

# Hydrogen Trucks Charging Stations

Data Analytics for Sustainability

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# TO SIZE THE H2 TRUCK CHARGING STATIONS NETWORK IN 2030 AND 2040 WE DEVELOPED A FLEXIBLE RULED-BASED MODEL

## METHODOLOGY

A very flexible and interpretable model which takes into account:

1. Initial assumptions/parameters:
  - a. On the manufacturers: tank size, autonomy, type of PL (short or long distance), etc.
  - b. On the stations: costs, storage onsite, footprint, etc.
  - c. On the market: split between the manufacturers, daily average distance (for each type of PL), the security buffer, the activation rate of the fleet, the desired split between the station types (small, medium, large), etc.
2. The current distribution of trucks (based on the government traffic data)
3. The chosen positioning strategy with the strategic positioning index: 1 = meet the demand, 1.5 = get 50% ahead of the demand, etc.

```
manufacturers_desc = {
    'man_1': {'name': 'Daimler Truck & Volvo', 'prototype': 'GenH2 Truck', 'technology': 'Hydrogen fuel cell'},
    'man_2': {'name': 'DAF', 'prototype': 'XF, XG and XG+', 'technology': 'Internal combustion', 'tank_size': 100},
    'man_3': {'name': 'Iveco & Nikola & Hyundai', 'prototype': 'Nikola TRE', 'technology': 'Hydrogen fuel cell'}
}

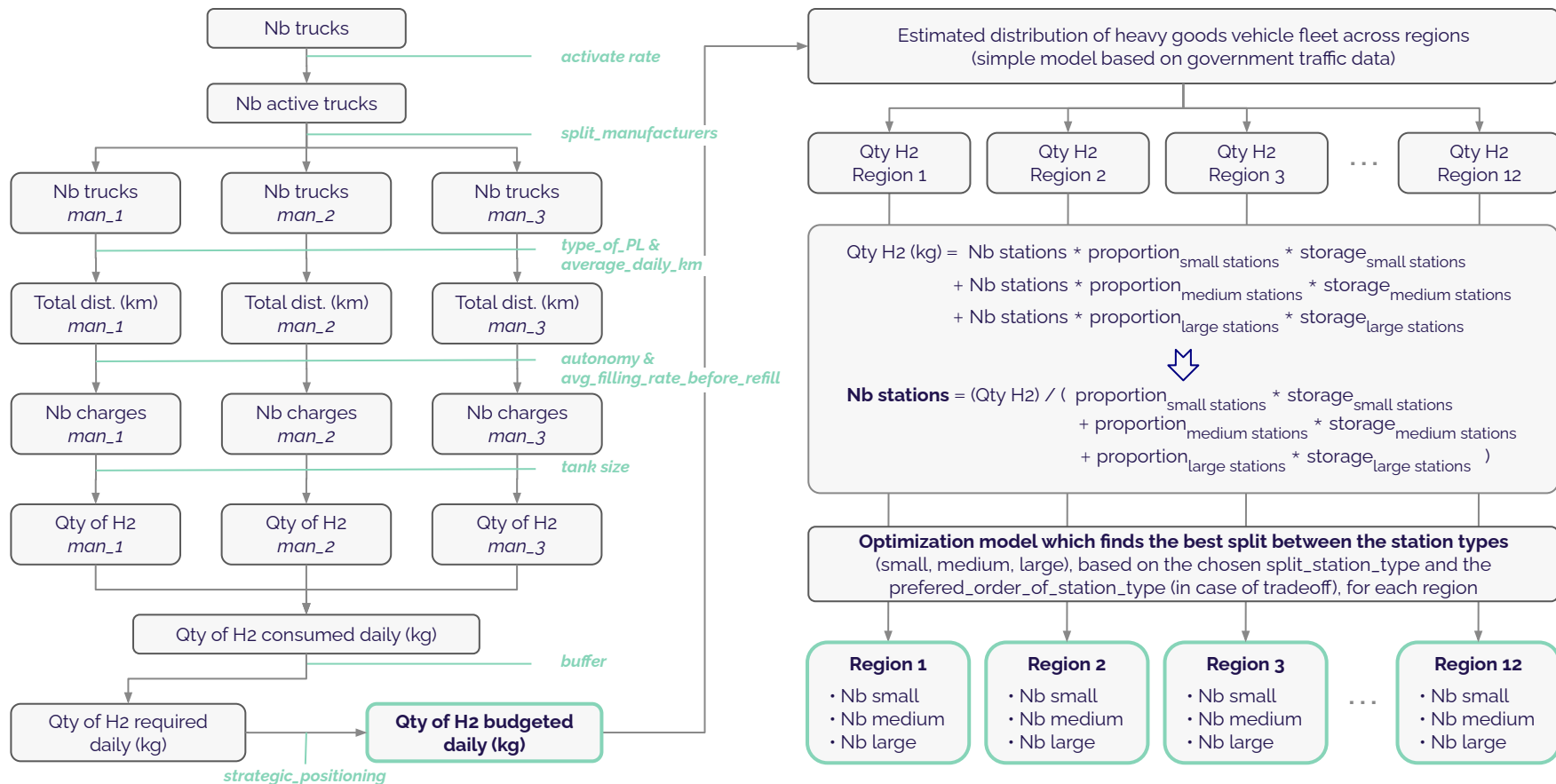
stations_desc = {
    'small': {'capex': 3*10**6, 'depreciation': 15, 'yearly_opex': 0.10, 'storage_onsite': 2000, 'construction_time': 12},
    'medium': {'capex': 5*10**6, 'depreciation': 15, 'yearly_opex': 0.08, 'storage_onsite': 3000, 'construction_time': 12},
    'large': {'capex': 8*10**6, 'depreciation': 15, 'yearly_opex': 0.07, 'storage_onsite': 4000, 'construction_time': 12}
}

parameters = {
    'year': 2030,
    'nb_trucks': 10000,
    'manufacturers_desc': manufacturers_desc,
    'split_manufacturer': {'man_1': 0.4, 'man_2': 0.4, 'man_3': 0.2},
    'activation_rate': 0.8,
    'avg_daily_km': {'short-distance': 290, 'long-distance': 458},
    'stations_desc': stations_desc,
    'split_station_type': {'small': 1/2, 'medium': 1/3, 'large': 1/6},
    'average_tank_filling_rate_before_refill': 0.2,
    'security_buffer': 0.2,
    'strategic_positioning_index': 1,
    'preferred_order_of_station_type': ['small', 'medium', 'large']
}
```



