

# eleven super case data challenge

To the attention of the master in Data Science &  
Business analytics' students

February 8<sup>th</sup> , 2021

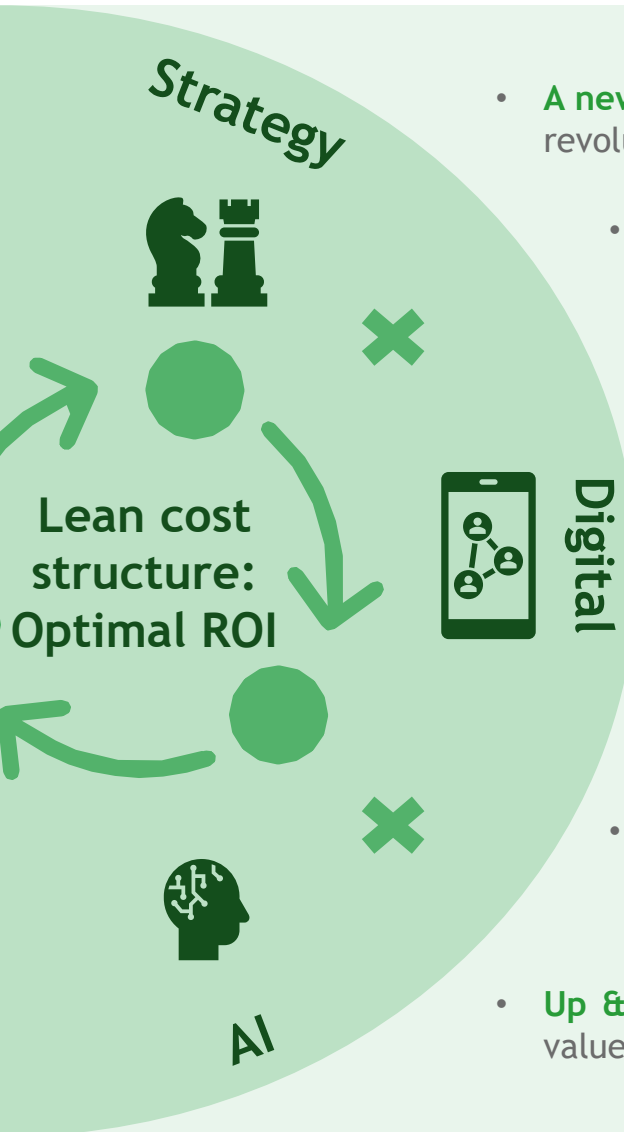


# AGENDA



1. About eleven
2. Case presentations
  - a) Airplane Interior Service
  - b) Aircrafts' take-off time prediction
3. Practical information
4. A few advice

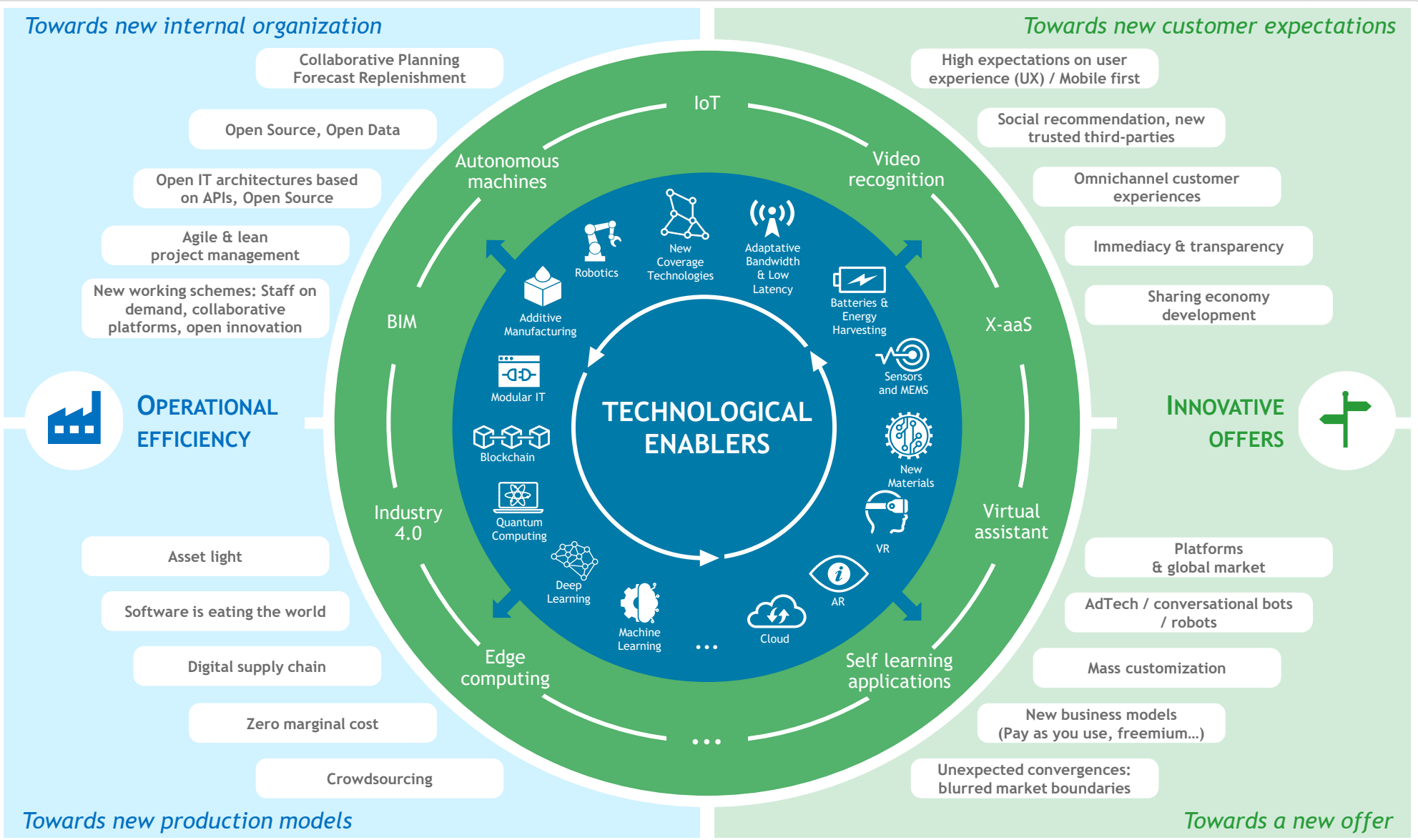
eleven is Europe's first ever specialist strategy firm specifically founded to accompany clients' transformation through the data and digital revolution, thanks to a unique combination of strategy perspective, entrepreneurship approach and a strong proximity with data and digital ecosystem



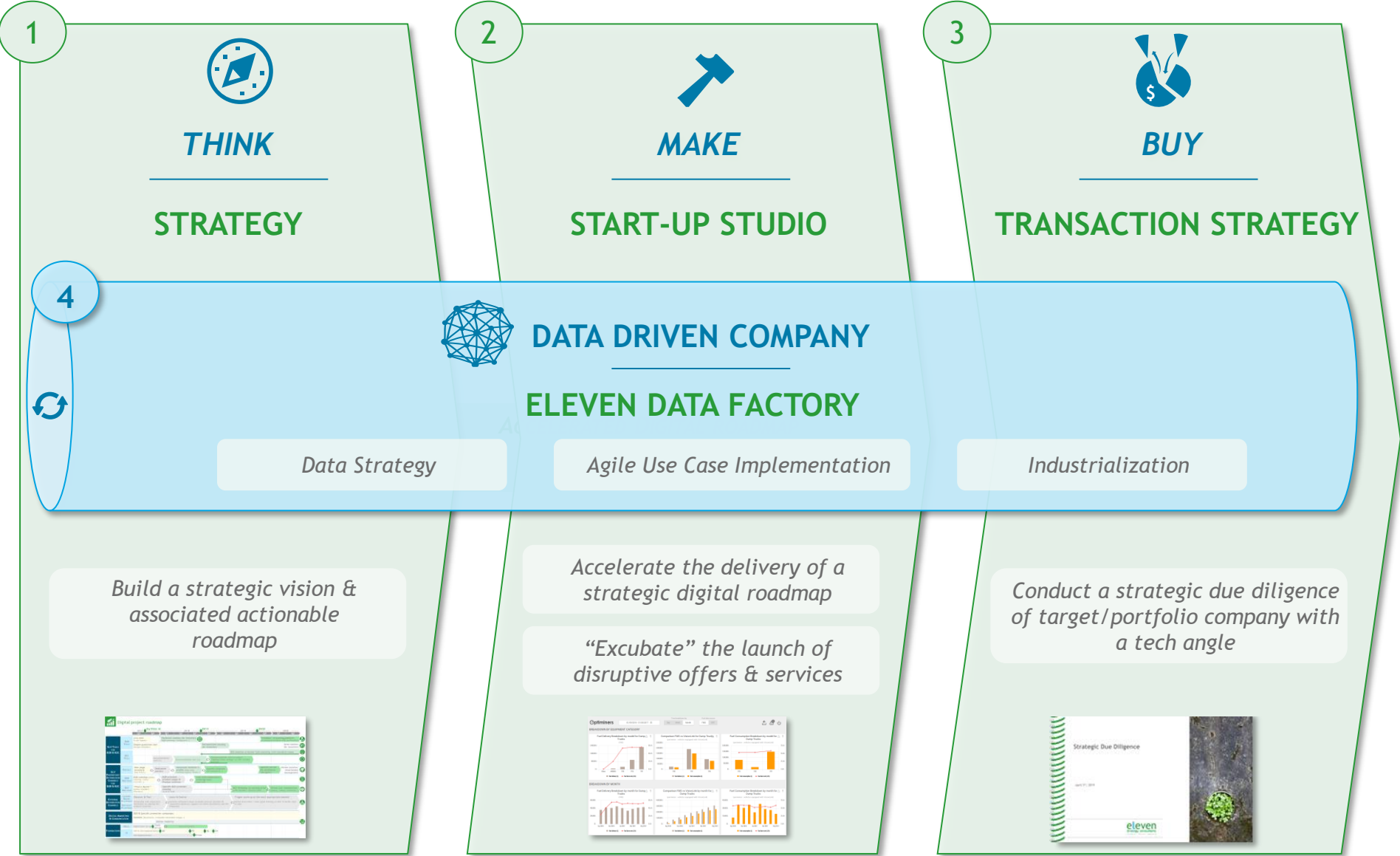
- **A new paradigm:** In today's world, immediately following and leveraging the ongoing digital revolution, the AI revolution is unlocking new, untapped value-creation opportunities
- **Our unique positioning:** eleven has been founded to specifically address strategic challenges raised by both digital and data, supporting C-level executives and organizations from strategic ignition to project scale-up and industrialization
  - **A unique blend of skills:** As strategic issues around digital raised more and more data challenges, eleven's unique expertise lied in its capacity to blend strategic problem-solving skills with tech and data science ones, thanks to its 50+ consultants that master the continuously evolving technological enablers at stake
  - **Distinctive entrepreneurial mindset and business model:** Our deeply-rooted nimble and entrepreneurial culture, combined with a lean cost structure enables us to offer optimal and demonstrated ROI for our clients' projects
- **CSR at the core of eleven's DNA:** Our approach enables our clients' projects to meet CSR expectations. Both financial and CSR impacts are tracked, proven and reported thanks to our mastering of digital and data levers
- **Up & ready for the new challenge:** We believe firms better equipped with data & digital value-levers are the best prepared to tackle the extended ongoing sanitary crisis era

