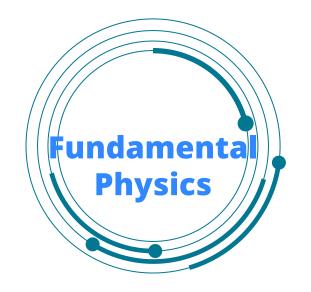
Data Science in



and the ridge to industry & society

Paul Van Branteghem

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/in/paulvanlorenzo



### About me

### **PAUL VAN BRANTEGHEM**

- Development team for the national AI strategy at SEDIA (Ministry of Economy) contracted by INECO
- Co-founder of Spain AI, a leading association in AI dissemination
- Freelance and CTO co-founder of Big Onion
- Physicist with a master's degree in advanced AI, meteorology, and a higher program in Digital Analytics and CRM
- Professor in various Data Science / Big Data programs (ISDI, CUNEF, EOI, INTEF)
- Experience at Siemens-Gamesa, in the AI department at BBVA, Deloitte Digital, and StratioBD

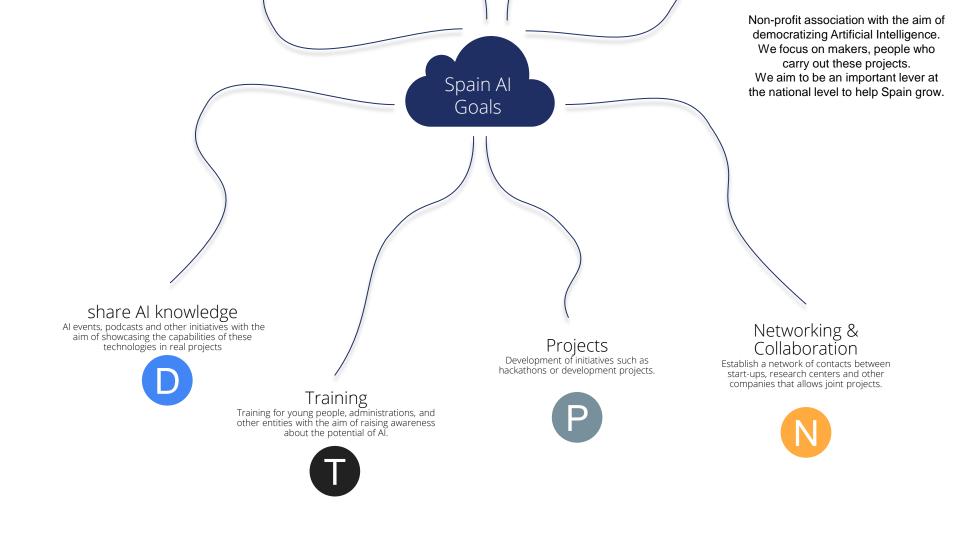








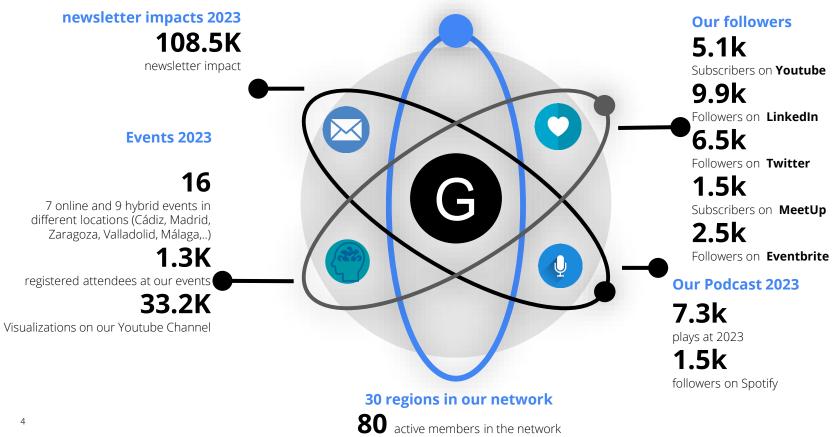






# About Spain Al

2023 Report



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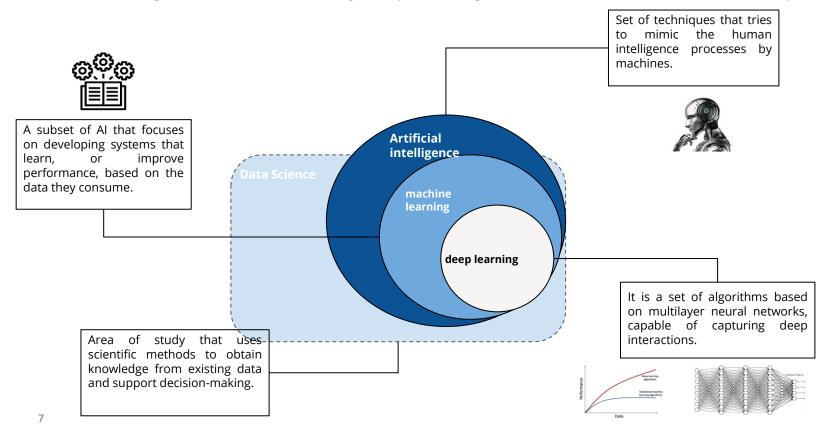
- Data Science
- ❖ Roles in Data Science
- From academia/research to industry

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- **❖** Data Science
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### Introduction to DS and AI

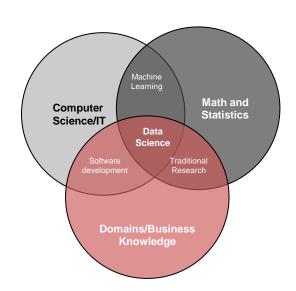
Artificial Intelligence, Machine Learning, Deep Learning and Data Science are different concepts.



### **Data Science**

When we talk about Data Science, we talk about a project typology that mixes 3 main domains: computer science, business knowledge and statistics.

- Computer Science: Use computer science techniques and tools to create algorithms, find patterns, ask questions, and launch experiments..
- Business Knowledge: Know what is the main objective,
   limitations, data and requirements of the project.
- Statistics: Define the best data transformations, cleaning, understand how the model works.