region	quantity_h2_to_reach	quantity_h2_proposed	small	medium	large	total
Auvergne-Rhône-Alpes	62993	64000	12	8	4	24
Bourgogne-Franche-Comté	26739	28000	6	4	1	11
Bretagne	11219	12000	3	2	0	5
Centre-Val de Loire	11978	12000	3	2	0	5
Grand Est	25802	26000	5	4	1	10
Hauts-de-France	14583	15000	4	1	1	6
Normandie	14007	15000	4	1	1	6
Nouvelle-Aquitaine	31159	32000	6	4	2	12
Occitanie	23645	24000	4	4	1	9
Pays de la Loire	10352	11000	2	1	1	4
Provence-Alpes-Côte d'Azur	12228	13000	3	1	1	5
Île-de-France	33274	34000	7	4	2	13
Total	277979	286000	59	36	15	110

# H2 TRUCK CHARGING STATIONS NETWORK MODEL IN DETAILS



```

1 national_strategy = size_the_network.define_best_national_strategy(parameters=parameters,
2                               manufacturers_desc=manufacturers_desc,
3                               stations_desc=stations_desc,
4                               verbose=True)
5 national_strategy

```

✓ 0.2s

Estimated number of H2 trucks in 2030: 10000

Number of daily active H2 trucks (activation\_rate: 0.8)

- Man.1 (Daimler Truck & Volvo): 3200 (40.0%)
- Man.2 (DAF): 3200 (40.0%)
- Man.3 (Iveco & Nikola & Hyundai): 1600 (20.0%)
- Total: 8000

Distance travelled daily based on the type of PL (in km, short-distance: 290; long-distance: 458)

- Man.1: 1465600
- Man.2: 928000
- Man.3: 732800
- Total: 3126400

Actualised autonomy of a truck (in km, average\_tank\_filling\_rate\_before\_refill = 0.2)

- Man.1: 800
- Man.2: 120
- Man.3: 320

Number of necessary daily charges for a truck:

- Man.1: 0.57
- Man.2: 2.42
- Man.3: 1.43

Daily consumed quantity of H2 (in kg)

- Man.1: 145920 (tank\_size: 80)
- Man.2: 27360 (tank\_size: 15)
- Man.3: 58368 (tank\_size: 32)
- Total: 231648

Daily required quantity of H2 (in kg, with a security buffer of 20%)

- Total: 277977.6

Daily budgeted quantity of H2 (in kg, with a strategic positioning index of 1)

- Total: 277978

Estimated number of stations needed

- Small: 52.5
- Medium: 35.0
- Large: 17.5
- Total: 105

First proposition - number of stations

- Small: 53
- Medium: 36
- Large: 18

First proposition - quantity of H2: 286000

> Performing optimization...