Our projects are built around ongoing major digital & data disruptions, including **new modes of production, consumption and collaboration**, which represent major challenges for existing players

*eleven's 'plate tectonics' of disruptions*



eleven addresses the entire spectrum of strategic topics around digitalization and disruption with its four core offers



eleven serves top large cap clients and leading mid cap players across several key industries as well as leading International large cap and smid cap Private Equity funds

ELEVEN'S KEY REFERENCES

CORPORATE

CAC 40

VINCI ENGIE SNCF

SOCIETE GENERALE AXA LVMH

ACCOR orange TOTAL

PRIVATE EQUITY

Large cap

BainCapital THE CARLYLE GROUP

ARDIAN VITRUVIAN PARTNERS

eurazeo Bridgepoint

Mid Cap players

sanef INDIGO ALCATEL onetouch. smart move.

FONCIA Club Med SERVAIR

euräus Van Cleef & Arpels sonepar

SMid cap

CATHAYCAPITAL Private Equity naxicap PARTNERS

mbo partenaires l'entrepreneur est capital ABENEX

andera PARTNERS 21 Partners

Our **partners** bring a **complementary range of expertise** to the table, ensuring to address all the dimensions of digital and data transformation



### MAXIME CARO

- Maxime joined eleven at its foundation after a first consulting experience in the United States. At eleven, Maxime has been working on projects related to software, data science, and to the launch of new businesses for both corporates & private equity funds.
- Academics: Mines de Paris, National University of Singapore



### AMBROISE HURET

- Ambroise began his career as a strategy consultant within Booz.Allen & Hamilton's Telecommunication Media and Technology practice. He co-founded several start-ups with successful exits to Monster.com, Dassault Systems and Bearing Point. Ambroise also teaches strategy at both the HEC Paris MBA and the HEC Paris MSc Strategic Management and is a professor on Coursera.
- Academics: HEC Paris, Singularity University



### CHRISTOPHER RISCHARD

- After 10 years in the United States selling internet solutions, Christopher spent 8 years at Booz&Co in Paris and later as a Principal in Madrid, focusing on digitalization strategy in EMEA. Based in London since 2014, Christopher continues to deliver commercial due diligence and disruptive digital strategy work across industries with an emphasis on private equity clients.
- Academics: ESSEC Paris, INSEAD MBA



### BERTRAND SEMAILLE

- Bertrand began his career in the Media and Entertainment field before joining the strategy practice of Bossard Consultants in Paris. He then founded and led the consulting team of the Cap Gemini affiliate dedicated to digital strategy. Bertrand also teaches digital strategy at the HEC Paris MBA.
- Academics: Sciences Po Paris, Pantheon Sorbonne - MSc Econometrics



### MORAND STUDER

- Morand has been working as a consultant for over a decade and boasts specific expertise the Artificial Intelligence field... gained through several cutting-edge projects he has been leading for global industry majors. Morand heads eleven Data Factory and teaches at the HEC Paris MSc Strategic Management.
- Academics: Ecole Polytechnique Paris, ENSAE, Sciences Po Paris, Singularity University



### STEVAN URIEN

- Stevan previously worked for several private equity investors in new technologies, where he led due diligences each time a high level of technical expertise was required. He is involved in numerous assignments alongside innovation divisions of corporate companies, accelerating time-to-market of innovations.
- Academics: Ecole Polytechnique Paris, ENSTA

Our data scientist consultants combine both **unique technical and functional skills** shaping the backbone of eleven unique **strategy-driven data science**



## Data Science

*eleven's consultants master most of Data Science techniques (supervised, non-supervised, deep learning, etc.) that enable them to quickly identify data challenges of companies in order to implement tools for **visualization, modelling, prediction** and **recommendation** aiming at **generating value through data***



## Hacking and industrialization

*Mastering the majority of Data tools (R, Python, Spark, Docker, Elasticsearch, TensorFlow ...) allows eleven consultants to **quickly adapt to the data environments** of each customer*



## Strategy

*eleven's consultants are able to **understand their clients' core business**, to focus on the highest stakes and to frame the problem optimally*



## Transformation

*eleven's consultants are able to adapt to various stakeholders and to demonstrate the value of the implemented initiatives in order to **generate the cultural changes** necessary to make the **transformation sustainable over the long term***



eleven leverages its **academic and R&D efforts** to constantly keep an edge in terms of **digital and AI technologies**

eleven's academic exposure to the talents at the cutting edge of digital, data, and technological evolutions



HEC  
PARIS



Organization of the MSc Data Science for Business Hackathon



CentraleSupélec

ESSEC  
BUSINESS SCHOOL



Organization of the MSc Data Science & Business Analytics Hackathon

eleven's differentiative R&D efforts



- ✓ Several consultants holding **PhD**
- ✓ Numerous **highly skilled engineers**

- ✓ R&D efforts deployed on the **cutting edge of innovation**: model explicability, computer vision (video), ...

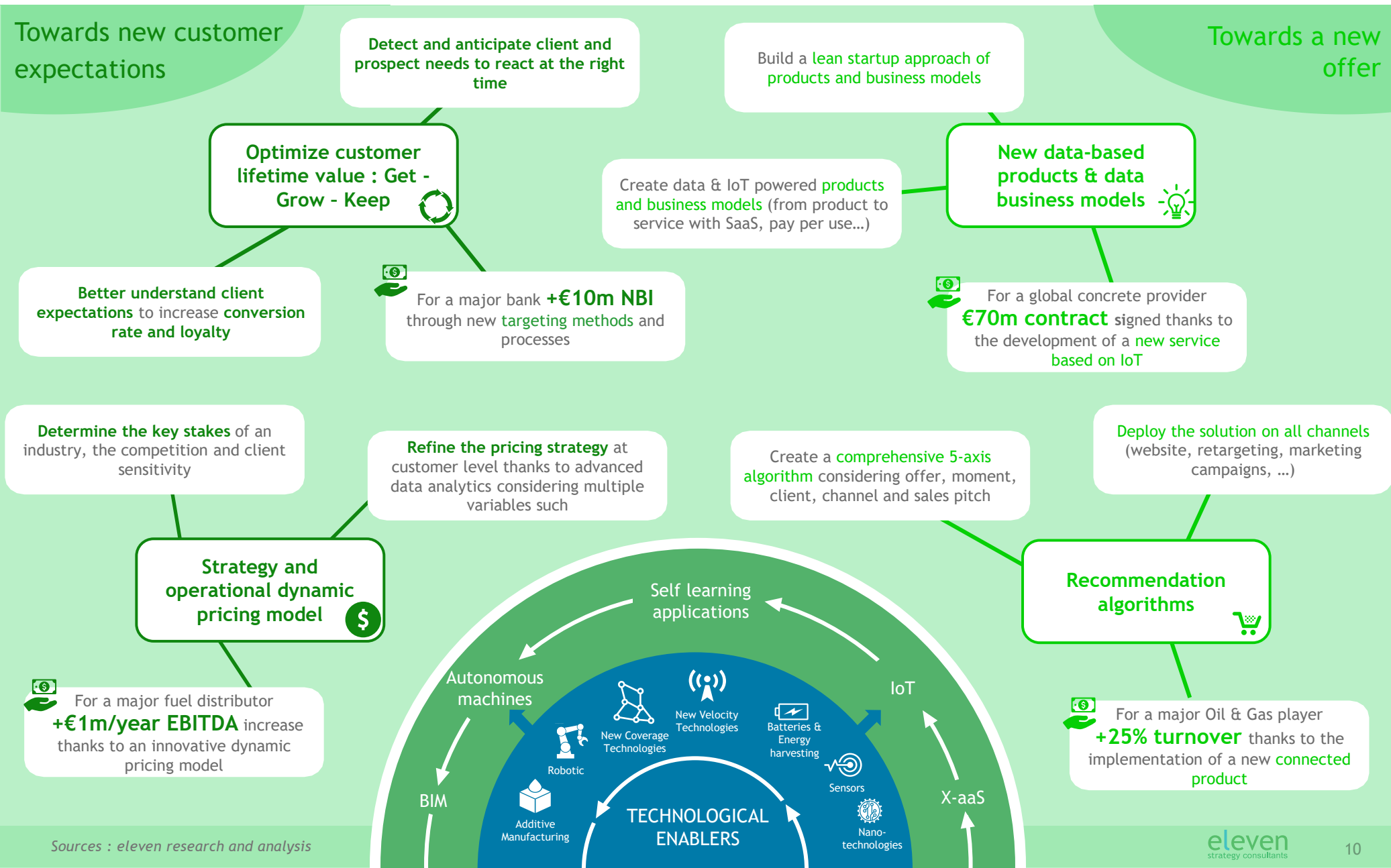


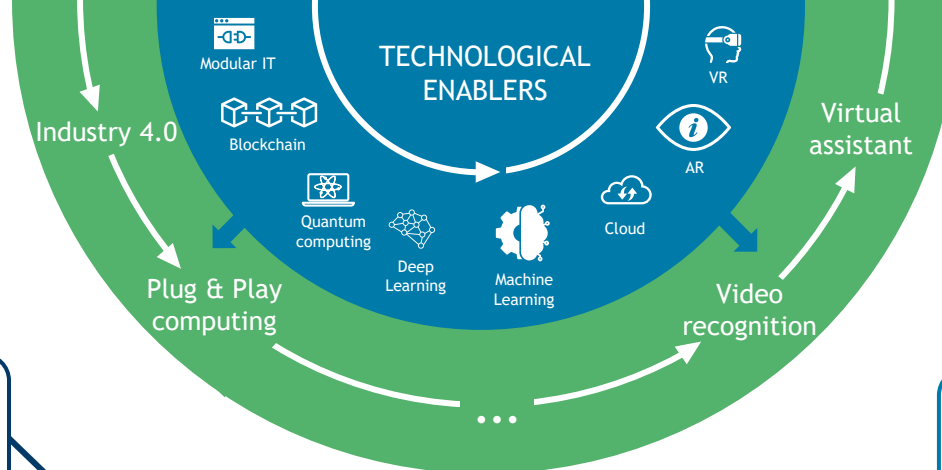
- ✓ **1200+ man-days** per year dedicated to R&D efforts