# **Roles Data Science Projects**



There is generally no Data Scientist who knows everything perfectly; Usually, these projects require multiple profiles.

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# **Roles Data Science Projects**

Within a DS project we can find multiple roles.



Data architect

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**Objective:** architects Data conceptualize and visualize data flow frameworks.

Background: Computer science, programming. It is usually a more experienced DE.

**Tools:** Scala, JAVA, MongoDB, Cloud Architectures.



Learning Engineer



Machine

Objective: Thev take experimentation models and convert them into productive models to deploy.

**Background:** Computer Science with knowledge of ML

Tools: Python, Spark, Scala, Tensorflow....



Product owner/ Manager

**Objective:** Define and manage business requirements.

Background: Business profile with high level analytical and development knowledge.

Tools: IIRA, Trello, Excels.



Obiective: Data engineers develop and maintain data frameworks.

Background: Computer science, programming.

**Tools:** Scala, IAVA, MongoDB,

Cloud Architectures. Data engineer



**Data Scientist** Research

**Objective:** Experiment with the data to process and analyze the best models.

**Background:** Statistics, physics, mathematics, engineering.

Tools: lupyter Notebooks, Python, R, VIsualization Tools (Tableau,...)



Data ss Role **Objective:** Understand and convert business requirements into data.

Background: ADE. finance, actuarial,...

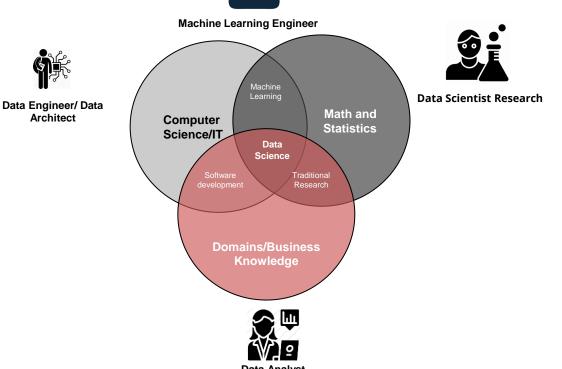
Tools: Excel. Oracle. Visualization Tools **Analyst/Busine** (Tableau, Microstrategy...)

computer science **Business** 

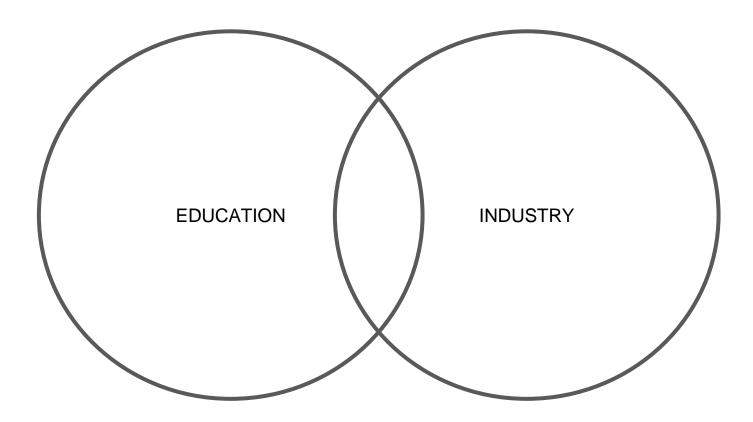
# **Roles Data Science Projects**

Each of the roles may have a greater importance in the project depending on the scope and

needs at each point of the project.



