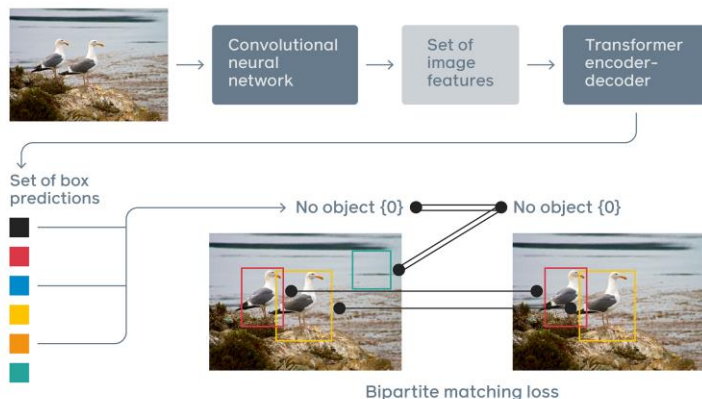




Human-Centered Data & AI

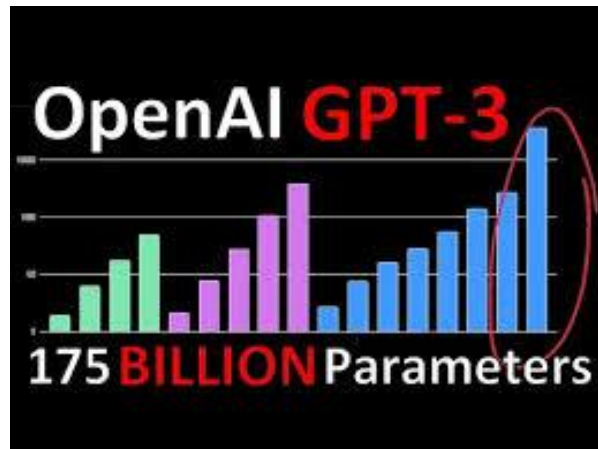
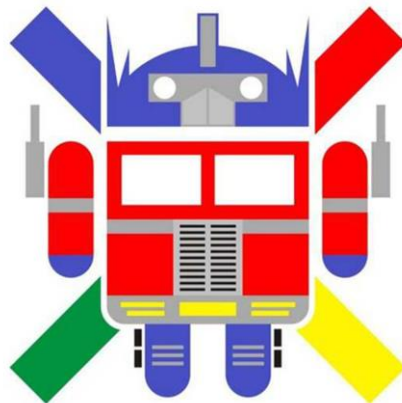
Pesquisas



Detection Transformers (DETR)

Facebook AI

<https://lnkd.in/ehqeAFe>



Paper: <https://t.co/EdFKftWRpv>

Video: <https://t.co/PmQ6PA5eWR>

Post: <https://lnkd.in/en4Ty2i>

<https://ai.googleblog.com/2017/08/transformer-novel-neural-network.html>

“

As ML matures from research to applied business solutions, so do we need to improve the maturity of its operation processes