- ✓ A portfolio of **100+ trainings** around digital and data trends and competencies



eleven data factory covers **all disruptions** brought by the **data revolution** from customer expectation to internal production models and organization





Design & optimize processes leveraging a blend of **NLP, Computer vision, scoring, sentiment analysis**,...

### AI-powered business processes automation



Integrate and industrialize AI within production chains to **anticipate and forecast** potential issues



For a financial services player  
**200 man-day/year** shrink bankers' time dedicated to customer analysis

Optimize industrial processes thanks to **real-time information** provided by IoT sensors

### Factory 4.0 and predictive maintenance



Shift from reactive and preventive to predictive maintenance processes leveraging **anomaly detection** & existing **in-house expertise**



For a major oil & gas company  
**+€5m/year**  
Of savings per mine thanks to the reduction of fuel consumption

Build practical and **actionable data roadmaps**, from the first Proof of Concept to the progressive **industrialization** of selected use cases

Issue recommendations on **data architecture, organization, governance, technical providers and tools**

### Data governance and organization strategy



For a videosurveillance player, **2 new business units created** on data-based new business models

Identify the relevant **key performance drivers** among hundreds of sources in order to **streamline operations**

Identify optimal **machine setups and processes** to increase production

### Performance assessment & optimization



For a food industry player  
**+€1m/year** cost savings thanks to quality assessment process improvement

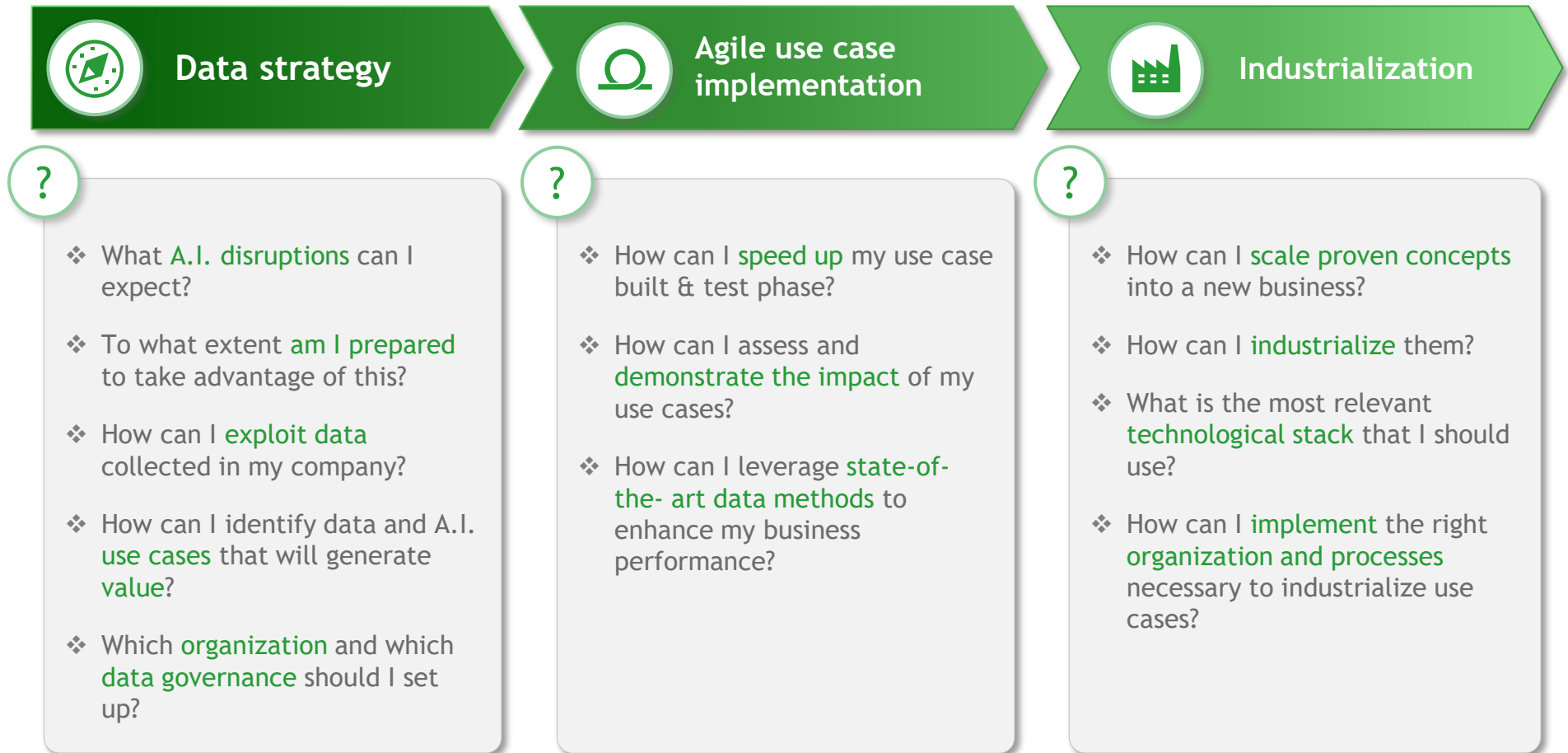
Towards new internal organization

Towards new production models<sup>11</sup>

# eleven data factory three complementary offers address the full spectrum of data-related needs from strategy to use case industrialization



## ELEVEN DATA FACTORY KEY EXPERTISE



# AGENDA



1. About eleven
2. Case presentations
  - a) Airplane Interior Service
  - b) Aircrafts' take-off time prediction
3. Practical information
4. A few advice

# Supercase 1

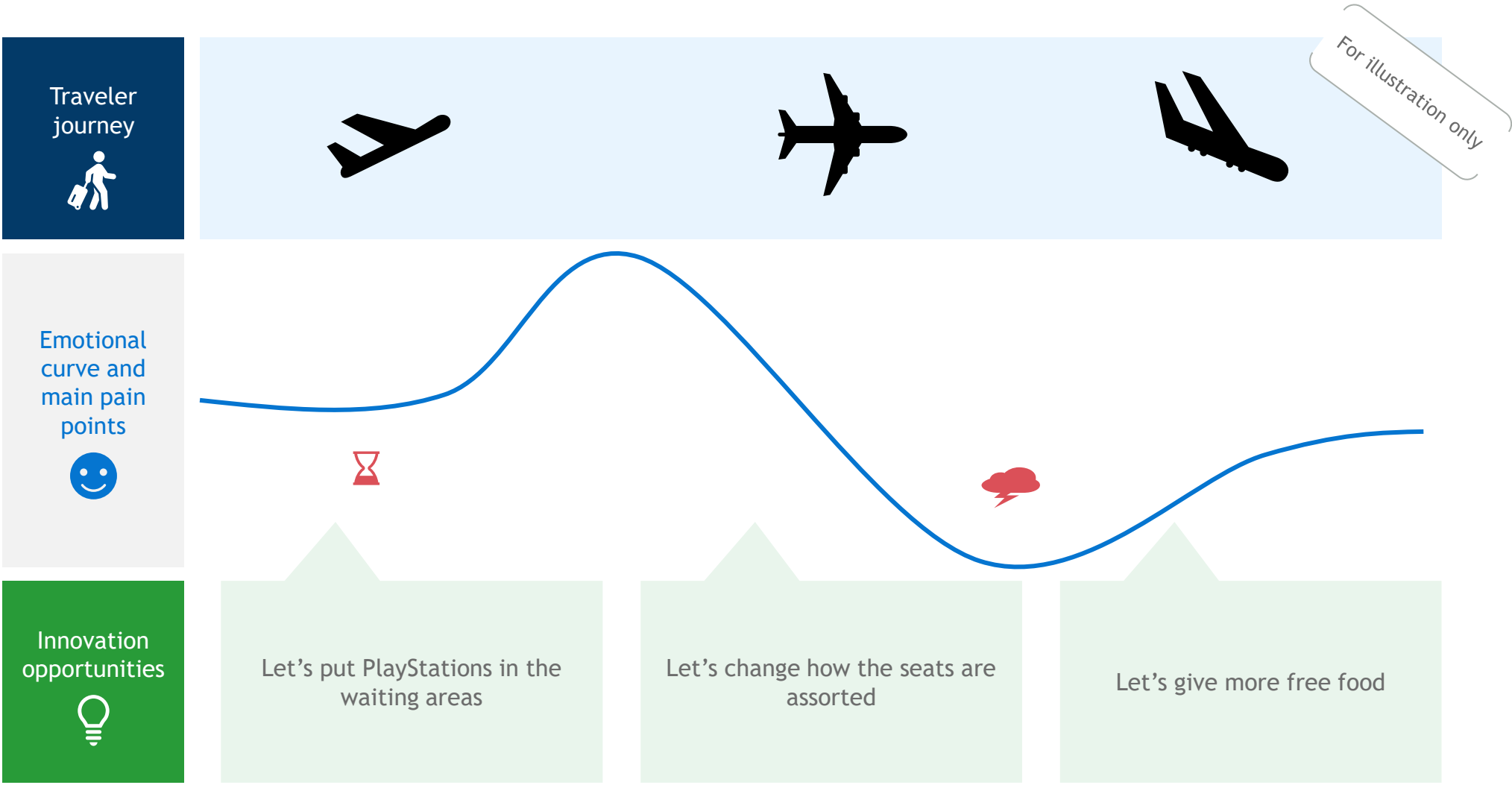
# Airplane Interior Services

*To the attention of the master in Data Science & Business analytics' students*

February 8<sup>th</sup> , 2021



The client, a player in the aeronautic industry, wants to understand how the user experience of airplane passengers can be improved





Several websites enable users to share their experience, thereby providing **valuable data sources** with large and various information



1/10

**"a horrible airline to fly"**  
Rachel Beale (United States) 28th July 2019  
✔ **Trip Verified** | Atlanta to Paris. Air France is just a horrible airline to fly, especially compared to delta. They do not update their movies on their tiny screen they put in front of you it's the same movie names repeated over and over to make it seem like there are choices. I found the food disgusting and they have no WiFi on the plane. I would not recommend flying Air France at all unfortunately due to work and my routes I've had to use them 4 times.