> Optimization done!

Final number of stations

- Small: 53
- Medium: 35
- Large: 17

Quantity of H2 to reach: 277978

Proposed quantity of H2: 279000

	region	quantity_h2_to_reach	quantity_h2_proposed	small	medium	large	total
0	Total	277978	279000	53	35	17	105

```

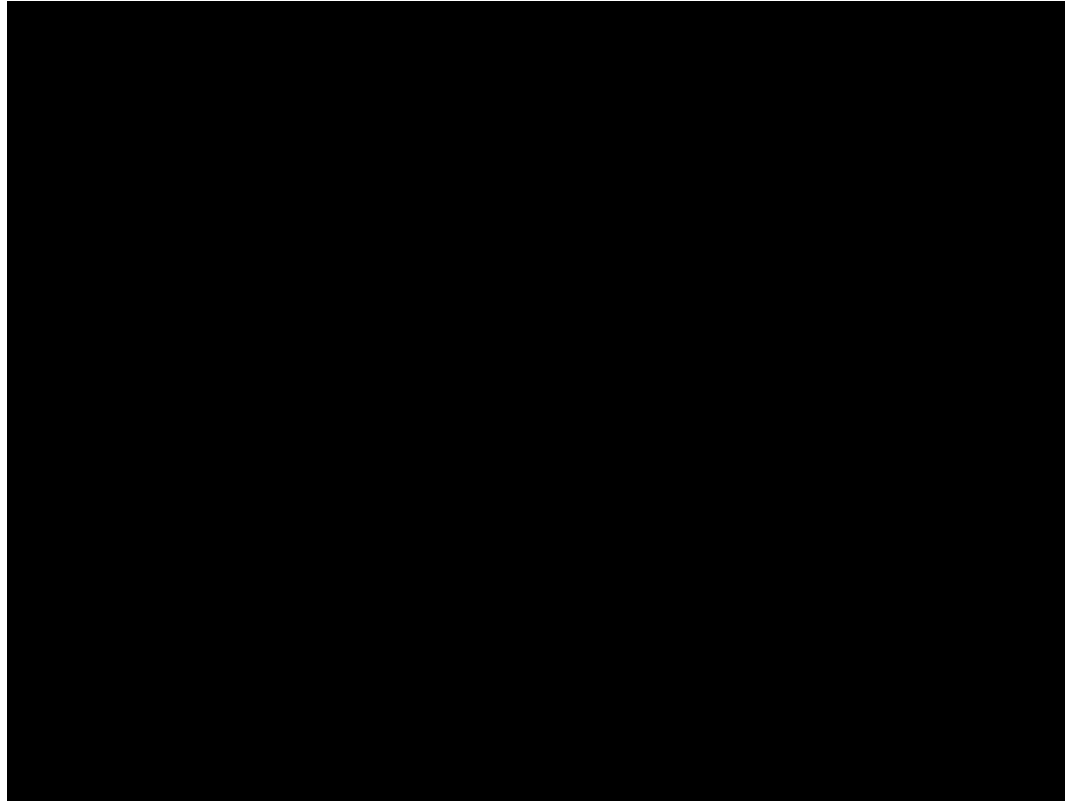
1 regional_strategies = size_the_network.define_best_regional_strategies(parameters=parameters,
2 | manufacturers_desc=manufacturers_desc,
3 | stations_desc=stations_desc,
4 | region_breakdown=region_breakdown,
5 | verbose=False)
6 regional_strategies

```

✓ 0.0s

	region	quantity_h2_to_reach	quantity_h2_proposed	small	medium	large	total
0	Auvergne-Rhône-Alpes	62993	64000	12	8	4	24
1	Bourgogne-Franche-Comté	26739	28000	6	4	1	11
2	Bretagne	11219	12000	3	2	0	5
3	Centre-Val de Loire	11978	12000	3	2	0	5
4	Grand Est	25802	26000	5	4	1	10
5	Hauts-de-France	14583	15000	4	1	1	6
6	Normandie	14007	15000	4	1	1	6
7	Nouvelle-Aquitaine	31159	32000	6	4	2	12
8	Occitanie	23645	24000	4	4	1	9
9	Pays de la Loire	10352	11000	2	1	1	4
10	Provence-Alpes-Côte d'Azur	12228	13000	3	1	1	5
11	Île-de-France	33274	34000	7	4	2	13
12	Total	277979	286000	59	36	15	110

