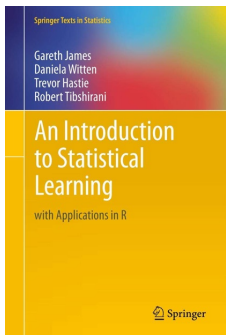
LinkedIn titles with data science role:

- Experimental scientist
- Decision scientist
- NLP engineer
- Computer vision engineer
- AI researcher ..etc.

Data science Knowledge areas

ML Essentials for beginners:

- Linear regression
- Logistic regression
- KNN, LDA
- Ridge and Lasso regression
- Feature selection
- Cross validation and bootstrap
- Bias variance tradeoff
- Decision trees
- Bagging , Random Forest
- Boosting
- SVMs
- PCA
- K-means clustering and Hierarchical clustering



	Causal inference and causal ML	Simulation	Tuning deep learning and other architectures RNN, LSTM, CNN	Operations research	Reinforcement learning	LLM	Bayesian approaches	Experiment design
LEVEL3								
	Anomaly detection	Deep learning	Computer vision	recommender systems	Ranking models	Time series models	NLP	
LEVEL2								
	ML model deployment	Model validation - Bias variance tradeoff - Learning curve Cross validation	Hyper parameter tuning	A/B testing	Feature selection methods	Feature engineering - data transformation - outlier detection - binning		
LEVEL1	Model explainers							
LEVEL0		SQL	Descriptive statistics	Statistical hypothesis test	Data visualisation	Basic ML algorithms from scratch		

Data scientist resume

Resume header

FIRST LAST
Bay Area, California • +1-234-456-789 • professionalemail@resumeworded.com • linkedin.com/in/username

SKILLS

Skills

- Python, Pandas, scikit-learn
- Apache-spark
- R
- SQL
- Tableau

Align skills to job description

Project portfolio

Projects

Movie Recommendation System

- Used k-nearest-neighbors to build a recommendation model for movies
- The model outperformed achieving 18% improvement over baseline

Attrition model

- Built a employee attrition model using random forest and achieved results with estimated savings of 10 million\$

Sales forecast model

- Implemented a time series model to forecast the sales and improved forecast accuracy by 40%

Project portfolio expected to have a regression, classification, recommendation problems along with a data analysis project

Resume bullet point template



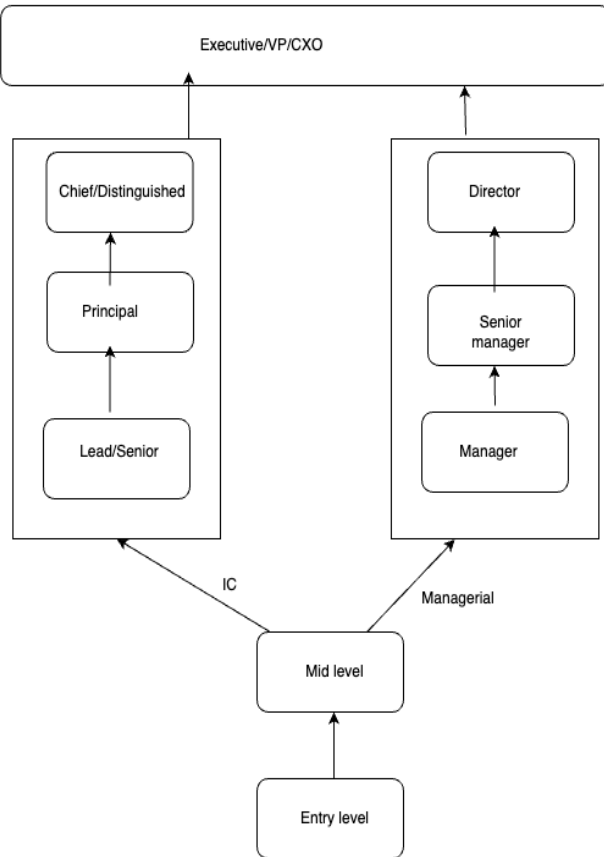
Public profiles

Github: <https://github.com/xyyhz>
Linkedin: <https://linkedin.com/in/xyyhz>
Blog: <https://medium.com/@xyyhz>

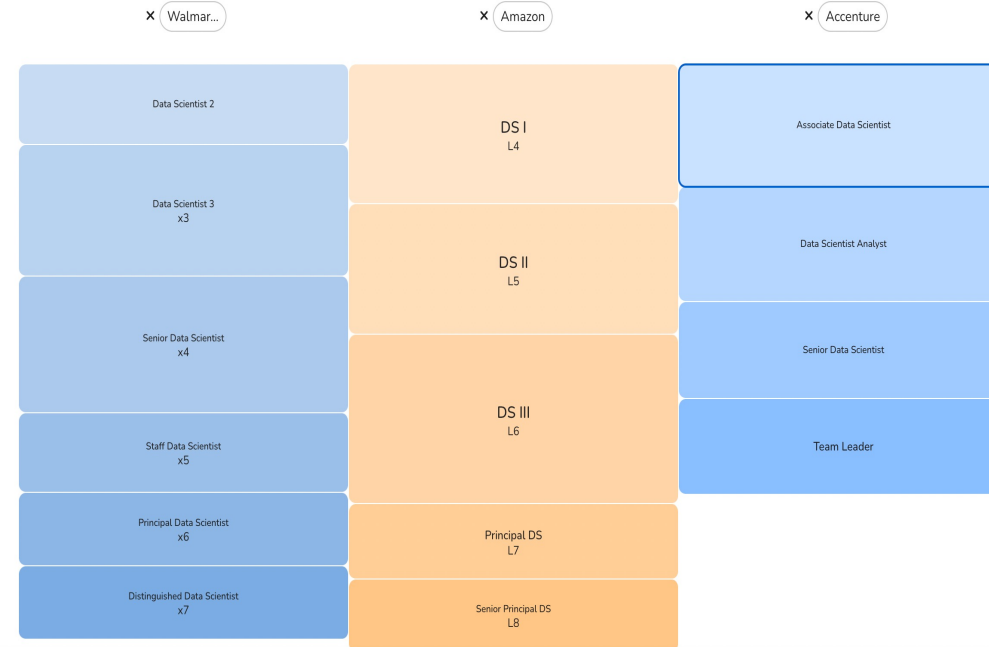
- GitHub home page
- Technical blog
- Personal website etc

Data scientist – Career progression

General tech career progression

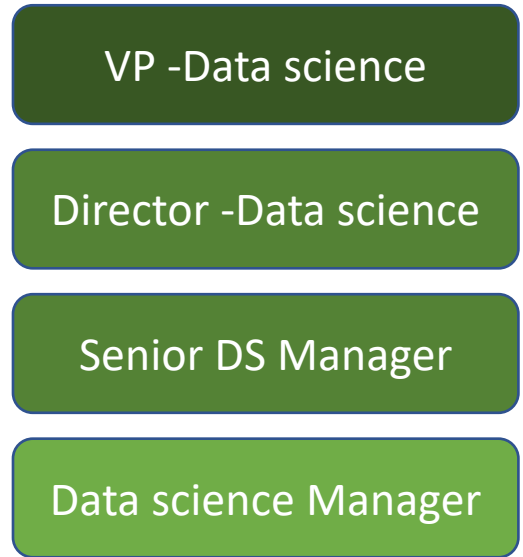


IC levels



<https://www.levels.fyi/t/data-scientist?countryId=254>

managerial levels



Thanks



All the best

LinkedIn connect



<https://www.linkedin.com/in/srimugunthan-dhandapani/>