Type Of Traveller	Family Leisure
Seat Type	Economy Class
Route	Atlanta to Paris
Date Flown	July 2019
Seat Comfort	★ ★ ☆ ☆ ☆
Cabin Staff Service	★ ★ ☆ ☆ ☆
Food & Beverages	★ ★ ☆ ☆ ☆
Inflight Entertainment	★ ☆ ☆ ☆ ☆
Ground Service	★ ☆ ☆ ☆ ☆
Value For Money	★ ☆ ☆ ☆ ☆
Recommended	✗



**Baloo** wrote a review Sep 9  
2 contributions

...

●●●●○

Moscow - Ulaanbaatar International Economy

**Delay**  
After getting inside the aircraft we were waiting about 2 hours on board, then they announced that they are gonna change our schedule. Because they have to change the aircraft. That's all somehow ok, but the worst part of the story is, that the pilot tried to explain what is...  
[Read more](#)

**Date of travel:** August 2019

Helpful

Share

**Praveen Chaudhary** wrote a review Sep 9  
Little India, Singapore • 44 contributions • 7 helpful votes

...

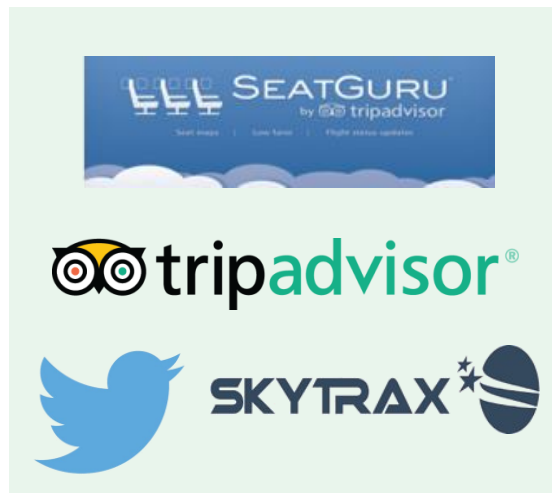
●●●●○

New Delhi - Moscow International Business Class

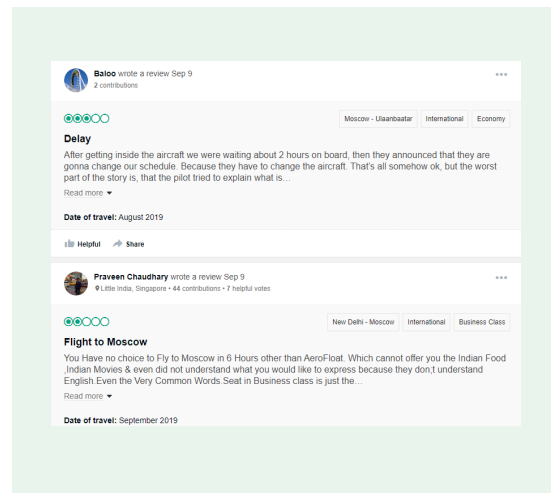
**Flight to Moscow**  
You Have no choice to Fly to Moscow in 6 Hours other than AeroFloat. Which cannot offer you the Indian Food ,Indian Movies & even did not understand what you would like to express because they don;t understand English.Even the Very Common Words.Seat in Business class is just the...  
[Read more](#)**Date of travel:** September 2019

Your goal is to leverage webscraping techniques, topic modelling and sentiment analysis algorithms as well as your business sense to provide insights on possible business opportunities

## Data Sources: Customer review websites



## Data: Customer reviews



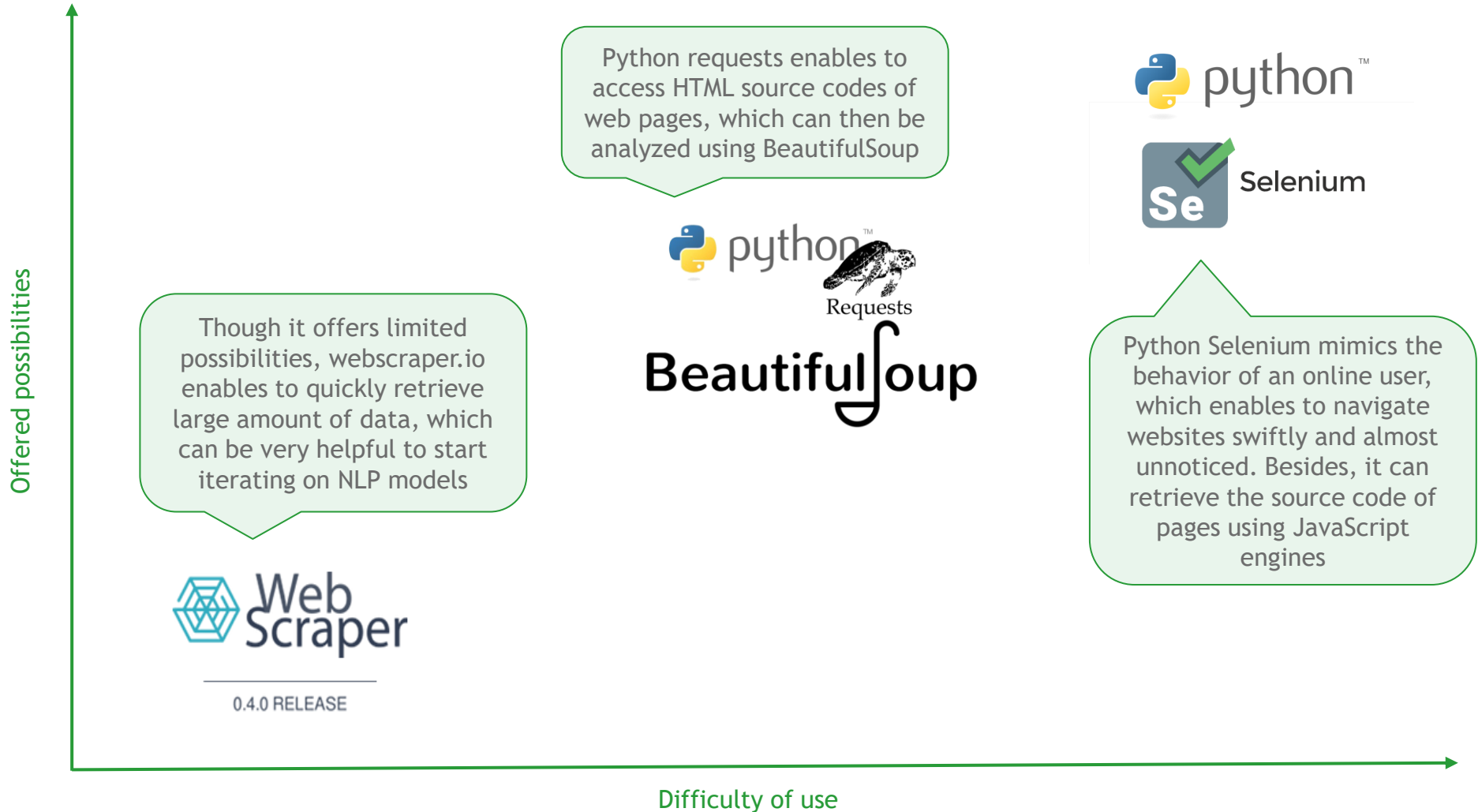
## Output: Business insights, new development opportunities