**Launching is easy,
Operating is hard.**

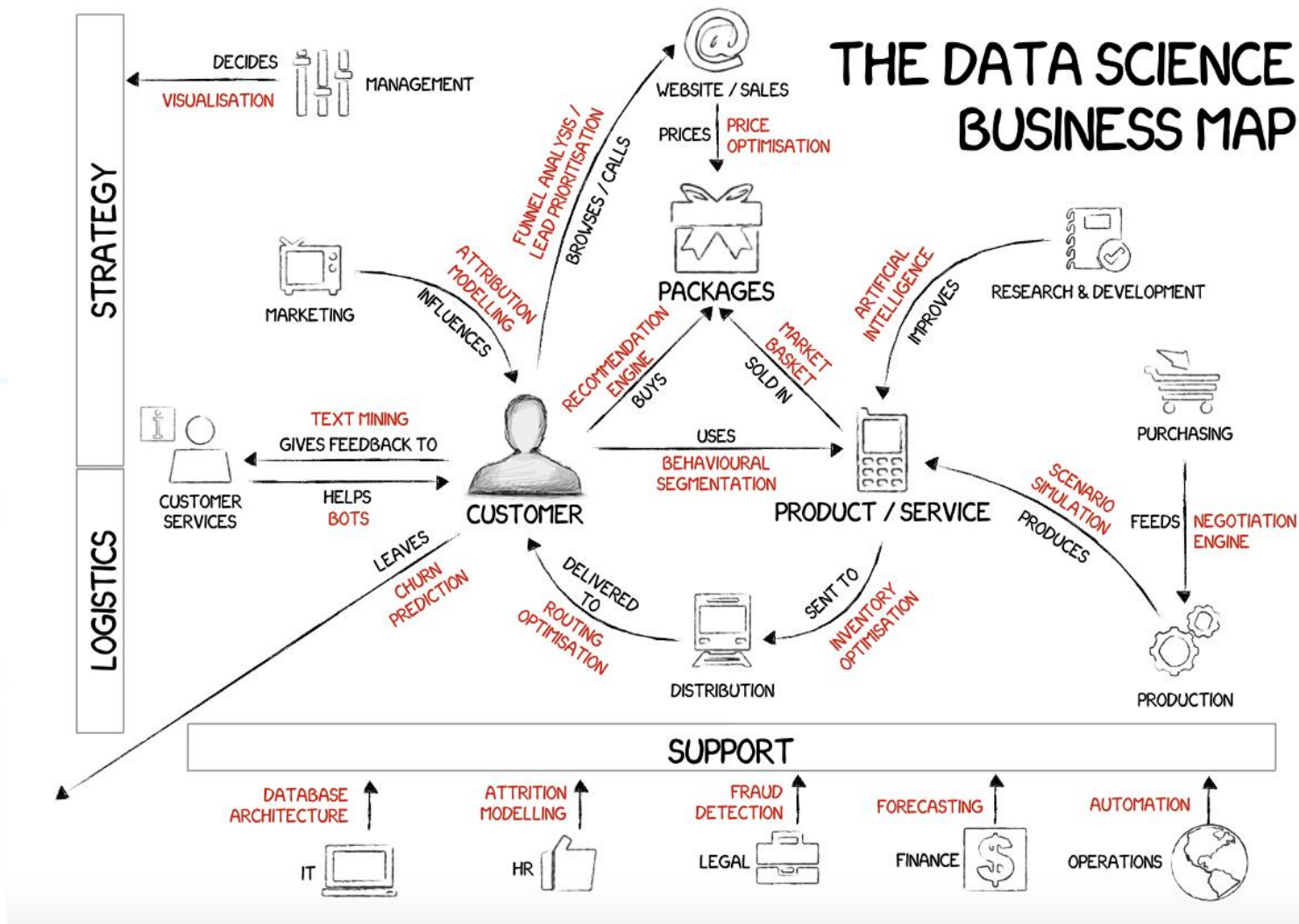
**"The real problems with a
ML system will be found
while you are continuously
operating it for the long term"**





How can we **transform** our business with Artificial Intelligence?

Machine learning é uma tecnologia central e transformadora que nos faz repensar como fazemos tudo “Sundar Pichai, CEO – Google”





Machine Learning | Produção

Além de treinar um modelo incrível...