# Visualisation of Scenarios and H2 stations per region



## 5 panels :

Description of the steps & positions taken

- Number of stations per region for the scenarios
- Optimal locations of H2 stations
- Competition strategies
- Optimal location of H2 production plants

# BASED ON AIR LIQUIDIC STRATEGIC POSITIONING, WE IDENTIFIED THREE SCENARIOS

Here, we considered a setup with those parameters:

- nb\_trucks: 10k (2030) and 60k (2040)
- activation\_rate: 80% - security\_buffer: 20%
- manufacturer\_split: {'man\_1': 1/3, 'man\_2': 1/3, 'man\_3': 1/3}
- station\_split: {'small': 1/2, 'medium': 1/3, 'large': 1/6}

We get those results:

- 2030: 231,648 kg of  $H_2$  → 93 stations (49 small, 34 medium, 10 large)
- 2040: 1,389,888 kg of  $H_2$  → 529 stations (269 small, 178 medium, 82 large)

## Conservative: Playing it Safe

### Scenario description

- Low coverage of  $H_2$  station need  
*strategic\_positioning\_index: 50%*
- Certainty to have high demand.
- Possible drawback : slow down of the trucks replacement.

### Number of stations to build

- 2030: 47 (25 - 17 - 5)
- 2040: 265 (135 - 89 - 41)

## Medium: Patiently Progressive

### Scenario description

- Cover the estimated demand  
*strategic\_positioning\_index: 80%*
- High probability to have matching demand.
- In line with current previsions.

### Number of stations to build

- 2030: 75 (40 - 27 - 10)
- 2040: 423 (215 - 142 - 66)

## Bold: Visionary Leadership

### Scenario description

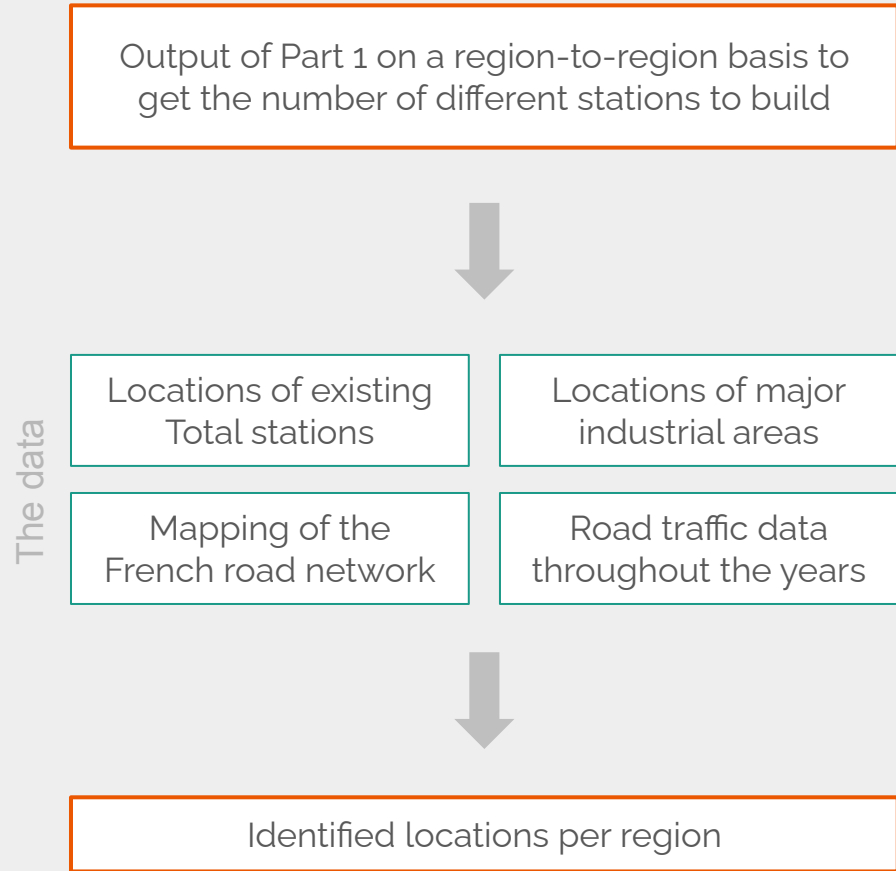
- Plan to build more than needed  
*strategic\_positioning\_index: 120%*
- Less demand than offer.
- Expectations : accelerate trucks replacement and have higher number of trucks than predicted

