

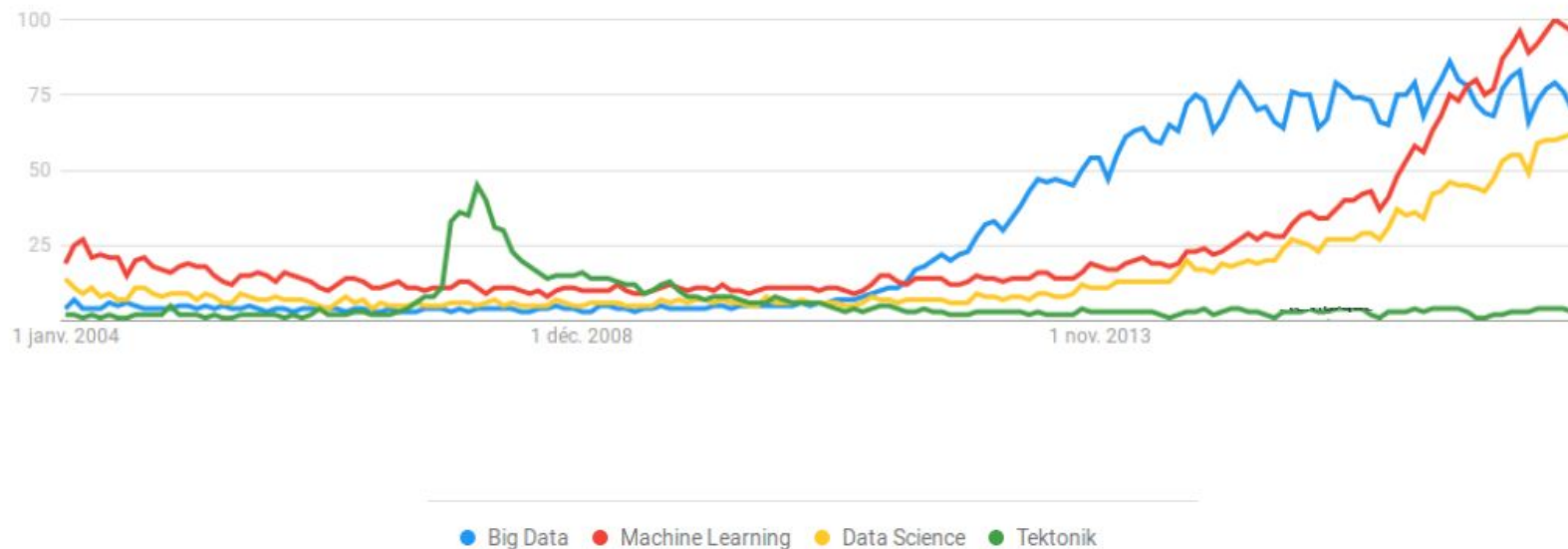


# THE AI ERA IS HERE

**Ignited by a new computing model,  
GPU Deep Learning**

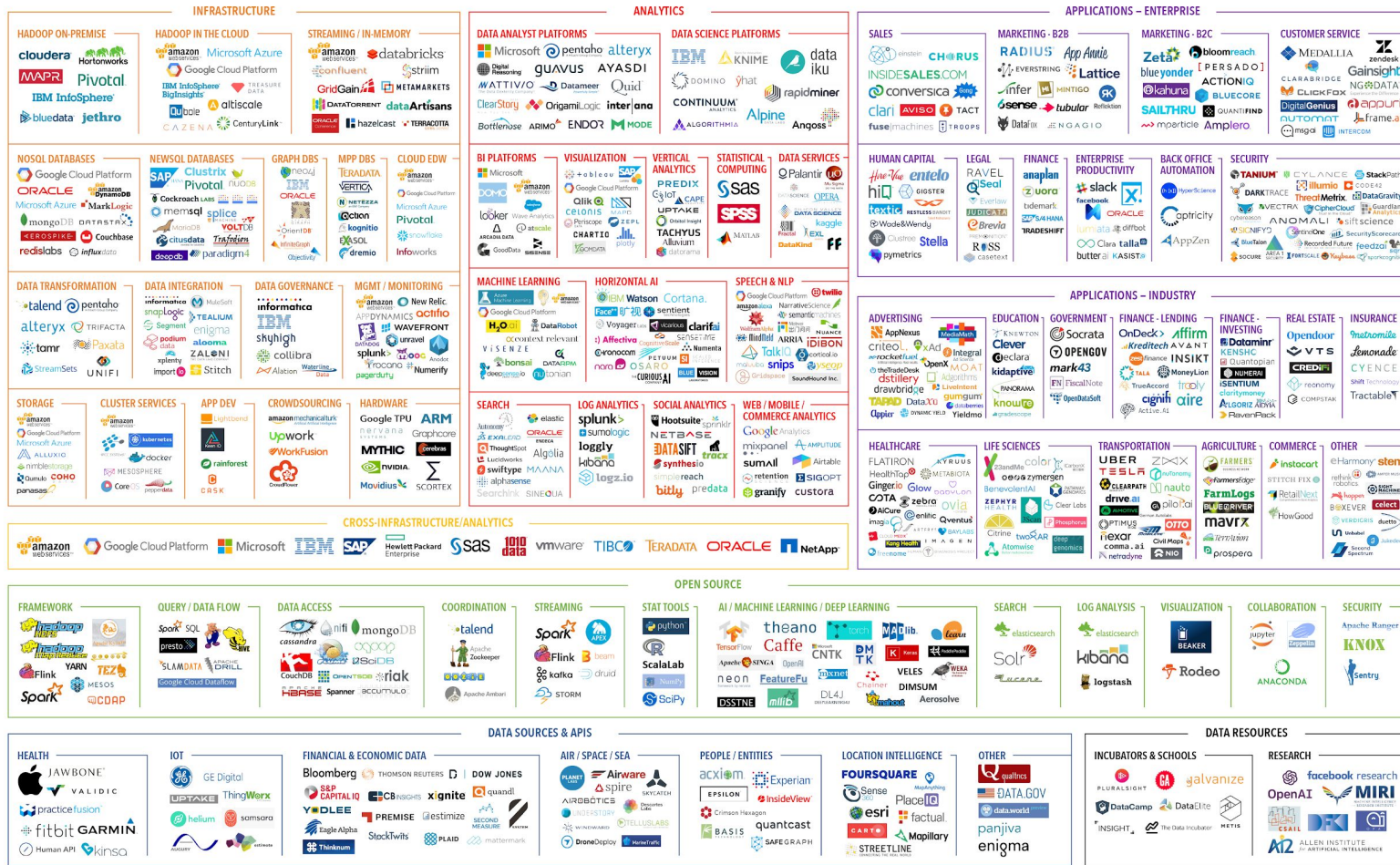
<https://www.slideshare.net/NVIDIA/the-ai-era-ignited-by-gpu-deep-learning>