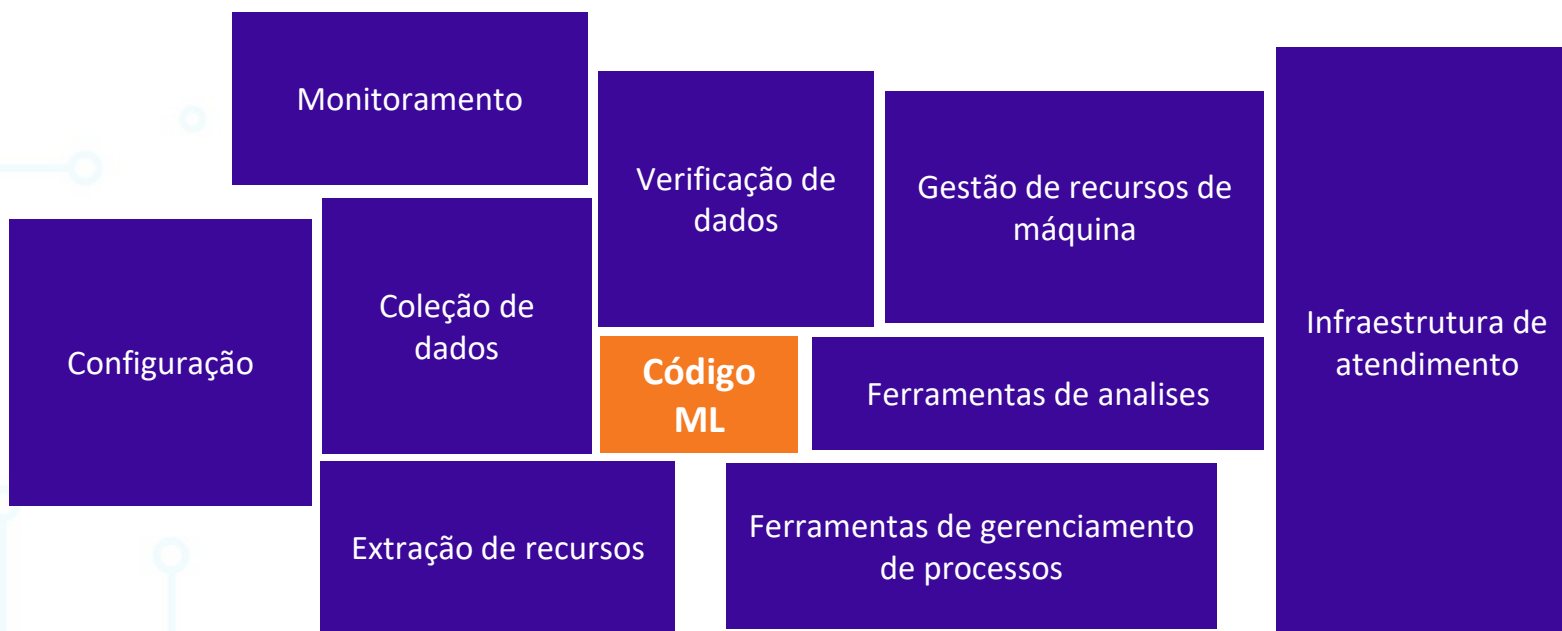
Código ML

Machine Learning | Produção



Machine Learning | Produção

Realidade: ML requer DevOps





Machine Learning | Produção

Modern Software Development

- Scalability
- Extensibility
- Configuration
- Consistency & Reproducibility
- Modularity
- Best Practices
- Testability
- Monitoring
- Safety & Security