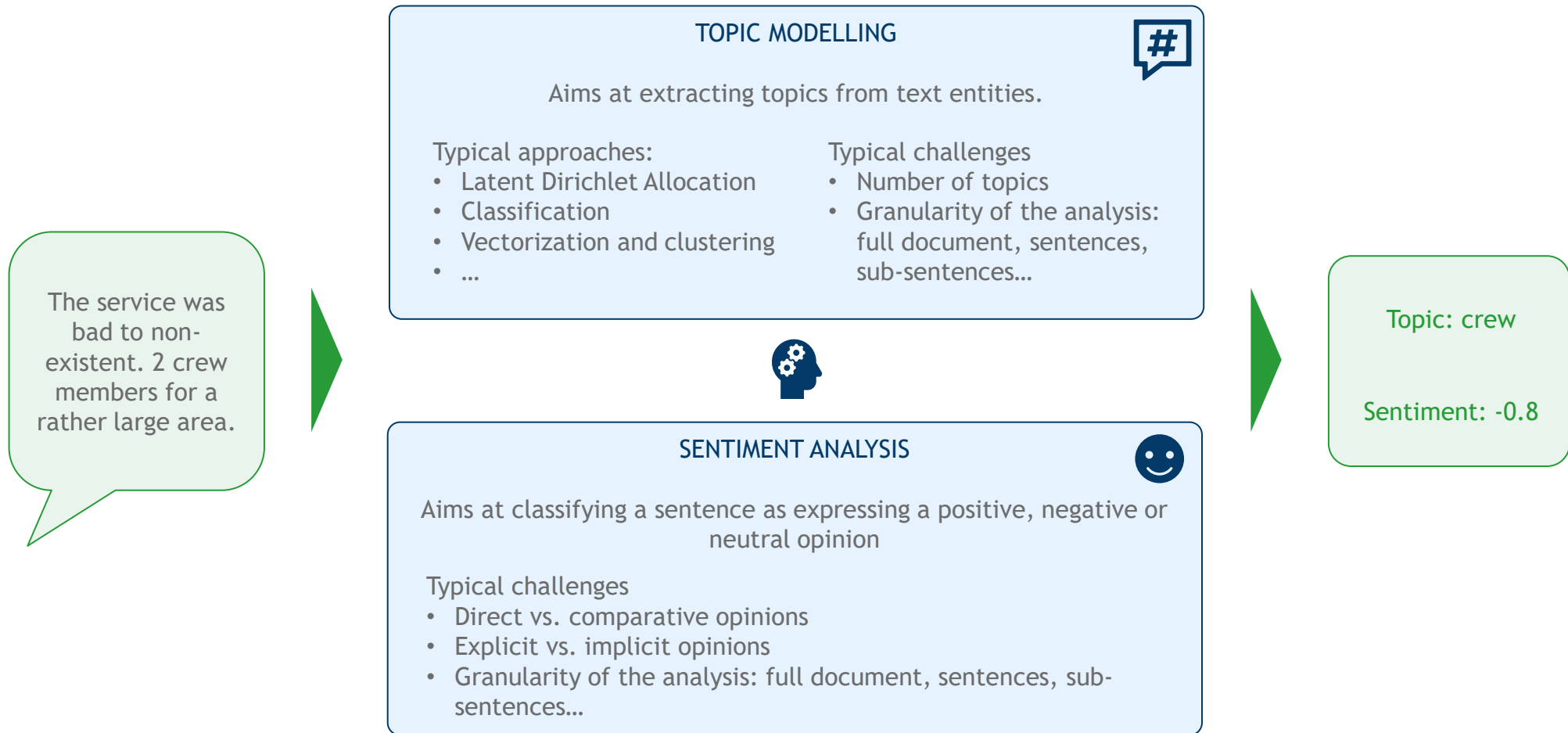
Webscraping

Topic Modelling,  
Sentiment Analysis

Several **web scraping techniques**, with varying ease of use and offered possibilities, enable to **retrieve information from websites**



# Topic modelling and sentiment analysis enable to derive valuable insights from customer reviews



**Resources :** You are free to **use any resources** you want, here are some recommendations to help you get started

## Languages



python™

We suggest you use either R or Python

## Bunch of useful libraries



Selenium

BeautifulSoup

Selenium, BeautifulSoup are powerful Python libraries for webscraping



TextBlob



Gensim, TextBlob are useful Python libraries for Topic Modelling and Sentiment Analysis

## Useful links



<https://www.seatguru.com/browseairlines/browseairlines.php>

[https://www.seatguru.com/airlines/Aegean\\_Airlines/Aegean\\_Airlines\\_Airbus\\_A320-200.php](https://www.seatguru.com/airlines/Aegean_Airlines/Aegean_Airlines_Airbus_A320-200.php)



<https://www.airlinequality.com/review-pages/a-z-airline-reviews/>



<https://www.tripadvisor.com/Airlines>

# AGENDA



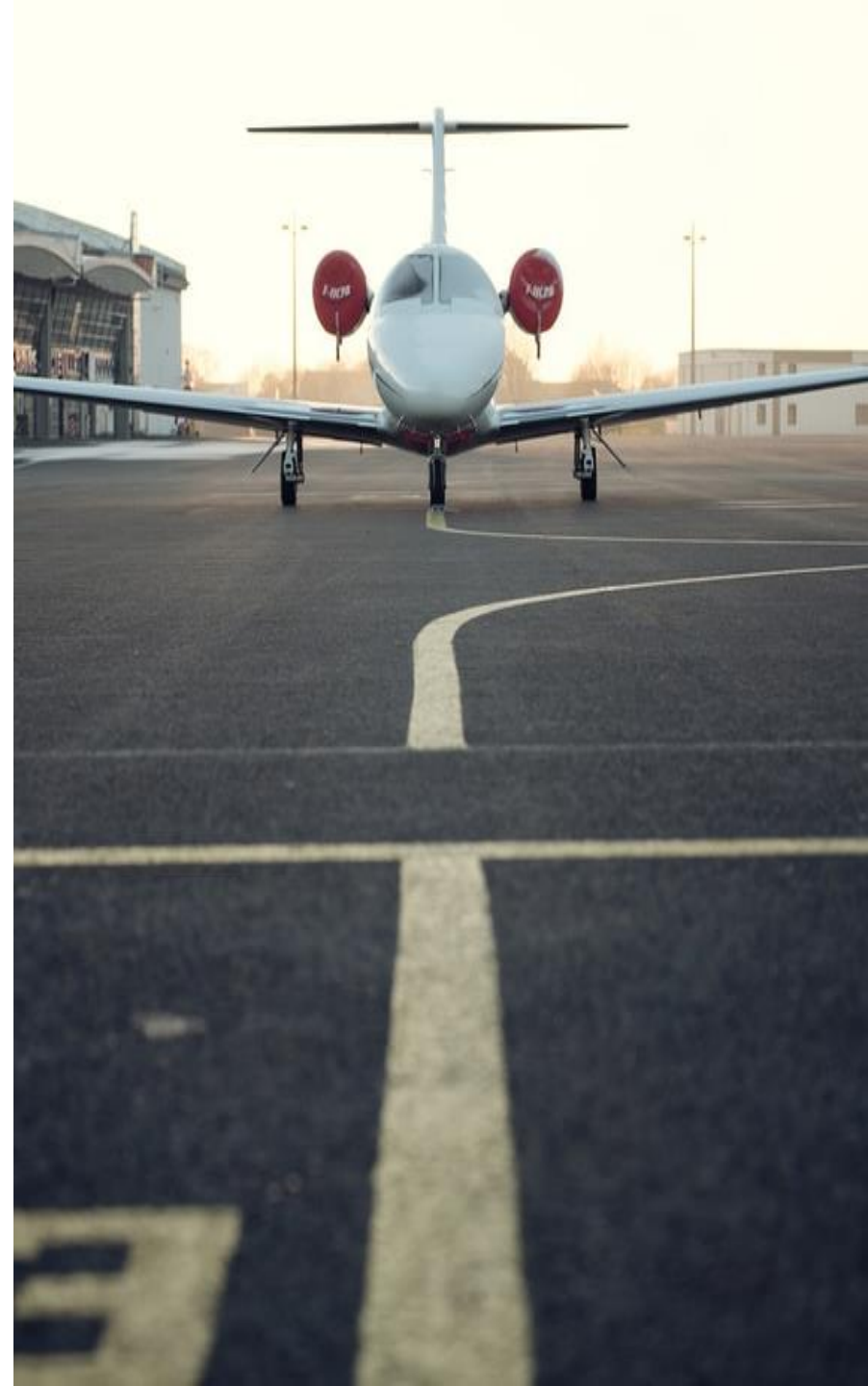
1. About eleven
2. Case presentations
  - a) Airplane Interior Service
  - b) Aircrafts' take-off time prediction
3. Practical information
4. A few advice

# Supercase 2

# Taxi-time prediction

*To the attention of the master in Data  
Science & Business analytics' students*