# Trends





# BIG DATA LANDSCAPE 2017



## BIG DATA &amp; AI LANDSCAPE 2018

## INFRASTRUCTURE

**HADOOP ON-PREMISE**

- Hadoop
- Cloudera
- MapR
- Pivotal
- IBM InfoSphere
- bluemix jethro

**HADOOP IN THE CLOUD**

- aws Microsoft Azure
- Google Cloud
- IBM InfoSphere BigInsights
- TREASURE DATA
- Databricks
- Oracle
- CAZENNE
- CenturyLink

**STREAMING / IN-MEMORY**

- aws databricks stream
- Confluent
- GridGain
- DATA-TORRENT
- DataArtisans
- ORACLE
- hazelcast
- TERRACOTTA

The collage displays logos for several database technologies, organized into five categories:

- NoSQL DATABASES:** Includes logos for Google Cloud, AWS, Oracle, Microsoft Azure, MongoDB, MarkLogic, Aerospike, DataStax, Amazon DB, Couchbase, Redis Labs, and Scylla.
- NewSQL DATABASES:** Includes logos for SAP, Clustrix, NuDB, Pivotal, Cockroach Labs, Open Spanner, Microsoft SQL Server, InfluxData, Oracle, VoltDB, Splice Machine, Cloudera, and Paradigm.
- GRAPH DBs:** Includes logos for Neo4j, Amazon Neptune, IBM, and Oracle.
- MPP DBs:** Includes logos for Vertica, IBM Data Warehouse Systems, Alation, Cognitio, Exasol, and Dremio.
- CLOUD EDW - AWS:** Includes logos for Google Cloud, Microsoft Azure, Pivotal, Snowflake, and Informatica.

The diagram is organized into four columns, each representing a different category of data management tools. Each column has a header and a list of companies with their logos.

- DATA TRANSFORMATION:** Includes logos for Talend, Pentaho, Alteryx, Trifacta, TM1r, and StreamSets.
- DATA INTEGRATION:** Includes logos for Informatica, SnapLogic, Tealium, Segment, Enigma, Alchem, Zaloni, Veeva, and Stitch.
- DATA GOVERNANCE:** Includes logos for Informatica, IBM, SailPoint, OneSpan, Collibra, Alation, and Watermark.
- MGMT / MONITORING:** Includes logos for AWS, New Relic, Octio, Rubrik, AppDynamics, Wavefront, Dynatrace, Splunk, SignalFx, Druva, Mapquest, Uniforest, Pagerduty, and Humany.

The image displays a collection of logos for various cloud and data technologies, organized into six categories:

- STORAGE**: Google Cloud, Microsoft Azure, Pure Storage, Alluxio, Nimbus Storage, Cloudera, and Cohesity.
- CLUSTER SVCS**: AWS, IBM, Docker, Mesosphere, and Core OS.
- APP DEV**: GitHub, Keen IO, and Rainforest.
- CROWD-SOURCING**: Amazon Mechanical Turk, Upwork, and Scale.
- HARDWARE**: Google TPU, ARM, Intel, Graphcore, Mythic, Nvidia, Movidius, Wave, and Hailo.
- GPU DBs**: Kinetica, Soream, and Blazeblock.