Machine Learning | Produção

Modern Software Development

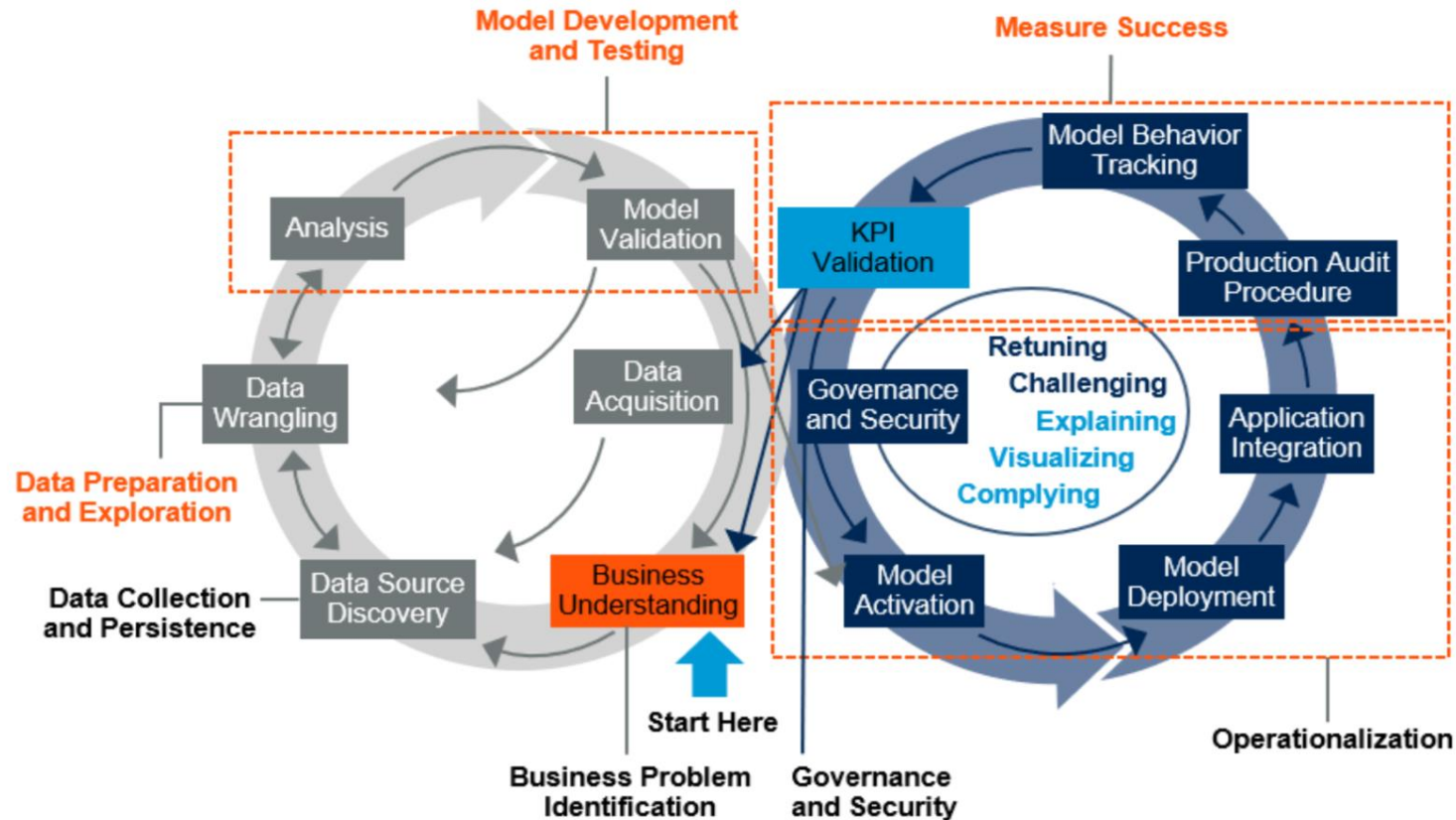
- Scalability
- Extensibility
- Configuration
- Consistency & Reproducibility
- Modularity
- Best Practices
- Testability
- Monitoring
- Safety & Security