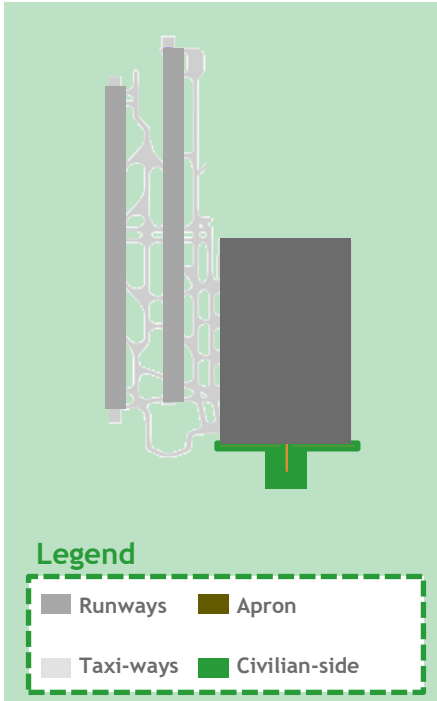
February 8<sup>th</sup>, 2021



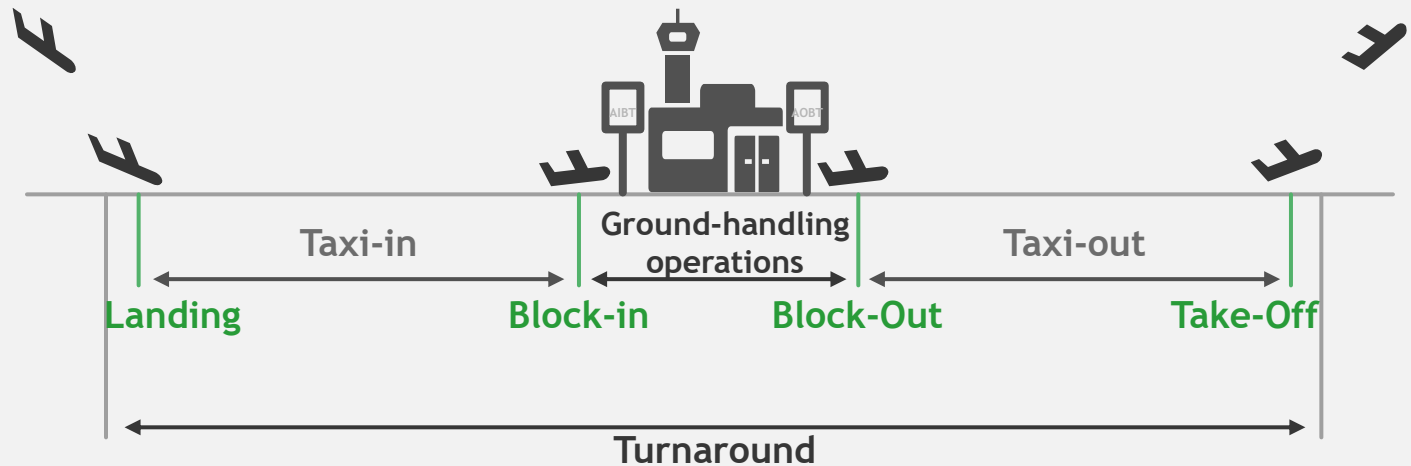


## Take-off time (TOT) prediction - Airport overview

### Airport overview

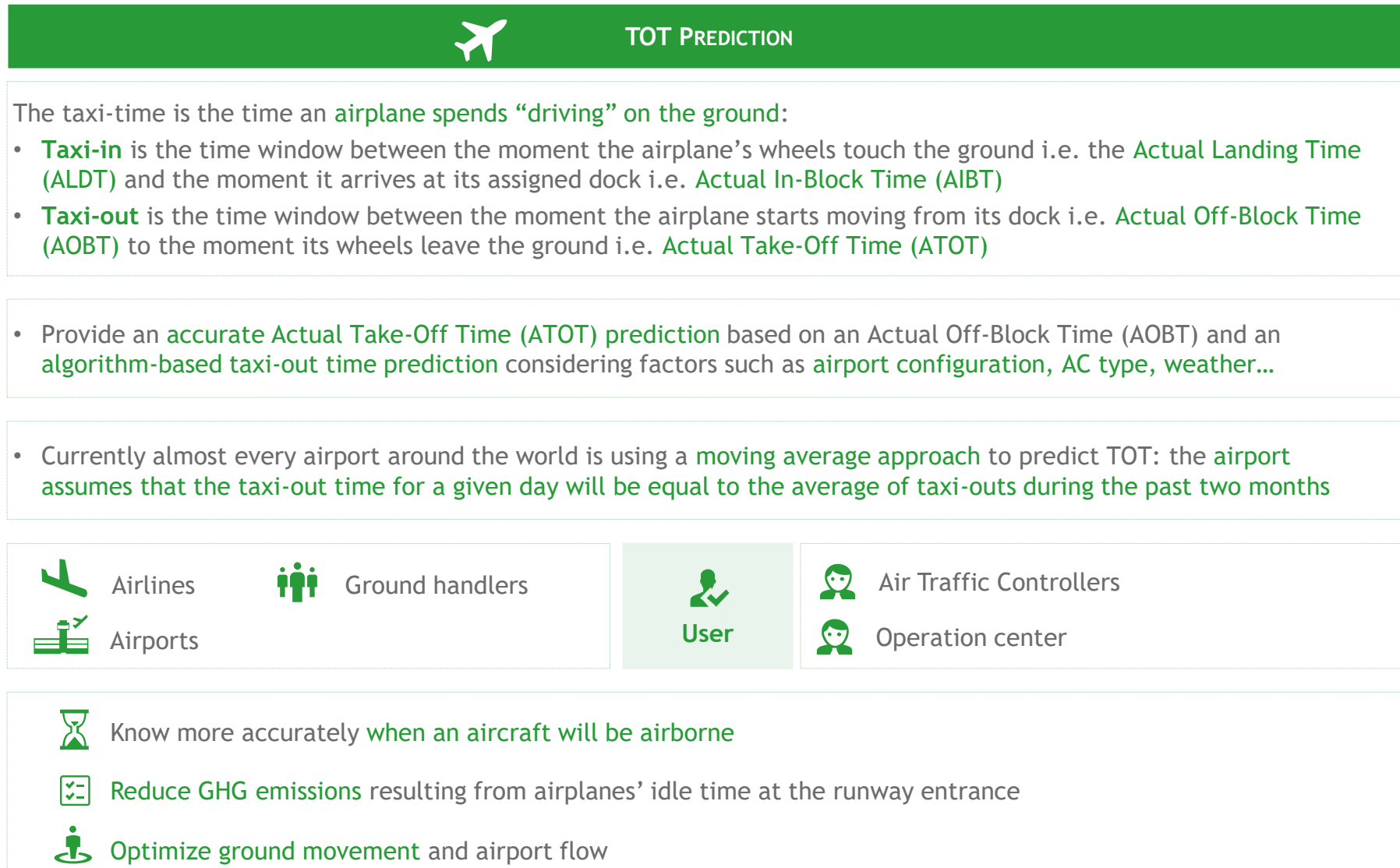


### Turnaround process breakdown



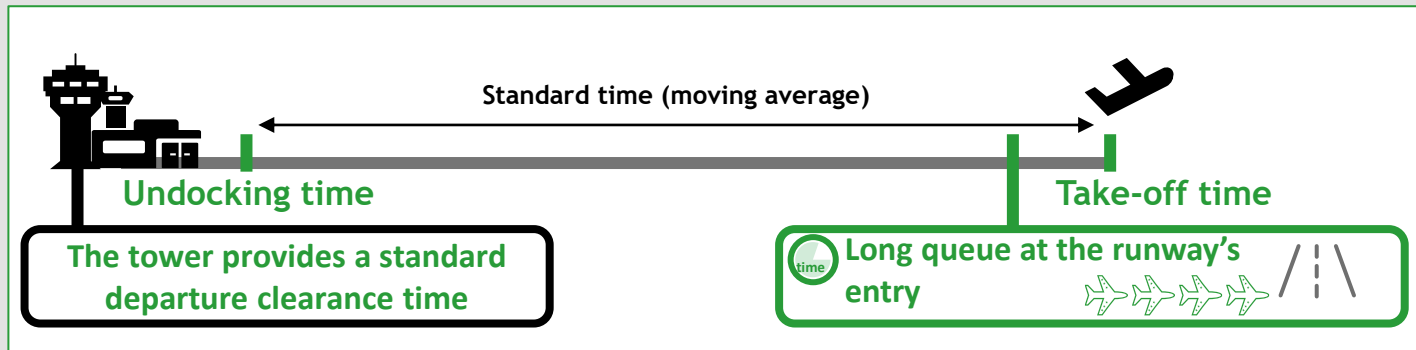
**TAXI-TIME PREDICTION CAN IMPROVE AIRPORT AND AIRLINES OPERATIONS AND REVENUES AS WELL AS REDUCE OVERALL GHG EMISSIONS**

# Take-off time (TOT) prediction - Use case description

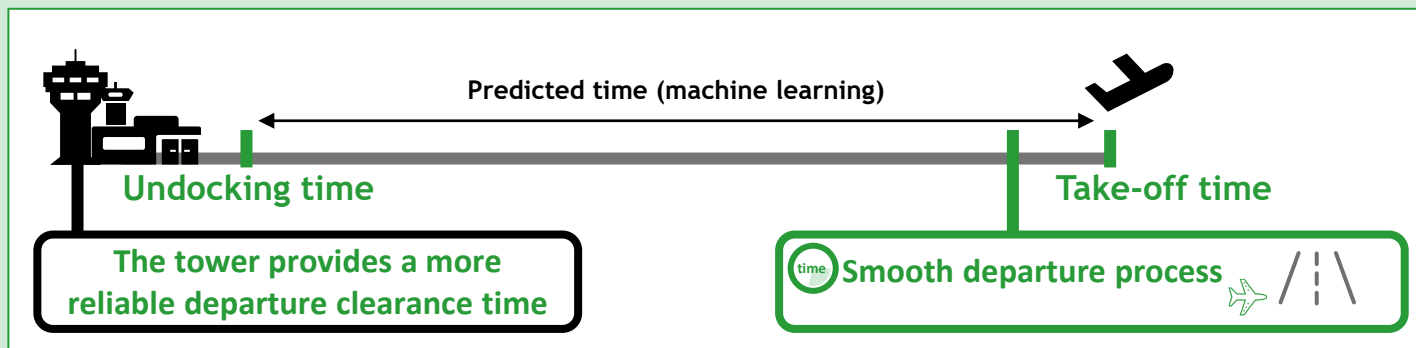


# Take-off time (TOT) prediction - Use case description

Status quo



A.I. solution



  
Awareness

  
Efficiency

  
Safety

  
Green

# Expected output

- A **PowerPoint presentation** that should include at least the following:
  - ✓ A presentation of your **models' results** and how they **compare** to the **status quo** (the moving average)<sup>1</sup>
  - ✓ An explanation of the **expected impact** of your best model **on ground operations** at the airport
  - ✓ A **final assessment** of your models by using **interpretability** methods
- Your **code** which should include:
  - ✓ Your **feature engineering code** specifying how you modified your data and why (make sure to **clearly comment your code** to explain why you processed the data the way you chose to)
  - ✓ Your models' parametrization, training code and testing code