---

## ANALYTICS

**DATA ANALYST PLATFORMS**

- Microsoft
- pentaho
- alteryx
- Digital Reasoning
- gUAVUS
- AYASDI
- ATTIVO
- Datameer
- Quid
- incorta.
- interana.
- ClearStory
- Origami
- red5

**DATA SCIENCE PLATFORMS**

- IBM
- KNIME
- dataiku
- DOMINO
- rapidminer
- CONTINUUM ANALYTICS
- ALGORITHMIA
- DATAWATCH

The collage is organized into three main categories, each with a red header:

- BI PLATFORMS:** Includes logos for Microsoft, AWS, Domo, Veeva Analytics, Looker, ATSCALE, Alcatraz Data, IBM, Google Data, and Birst.
- VISUALIZATION:** Includes logos for Tableau, SAP, Google Cloud, Celonis, Qlik, Perceptics Data, ZEPL, Alteryx, CHARTIO, and Xoriant.
- MACHINE LEARNING:** Includes logos for AWS, Google Cloud, H2O, DataRobot, gamalio, ELEMENT, VIZENSE, Versive, and bonsai.

**COMPUTER VISION**

Microsoft Azure  
Amazon Rekognition  
Clarifai  
Ever AI  
TwentyOne

**HORIZONTAL AI**

IBM Watson Cortana, Face++ 人脸识别  
Sentient  
i) Affective  
Numeta  
narraglogics  
ALUS VISION

**SPEECH & NLP**

Google Cloud  
Amazon Alexa  
Semantic Machines  
MobiVox  
SoundHound Inc.  
MindHive  
Grindstone  
yycorp

SEARCH	LOG ANALYTICS	SOCIAL ANALYTICS	WEB / MOBILE / COMMERCE ANALYTICS
 elasticsearch  ORACLE  HADOOP  EXPLAIN  Lucidworks  swiftype  alphasense  omni-us	 splunk  sumologic  LOGGLY  TIMBER  kibana  logz.io	 Hootsuite  NETBASE  synthesio  semperache  bitly  SimilarWeb	 Google Analytics  mixpanel  sumall  RESCI  granify
 PRIVACYCLOUD  covoco  ATTIVO  algolia  MAANA  SINEQUA		 sprinklr  track  predata	 AMBITUS  aimaair  SIGOPT  custora

## APPLICATIONS – ENTERPRISE

[illegible]

## APPLICATIONS – INDUSTRY

[illegible][illegible]

## CROSS-INFRASTRUCTURE/ANALYTICS



OPEN SOURCE

A horizontal bar chart displaying various big data technologies organized into functional categories:

- FRAMEWORK**: Hadoop, MapReduce, Tez, YARN, Flink, Mesos, Spark.
- QUERY / DATA FLOW**: Spark SQL, Presto, SLAMMATA, Drill, Hive, Pig, Impala, Kudu.
- DATA ACCESS**: mifi, mongoDB, Cassandra, Redis, Couchbase, Aerospike, ScioDB, Riik, HBase, Hudi, Apache Ambari.
- COORDINATION**: talend, Talend Data Collector, Databricks, Cloudera, Storm.
- STREAMING**: Spark Streaming, Flink, Kafka, Flume, Druid, Storm.
- STAT TOOLS**: Python, ScalaLab, SciPy.
- AI / MACHINE LEARNING / DEEP LEARNING**: TensorFlow, Theano, PyTorch, Caffe, Microsoft Cognitive Toolkit, OpenAI, DM TK, IK, FeatureFusion, Veles, DMSUM, Neon™, DSSTNE, ml4d, MAHOUT, Aerosolve.
- SEARCH**: Elasticsearch, Solr.
- LOGGING & MONITORING**: ELK Stack (Elasticsearch, Logstash, Kibana), Prometheus.
- VISUALIZATION**: BeakerX, Rodeo.
- COLLABORATION**: Jupyter, Anaconda.
- SECURITY**: Apache Sentry, Knox, Sentry.

## DATA SOURCES & APIs

## DATA RESOURCES

**DATA SERVICES**

- Palantir
- OPERA
- DATA ENGINEERING
- fractal
- EXL
- kaggle
- DataKind

**INCUBATORS & SCHOOLS**

- PLURALSIGHT
- DataCamp
- INSIGHT
- galvanize
- DataElixir
- The Data Incubator

**RESEARCH**

- OpenAI
- facebook research
- MIRI
- MSL
- VECTOR INSTITUTE
- AI2
- AI
- AI2
- AI2



# Data Scientist: The Sexiest Job of the 21st Century





# Keep Calm with Michael Jordan

Artificial Intelligence—The Revolution Hasn't Happened Yet

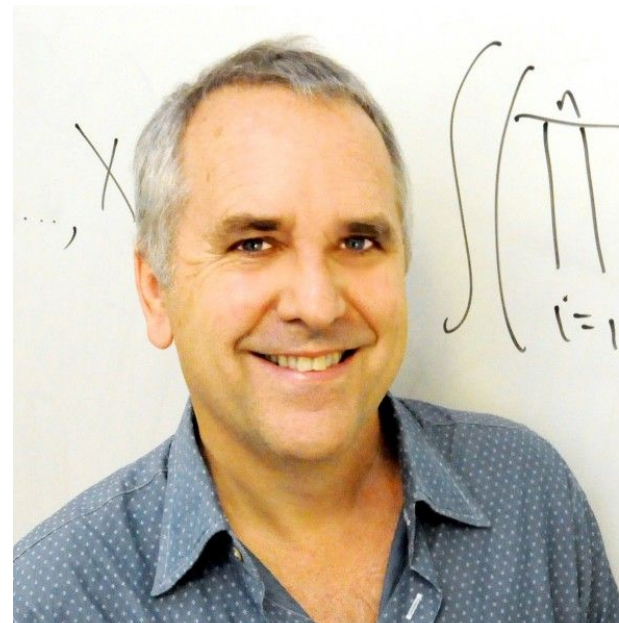
<https://medium.com/@mijordan3/artificial-intelligence-the-revolution-hasnt-happened-yet-5e1d5812e1e7>