# My experience moving from university to industry



# My experience moving from university to industry as a physicist

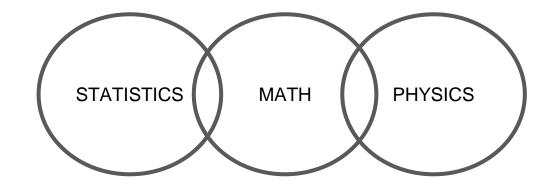
Most profiles coming from physics, mathematics, or statistics are closer to the role of Data Scientist Research.



Data Scientist Research **Objective:** Experiment with the data to process and analyze the best models.

**Background:** Statistics, physics, mathematics, engineering.

**Tools:** Jupyter Notebooks, Python, R, VIsualization Tools (Tableau....)



- Basic Programming Skills
- Strong Statistical Foundation
- Business Knowledge Gap

# My experience moving from university to industry as a physicist/meteorologist

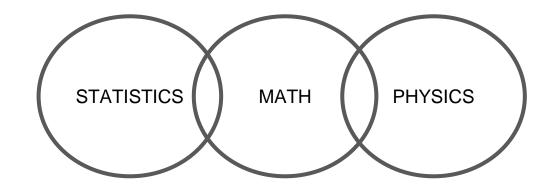
Most profiles coming from physics, mathematics, or statistics are closer to the role of Data Scientist Research.



Data Scientist Research **Objective:** Experiment with the data to process and analyze the best models.

**Background:** Statistics, physics, mathematics, engineering.

**Tools:** Jupyter Notebooks, Python, R, VIsualization Tools (Tableau....)



In my case, I was lucky. I learned to program and understand computers better through the use of CESGA in developing weather prediction models. However, I still lacked many good coding practices and the use of collaborative tools.

A good foundation in time series modeling

- Basic Programming Skills
- Strong Statistical Foundation
- Business Knowledge Gap

# What skills did I find I needed to improve when I started working in the industry?

When I started working in the industry, I realized there were certain skills/goals that I needed to work on to improve my career path.

Model Metrics vs Performance Metrics





Clean Code & best practices



**Soft Skills** 



focus on production



First, understand the business



**Networking** 



Learn to adapt

# **Example: Demand forecasting Model for a top retailer**

Daily Demand Forecasting by SKU

- + 1000 shops
- + 1000 products per shop

~1M models in production for daily demand forecasting

- R didn't work
- ARIMA didn't work
- RMSE, MAE or other regression metrics are not the main metric that should be taken into account
- RMSE, MAE, or other regression metrics are not the main metrics that should be taken into account.
- First, consider the production requirements, and then try to find the best model.
- Huge team-> good and clean coding skills and collaborative tools

# Important initiatives that have helped me in my transition from being a student to my professional career





Master's thesis enterprise collaboration

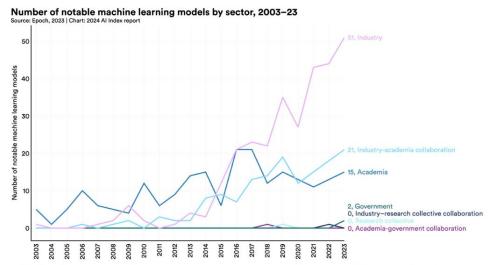


self-learning & continuous learning (Moocs)



Workshops/Even ts/networking/Ha ckathons

# You can do research in Industry



In 2023, industry produced 51 notable machine learning models, while academia contributed only 15. There were also 21 notable models resulting from industry-academia collaborations in 2023, a new high.

### Al Index Report 2024 By Stanford University











# Developing LLMs has a high cost that is affecting academic research in this field



Fei-Fei Li says Stanford's Natural Language computing lab has only 64 GPUs and academia is "falling off a cliff" relative to industry



1111

With President Biden one of the things we talked about is the moonshot mentality to invest in public sector AI ...

America's public sector academia is falling on the clift very fast in terms of Al resources.""

May 2024, Fei Fei Li, Stanford professor

# Tankyou. Questions?