## PRESENTATION



## CODE

```
# Get parameter of vehicle to consider for the selected context
engine = create_engine('mysql+pymysql://root:root@localhost:3306/eleven')
query = "SELECT v.id, v.equipment_plant_number FROM equipment_contexts_vac_vehicles v as v, refills r as r WHERE v.vehicle_to_consider = r.pl_refill_id; query(query, context) # Dataframe with 'id' and 'equipment_plant_number'"

vehicle_to_consider_id = str(tuple(list(set(vehicle_to_consider["id"]))))
vehicle_to_consider_name = str(tuple(list(set(vehicle_to_consider["equipment_plant_number"]))))

print("Vehicle to consider for the context " + str((parameter["context_id"])) + " in the batch " + str((parameter["batch_number"])))
print(set(vehicle_to_consider["equipment_plant_number"]))

# Import DRIVERS LOGS from RDD logs with INPUT = (vehicle_to_consider, startDate, endDate)
if parameter["id_log"] == "Apprenti":
    engine = create_engine('mysql+pymysql://root:root@localhost:3306/eleven')
    query = "SELECT log_id, log_driver_id, name, d.equipment_plant_number, log_action, log_date FROM apprenti_drivers_data d; pl.read_sql_query(query, context) # Dataframe with 'id', 'equipment_plant_number', 'action', 'date'"
    drivers_data = pd.read_sql_query(query, context) # Dataframe with 'id', 'equipment_plant_number', 'action', 'date'"
    drivers_data["action"] = drivers_data["action"].apply(lambda x: int(x))

    drivers_data = drivers_data[drivers_data["equipment_plant_number"] == vehicle_to_consider["equipment_plant_number"]]
    drivers_data = drivers_data[drivers_data["log_id"] > startDate & (drivers_data["log_id"] <= endDate)]
    drivers_data["driver_name"] = drivers_data["equipment_plant_number"] + str(drivers_data["name"])

    id_to_consider = str(tuple(list(set(drivers_data["driver_id"]))))
    query = "SELECT name, name as id FROM refills WHERE id is " + str(id_to_consider) + " AND (equipment_plant_number, 'action', 'date'"
    drivers_data["driver_name"] = drivers_data["equipment_plant_number"] + str(drivers_data["name"])

    drivers_data = drivers_data[drivers_data["action"] == "id"]
    print("Drivers Data InfoFrame with driver log data")
    print("----- Number of rows : " + str(drivers_data.shape[0]))
    print(drivers_data.head())
    print(drivers_data.shape)
    print(set(drivers_data["equipment_plant_number"]))
    print(set(drivers_data["equipment_plant_number"]))) vehicles"
    print(set(drivers_data["driver_name"]))
    print(set(list(set(drivers_data["driver_name"]))) + " names")

if parameter["id_log"] == "Pilotage":
    # ... (code continues) ...

    df_refills["last_refill_date"] = df_refills["last_refill_date"] # last transaction is the last_refill
    df_refills["last_refill_date"] = df_refills["last_refill_date"]
    df_refills["vehicle_name"] = v
    df_refills["hours"] = df_refills["hours"]

    for i in range(1, len(df_refills)):
        df_refills.loc[i, "hours"] = (df_refills.loc[i, "last_refill_date"] - df_refills.loc[i-1, "last_refill_date"])
        diff = df_refills.loc[i, "last_refill_date"] - df_refills.loc[i-1, "last_refill_date"]
        days, seconds = diff.days, diff.seconds
        df_refills.loc[i, "hours_between"] = days * 24 + seconds / 3600
        df_refills["hours"] = df_refills["hours"] + df_refills.loc[i, "hours_between"]
```

Provided input

- An airport terms glossary: Glossary > Glossary.xlsx
- Historical airport and weather data :
  - Airport data:
    - ❑ Data > Airport data > training\_set\_airport\_data.csv
    - ❑ Data > Airport data geographic\_data.csv
  - Weather data:
    - ❑ Data > Weather data > training\_set\_weather\_data.csv
- Academic papers on the taxi-time prediction subject: Taxi time academic papers > Paper 1.pdf...Paper 6.pdf
- Aircraft (A/C) types' characteristics: AC characteristics > ACchar.xlsx
- A test set: Test set - this folder contains weather data, airport data and geographical data for your model testing

AIRPORT & WEATHER DATA

The image shows two overlapping CSV files. The top file is training\_set\_weather\_data.csv, which contains columns for date, time, airport, weather, and taxi time. The bottom file is training\_set\_airport\_data.csv, which contains columns for airport, latitude, longitude, and taxi time.

GLOSSARY

The image shows a screenshot of the Glossary.xlsx file. It contains a list of airport codes and their corresponding names, along with a column for the type of airport (e.g., international, domestic).

TEST SET

The image shows two overlapping CSV files. The top file is test\_weather\_data.csv, which contains columns for date, time, airport, weather, and taxi time. The bottom file is test\_airport\_data.csv, which contains columns for airport, latitude, longitude, and taxi time.

RESEARCH PAPERS



A/C TYPE CHARACTERISTICS

The image shows a screenshot of the ACchar.xlsx file. It contains a list of aircraft types and their corresponding characteristics, such as length, wingspan, and maximum takeoff weight.

# APPENDIX



# Example of model performance comparison sheet

## Chosen models' description and performance overview

		Off-Block Taxi Prediction			
		Rolling average	Model 1	Model 2	...
Key elements		<ul style="list-style-type: none"><li>Average of the last 2 months of taxi-times</li></ul>	<ul style="list-style-type: none"><li>Model details:</li><li>Used features:<ul style="list-style-type: none"><li>...</li><li>...</li></ul></li></ul>	<ul style="list-style-type: none"><li>Model details:</li><li>Used features:<ul style="list-style-type: none"><li>...</li><li>...</li></ul></li></ul>	<ul style="list-style-type: none"><li>Model details:</li><li>Used features:<ul style="list-style-type: none"><li>...</li><li>...</li></ul></li></ul>
Results	Average error vs. real data	2.8 min	xx min	xx min	xx min
	First quartile max error	1.7 min	xx min	xx min	xx min
	Third quartile max error	3.9 min	xx min	xx min	xx min

# AGENDA



1. About eleven
2. Case presentations
  - a) Airplane Interior Service
  - b) Aircrafts' take-off time prediction
3. Practical information
4. A few advice



## The Teams workplace:

For this challenge, we opened a **Teams workplace**

On this workplace, you will find three channels:

- 1) **General**: for all questions and information related to the organization of the challenge
- 2) **Airplane Interior Service**: for all questions specifically related to the Airplane Interior Service case
- 3) **Taxi-time**: for all questions specifically related to the Taxi-time prediction case

*Please use the right channel to ensure fluidity of the interactions*

*Before asking something, also make sure that the requested information has not been given already ;)*



# Schedule for the week



	Monday 8 <sup>th</sup>	Tuesday 9 <sup>th</sup>	Wednesday 10 <sup>th</sup>	Thursday 11 <sup>st</sup>	Friday 12 <sup>th</sup>
Morning session	Case Presentations (9h-10h, Teams)	Training : Natural Language Processing (10h-11h, Teams)	Training : Interpretability (10h-11h, Teams)	Free working session	Free working session
	Training : Github and Python bestpractices (10h-11h, Teams)				
	Q&A : Taxi-Time (11h-12h, Teams)	Free working session	Free working session	Q&A : Taxi-Time (11h-12h, Teams)	
Afternoon session	Q&A : AIS (13h30-14h30, Teams)	Q&A : Taxi-Time Q&A : AIS (13h30-14h30, Teams)	Q&A : AIS (13h-14h, Teams)	Q&A : AIS (13h-14h, Teams)	Oral Presentations (13h40-15h40, Teams)
	Free working session		Q&A : Taxi-Time (13h30-14h30, Teams)	Free working session	Closing session (16h - 17h, Teams)
		Free working session	Free working session		

▶ The Q&A will be 1-hour sessions where you can log in each day to ask questions related to your topic. You will also be able to directly contact the coaches on Teams if you have specific questions

# Expected output

You are expected to deliver the following:

- The file with your **code** (for this assignment we highly recommend using Python and/or R)
- A **PowerPoint presentation** of your work (including your experiment process, your thoughts, the hardships you had to overcome...)

## PRESENTATION



## CODE

```
# get perimeter of vehicle to consider for the selected context
engine = create_engine('mysql://root@localhost:3306/eleven112-113-114?param=(%s)')
query = "SELECT v.id, v.equipment_plant_number FROM considering_contexts_has_vehicles as ddc, 'det_vehicle' as v WHERE
vehicle_to_consider = pl.read_sql_query(query,connection) # Database with 'id' and 'equipment_plant_number'"

vehicle_to_consider_id = str(tuple(list(set(vehicle_to_consider['id']))))
vehicle_to_consider_name = str(tuple(list(set(vehicle_to_consider['equipment_plant_number']))))

print("Vehicle to consider for the context: "+str(param["context_id"])+ " in the batch: "+str(param["batch_number"]))
print(set(vehicle_to_consider['equipment_plant_number']))

# Import DRIVERS LOGS from RED logs with IMPUT = (vehicle_to_consider, startdate, endDate)
if param["id_log"] == "approach":
    engine = create_engine('mysql://root@localhost:3306/eleven112-113-114?param=(%s)')
    query = "SELECT log_id, log_driver_id, name, d.name, r.equipment_plant_number, log_action, log_date FROM approa
drivers_data = pl.read_sql_query(query,connection) # DataFrame with 'id', 'equipment_plant_number', 'action', 'date'
drivers_data['action'] = drivers_data['action'].apply(lambda x: int(x))

drivers_data = drivers_data[drivers_data['equipment_plant_number'].isin(list(vehicle_to_consider['equipment_plant_
drivers_data = drivers_data[drivers_data['log'] > startdate] & [drivers_data['log'] <= enddate]]
drivers_data['driver_name'] = drivers_data['name'] + " " + drivers_data['name']

id_to_consider = str(tuple(list(set(drivers_data['driver_id']))))
query = "SELECT name, 'name', id FROM 'det_driver' WHERE id IN "+id_to_consider
driver_names = pl.read_sql_query(query,connection) # DataFrame with 'id', 'equipment_plant_number', 'action', 'date'
drivers_names['driver_name'] = drivers_names['name'] + " " + drivers_names['name']

drivers_data = drivers_data[['action','date','id','equipment_plant_number','driver_id','driver_name']]
print(drivers_data.info())
print("----- Number of rows: " + str(drivers_data.shape[0]))
print(drivers_data.head())
print(drivers_data.shape)
print(list(drivers_data['equipment_plant_number']))
print(str(list(set(drivers_data['equipment_plant_number'])))+" vehicle")
print(list(drivers_data['driver_name']))
print(str(list(set(drivers_data['driver_name'])))+" names")

if param["id_log"] == "approach":
    df_refills['last_refill_date'] = df_first['last_refill_date'] # last transaction is the last_refill
    df_refills['date'] = df_last['date']
    df_refills['vehicle_name'] = v
    df_refills['hours'] = df_last['hour']

    for i in range(1, len(df_refills)):
        df_refills.loc[i, 'hours'] = (df_refills.loc[i, 'date'] - df_refills.loc[i-1, 'date']).total_seconds() / 3600
        diff = df_refills.loc[i, 'date'] - df_refills.loc[i-1, 'last_refill_date']
        days, seconds = diff.days, diff.seconds
        df_refills.loc[i, 'hours_between'] = days * 24 + seconds / 3600
        df_refills['hours'] = min(df_refills.loc[i, 'hours'], df_refills.loc[i, 'hours_between'])
```

started between two refil:  
vehicleName  
don't know the consump  
t) # some transaction