# Keep Calm with Michael Jordan

## Artificial Intelligence—The Revolution Hasn't Happened Yet

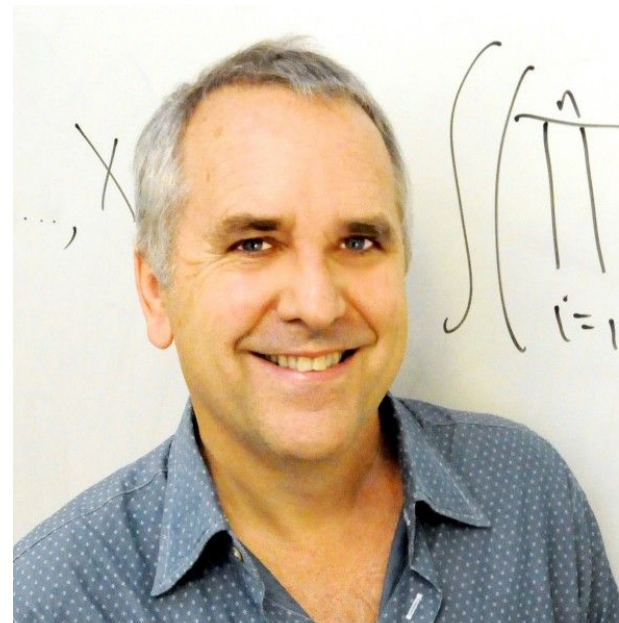
<https://medium.com/@mijordan3/artificial-intelligence-the-revolution-hasnt-happened-yet-5e1d5812e1e7>



# Keep Calm with Michael Jordan

## Artificial Intelligence—The Revolution Hasn't Happened Yet

- We are currently building Machine Learning blocks
- Blocks are being put together in ad-hoc ways
- We are far from what we call “AI”



<https://medium.com/@mijordan3/artificial-intelligence-the-revolution-hasnt-happened-yet-5e1d5812e1e7>

# Machine Learning Craftsmanship

Kelvin MOUTET

Machine Learning Engineer @ Prevision.io



# Software Craftsmanship

*As aspiring Software Craftsmen we are raising the bar of professional software development by practicing it and helping others learn the craft. Through this work we have come to value:*

*Not only working software ,*

*but also well-crafted software*

*Not only responding to change,*

*but also steadily adding value*

*Not only individuals and interactions,*

*but also a community of professionals*

*Not only customer collaboration,*

*but also productive partnerships*

*That is, in pursuit of the items on the left we have found the items on the right to be indispensable.*

<http://manifesto.softwarecraftsmanship.org/>

# PLAN

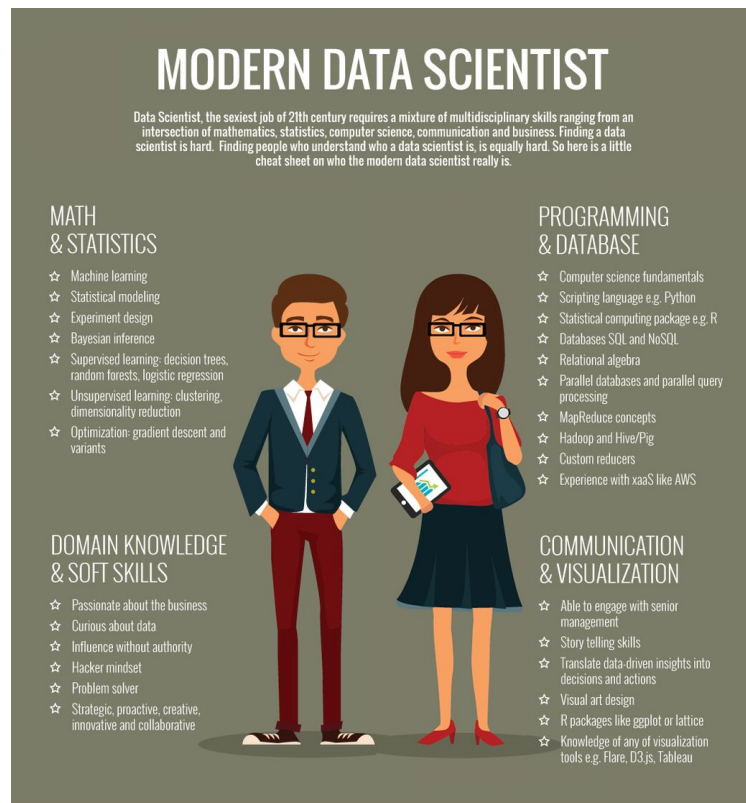
2 parts only !