Machine Learning Development

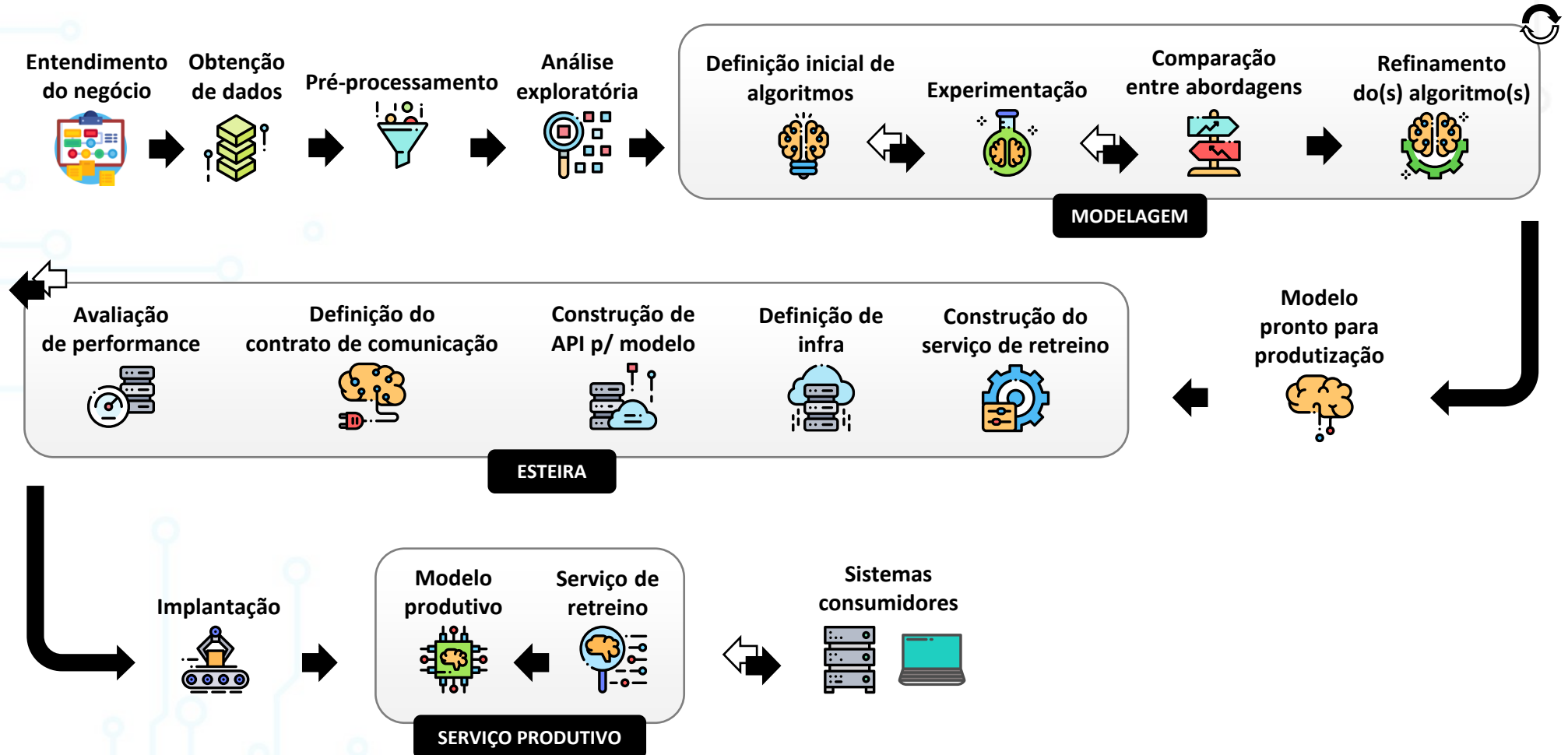
- Labeled data
- Feature space coverage
- Minimal dimensionality
- Maximum predictive data
- Fairness
- Rare conditions
- Data lifecycle management

Ciclo de vida de projetos de ML



Source: Gartner

Ciclo de vida de projetos de ML



Machine Learning

Freedom to experiment (and fail) is important

“Take your time
and succeed”



“Fail fast
and iterate”



Data Science Project Management says:

“Potentially Releasable Increment” Challenges: Scrum prescribes teams to deliver “potentially releasable Increment of ‘Done’ product” at the end of each sprint (Sutherland & Schwaber, 2017). However, Sutherland reports that “80% of the hundreds of Silicon Valley teams I polled could not do this” (Sutherland, 2015). This challenge is particularly daunting for data science teams and might not even be necessary. Much of the data science process, especially in the early exploratory phases, may not be intended for release outside the data science team. Moreover, testing requirements may extend beyond what is reasonable in a sprint, especially if a high degree of accuracy is needed (Jennings, 2017).



Processo atual mais usado para Ciência de Dados

Destaca as naturezas cíclicas e interativas da Ciência de Dados

Production Machine Learning

“Hidden Technical Debt in Machine Learning Systems”

NIPS 2015

<http://bit.ly/ml-techdebt>

Rules of Machine Learning: Best Practices for ML Engineering

Martin Zinkevich

<https://developers.google.com/machine-learning/guides/rules-of-ml/>

43 Rules

Rule #1: Don't be afraid to launch a product without machine learning.

Rule #23: You are not a typical end user.

Thanks !



Vinicius Fernandes Caridá

vfcarida@gmail.com



@vinicius caridá



@vinicius caridá



@vfcarida



@vfcarida