## Final presentation details and best practices:

On **Friday 12<sup>th</sup>**, you will have to present your work in front of a **jury** during a **closed-door session**

The modalities of the presentation will be as follow:

- **15min group pitch** based on a PowerPoint presentation
- **~10min Q&A session** with the jury
- **~5min debrief** from the jury

For each supercase, a winner will be announced. The two winners will then **present their work to the other students** (same modalities with questions from the students)

---

The presentation must be **as professional as possible**. Here are some advices and best practices that may be useful:



- **Structure your presentation:** start by stating the problem that you want to solve, then present the way you tackled it, and finally describe your solution. The “story” of the presentation should be natural and easy to follow
- **Be concise and precise:** focus on the most important messages, as you only have 20 minutes to present the work achieved for the entire week. You should limit the number of slides you present (you can still add appendices if needed)
- **Be organized as a team:** split up the speaking time between the team members beforehand to make it smoother
- **Be honest:** tell where you encountered issues or challenges
- **C-suite level:** you should convince both the CEO and the CTO/CDO of the company

## Evaluation criteria:

Although different in their essence, all supercases will be graded based on **similar criteria**:

1. The **engagement** of the team during the week (how far you've gone, how autonomous you have been, etc.)
2. The **creativity** and **relevance** of the **methodology** (i.e. scientific approach) you choose regarding the problem you have to solve and the data you had
3. **Coding assessment** (see next slide)
4. The **performance** of your model (specific to each supercase)
5. The explanation of your **technical choices** and your ability to present them in non-technical terms
6. The critical **business views** on your **current results**, and the **next steps** you could consider to improve them
7. The **quality of your final presentation**: how professional it looks (**slide quality**), how clear and complete it is (**storytelling**), how pertinent your answers are, etc.
8. Your **relative overall performance** compared to other groups



Please note that all groups will be graded at the end of the week

## Code evaluation criteria :

1. **Your code must be easy to run:** someone outside your group must be able to run your code autonomously
  - Write a clear readme with instructions
  - Set-up an environment to run the project with the necessary packages
  - Write a single/few main files
  - ...
2. **Your code must be well-structured:**
  - Use functions/OOP
  - Break-up the pipeline in several files for each steps (preprocessing, feature engineering, etc)
  - ...
3. **Your code must be easy to understand:**
  - Document your code with clear comments
  - Use understandable variable/function names
  - Respect pep8 recommendations\*
  - ...



\*<https://www.python.org/dev/peps/pep-0008/>

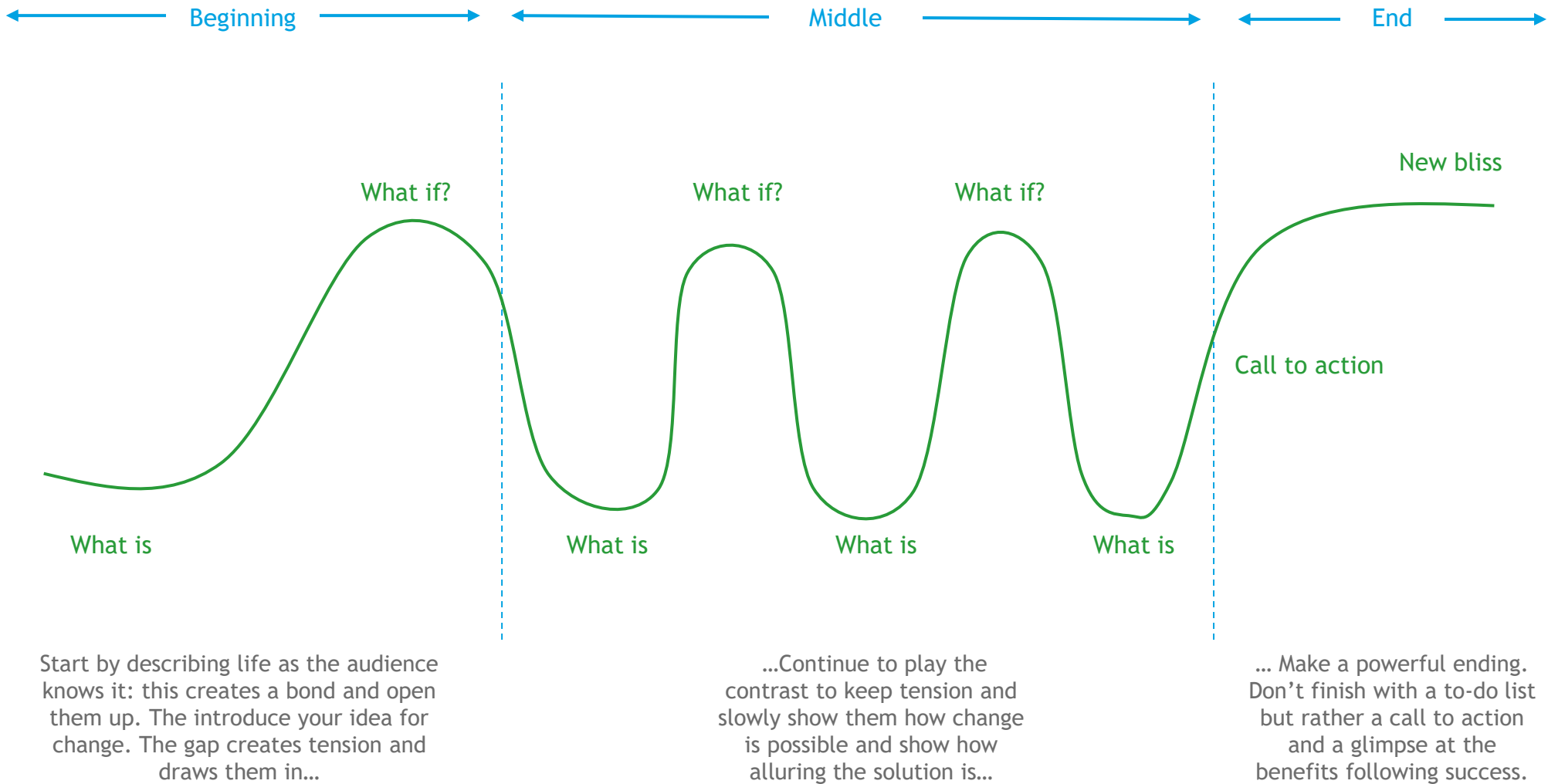
# AGENDA



1. About eleven
2. Case presentations
  - a) Airplane Interior Service
  - b) Aircrafts' take-off time prediction
3. Practical information
4. A few advice

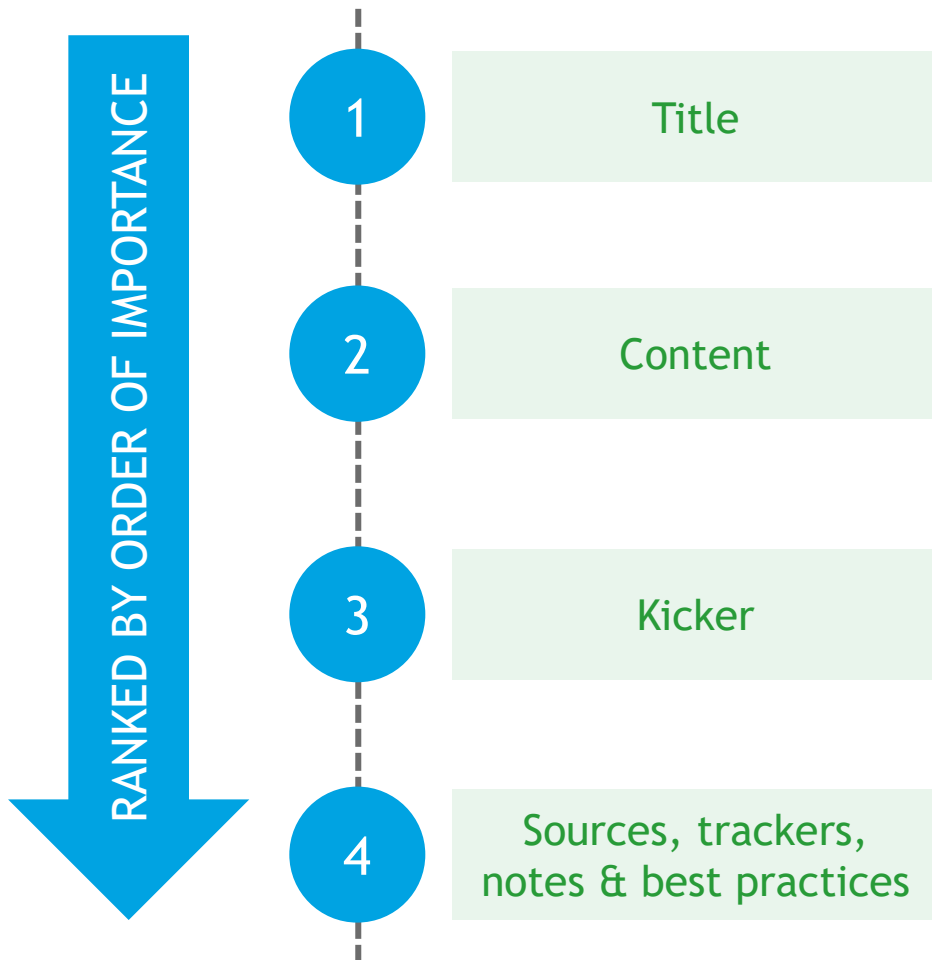
## Presentation best practices (1/2): the anatomy of a story

By reminding people of the status quo and then revealing the path to a better way, they set up a conflict that needs to be resolved.





### How to structure a slide



### Other key points on slide design



1 message per slide !



The content of the slide must illustrate/ demonstrate/ have a link with the title



The title can be either descriptive-oriented or action-oriented



Once put all together, your titles should read like a coherent story

# Any Questions ?

Enjoy the challenge ...







Land you project !