# 1) Part 1



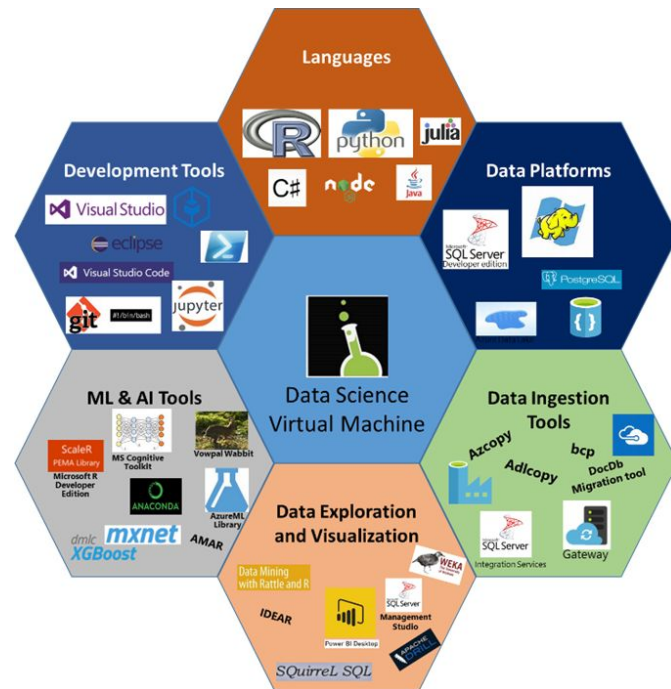
# Machine Learning Bare Necessities

## - Data Scientists !



# Machine Learning Bare Necessities

- Data Scientists
- Tools !



# Machine Learning Bare Necessities

- Data Scientists
- Tools
- Data !



# Machine Learning Bare Necessities

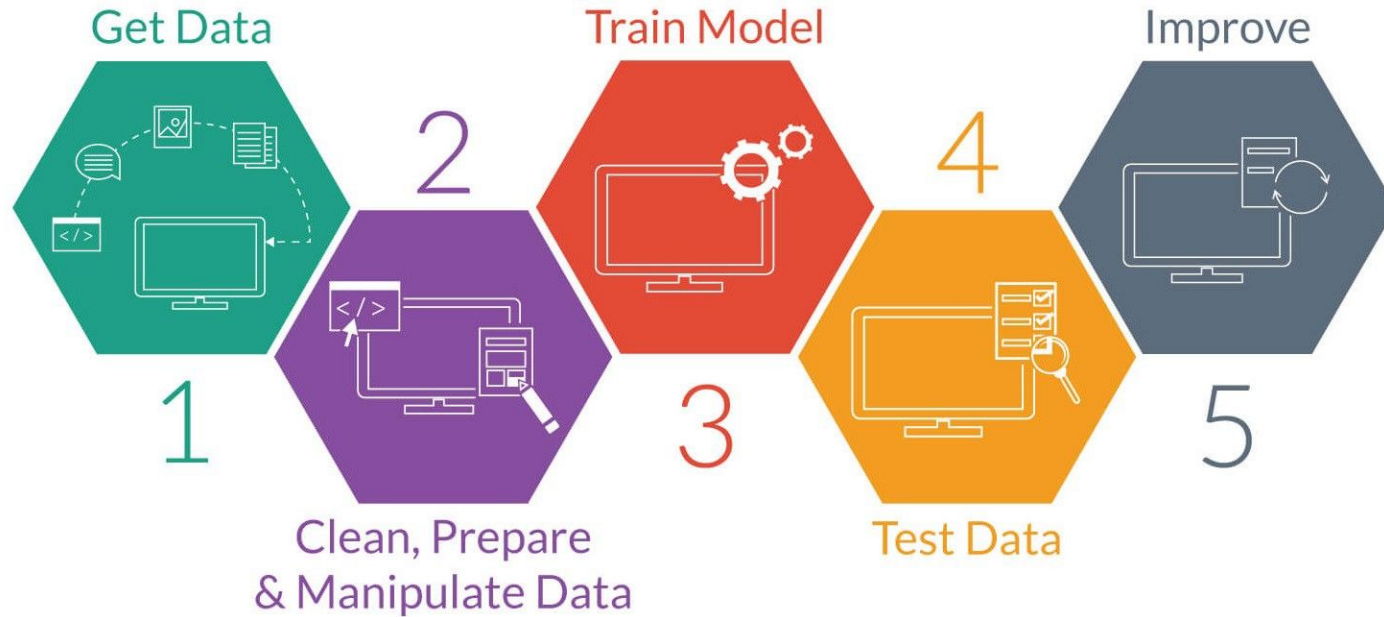
- Data Scientists
- Tools
- Data
- BIG DATA !





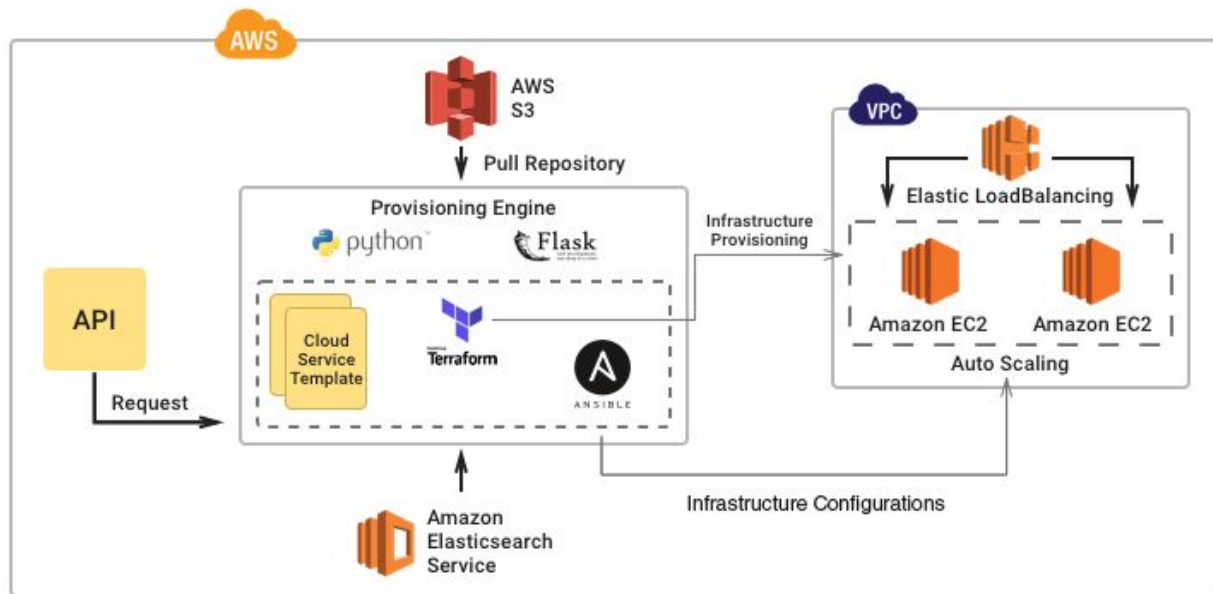
Let's go !

# Machine Learning core steps



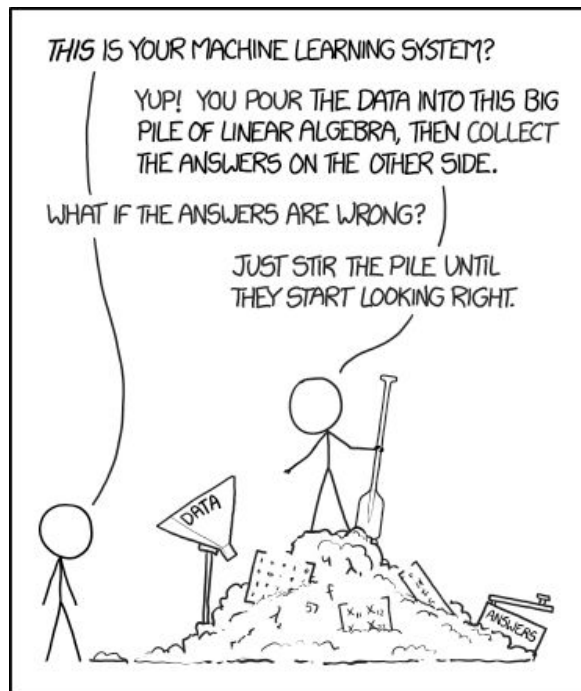
Go to Production !

# Machine Learning Infrastructure



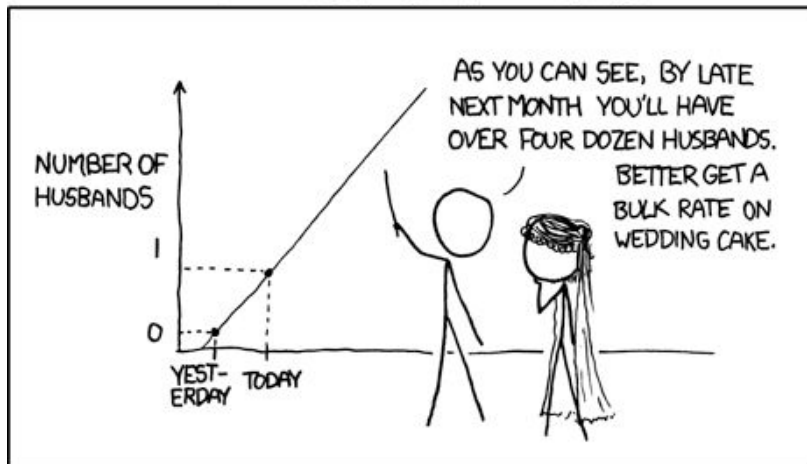


Welcome to AI ERA !!!



# Machine Learning “Success” Stories

MY HOBBY: EXTRAPOLATING



**Jac Rayner** @GirlFromBlupo

Dear Amazon, I bought a toilet seat because I needed one. Necessity, not desire. I do not collect them. I am not a toilet seat addict. No matter how temptingly you email me, I'm not going to think, oh go on then, just one more toilet seat, I'll treat myself.

Why does it fail ?

# Software project failure !



## The Top Six Reasons

- Incomplete requirements
- Lack of user involvement
- Lack of resources
- Unrealistic expectations
- Lack of executive support
- Changing requirements and specifications

[www.svsgit.com](http://www.svsgit.com)  
408-356-5793

# Machine Learning Project Failure (not exhaustive)

- Problem definition is bad or unclear (incomplete requirements)
- Scope is too large (unrealistic expectations)
- Overconfidence on data scientists (lack of resources)
- (Big) Data is too small or bad ! (lack of resources)
- True AI is not here (unrealistic expectations)
- Pessimistic on core business knowledge (lack of user involvement)
- Tools/ML Driven Development instead of Core Business Development
- Machine Learning algorithms understanding/coverage

~~1) Part 1~~

2) Part 2



## 2) Machine Learning Craftsmanship

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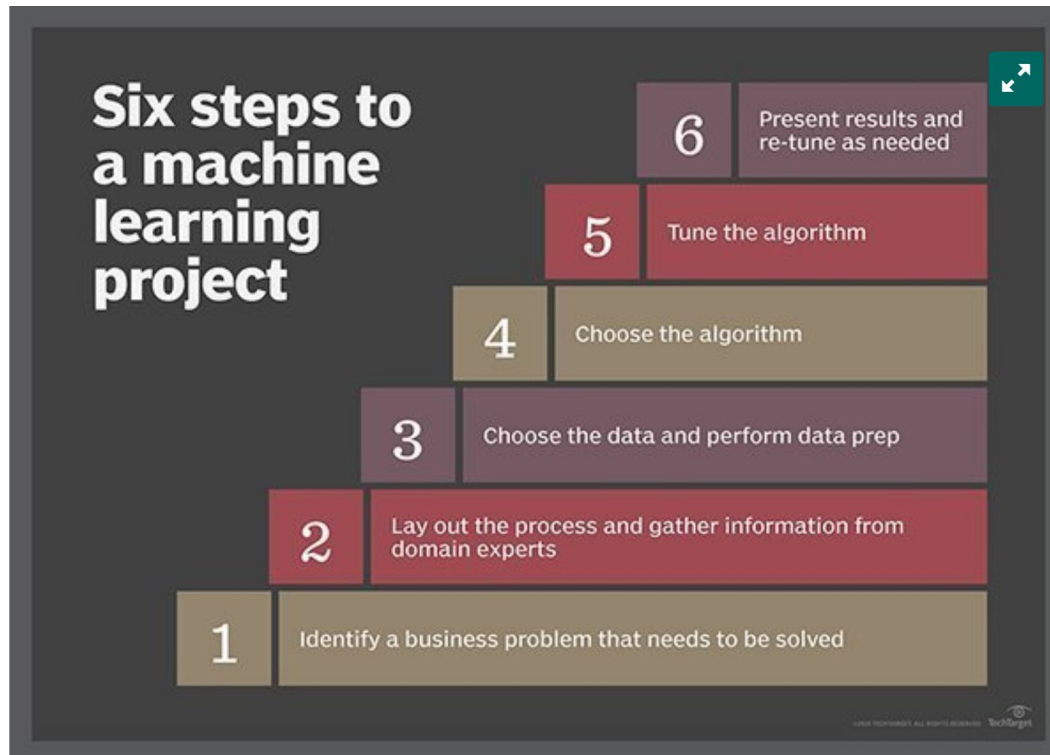
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# Machine Learning Craftsmanship

**Mindset !**

# A good way to start a Machine Learning project



# Examples !

# Machine Learning problem solving

Inspired by those two articles :

<https://towardsdatascience.com/1-year-doing-data-science-in-the-real-world-54f49b591991>

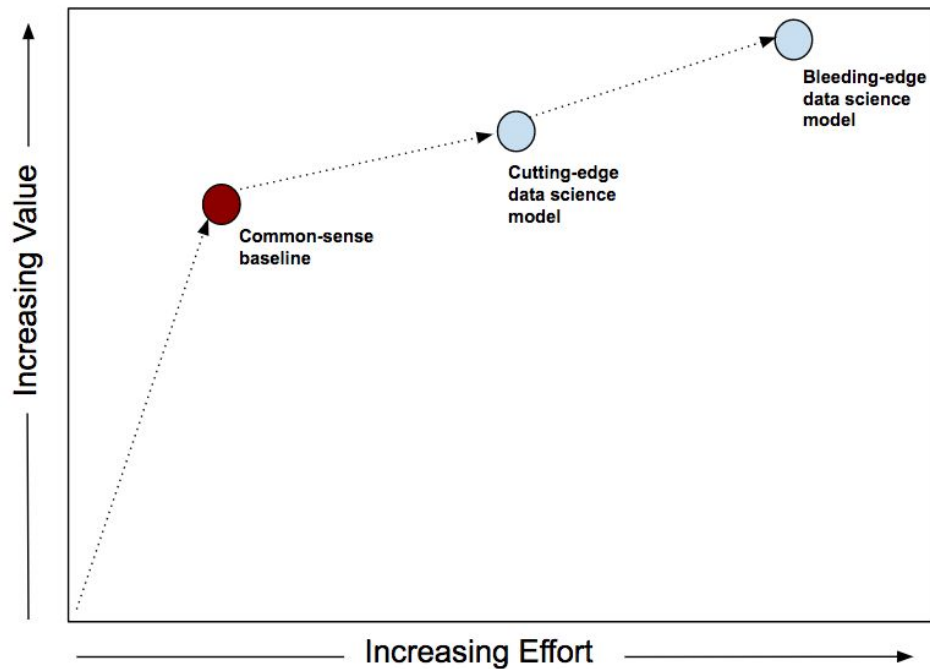
by **Jonny Brooks-Bartlett**, Data scientist Deliveroo

<https://towardsdatascience.com/first-create-a-common-sense-baseline-e66dbf8a8a47> by

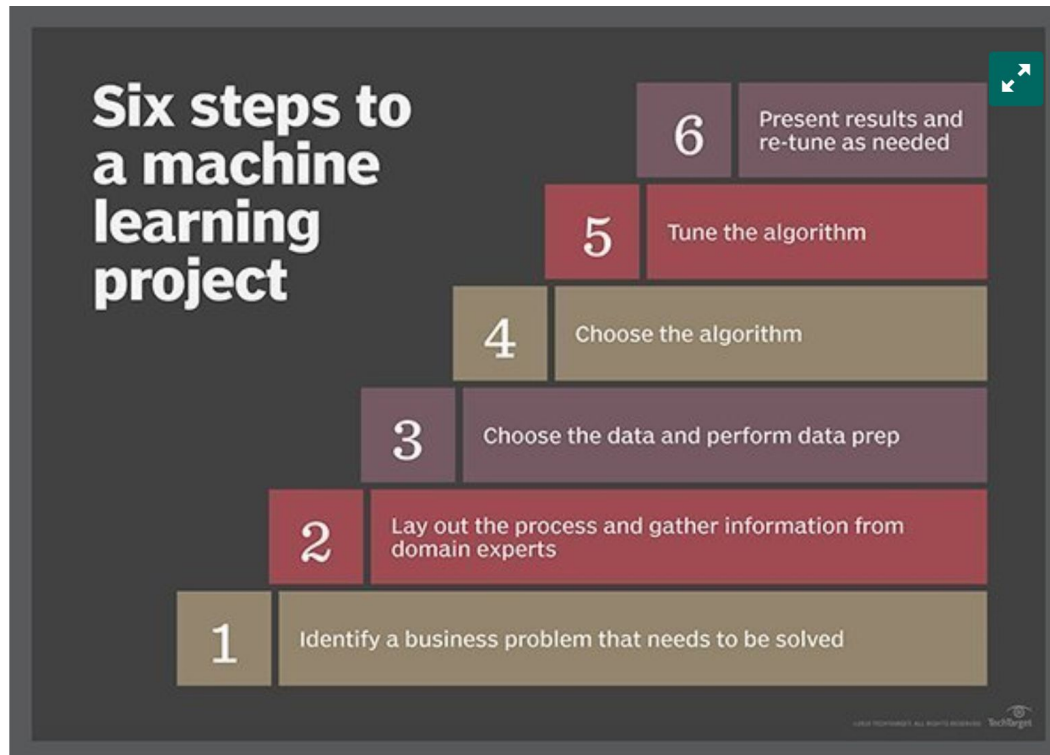
**Rama Ramakrishnan**, Senior Vice President, Salesforce.



# Identify your problem then build value !



# A good way to start a ML project



# Customers targeting

Send an advertising of your latest new products, at most, to 100,000 customers from the database.

How do you solve this ?



# Customers targeting

Pick the most **loyal** customers

- a) Shop a lot (frequency) and b) spends (money) a lot over a period (year, month)
- c) Recently shopped with you (number of day, week since last transaction)

Decile the customer databases, sort and pick the 100 000 first !

This is a core business tools called **Recency-Frequency-Monetary (RFM) heuristic**



# Product recommendations

Build a personalized product recommendation zone that will be displayed on the home page.

How do you do it ?



# Product recommendations

Transform the problem to : What's the simplest recommendation you can show visitors?

- Top selling products over all or by categories on a time window
- It works for all visitors (with or without informations)
- Personalisation with tweak (remember categories visited)

This is a good baseline before doing ML based methods



What next ?



# Machine Learning common issues !

# Machine Learning algorithms issues

[Data] Cold Start / Warm Start

[Data] Conceptual Drift / Covariate shift

[Model] No free lunch Theorem against you

[Data/Model] Reproduce

[Data/Model] Data Bias / Leak of information

[Data/Model] Cross validation and test validation

# Machine Learning project issues

[Model Selection] Usability (under core business constraints)

[Model selection] Core business metrics vs ML metrics !

[Model evaluation] A/B testing

[Production] Automatisation update mode

[Production] Monitoring (speed, failure, user feedback)

# Conclusion

- ❏ Talk with core business, as software engineers do, and build a common sense baseline !
- ❏ Fear over engineering and useless Machine Learning
- ❏ Keep learning :
  - Mooc, tutorials, books, articles, talks
  - Forum, blogs, twitter, people

# Machine Learning Craftsmanship

Kelvin MOUTET

Machine Learning Engineer @ Prevision.io