### Number of stations to build

- 2030: 111 (71 - 37 - 12)
- 2040: 635 (323 - 213 - 99)

## Part 2 - Location Identification

- Identify what makes a good location for a H2 Station
- Find best spots for such stations across the country





# First model - Greedy Permutations

- Use “points de repères” from motorway traffic data as potential locations for stations
- Proceed on a station type basis (big then medium then small)
- Assign a “fitness” score to each PR based on profitability threshold, traffic and distance to industrial areas
- Randomly select a permutation of PRs and evaluate their total fitness, compounded by their distance to each other
- Output the “fittest” found permutation

**Profitability = F (Traffic, consumption, Station type)**

**Point fitness = F (profitability, distance to IA)**

**Permutation fitness = F (point fitness, distance between points)**

# Greedy Permutations

- Explainability
- Customizability
- Large network
- Computational cost
- Simple approach
- Randomness

Example: Say we want 1 large, 2 medium and 2 small stations in the department 66

	point	coords	fitness
--	-------	--------	---------

0	227A000911	(698585.43, 6196493.63)	1.319786
---	------------	-------------------------	----------

	point	coords	fitness
--	-------	--------	---------

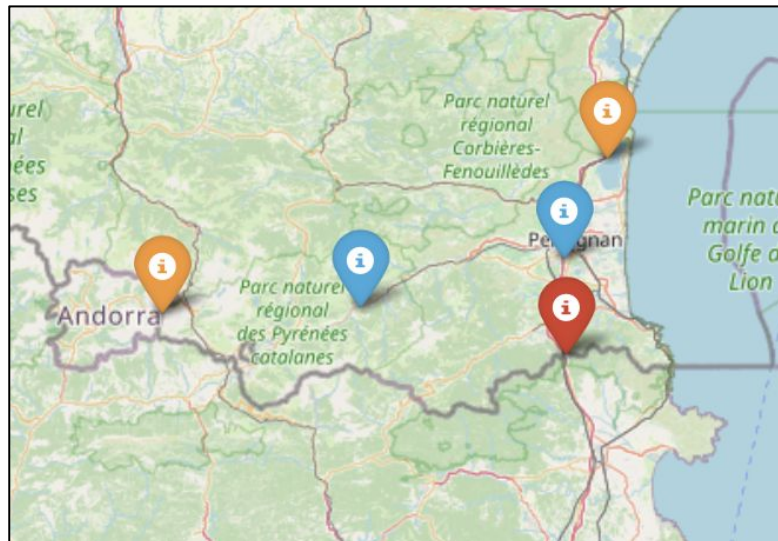
0	280A000966	(688911.76, 6151643.75)	1.388219
---	------------	-------------------------	----------

1	5N002266	(596176.86, 6161871.46)	1.855467
---	----------	-------------------------	----------

	point	coords	fitness
--	-------	--------	---------

0	56N011666	(641612.81, 6162697.65)	1.017484
---	-----------	-------------------------	----------

1	255A000966	(688619.9, 6173843.4)	1.111118
---	------------	-----------------------	----------



# Alternative model - priority based

1. Create area around each road section

## Number of areas based

2. Define the number of areas you want
3. Pick the busiest road section and increase the area around it until it encompasses area of france / number of areas
4. Do the same for the next busiest road section that is not in the previous area
5. Iterate until you have the number of areas you want
6. For each area, determine the center and pick the closest existing station

## H2 need based

2. Define the weight of H2 you want to serve
3. Pick the busiest road section and increase the area around it until it encompasses the maximum traffic servable by the chosen station size
4. Do the same for the next busiest road section that is not in the previous area
5. Iterate until all the weight can be served
6. For each area, determine the center and pick the closest existing station

## Part 3 - Deployment Plan

- Monopoly scenario
- Duopoly competition
- Late entrant

**Demand of H2** will be modeled in an exponential following the model of EV trucks deployment(1).

Find the timeline for deploying stations according to return on investment



Prioritise stations with the best score according to profitability & traffic

Adapt to the competition by maximizing relative payoff



Defined steps & strategy for deployment

(1) <https://www.iea.org/data-and-statistics/data-tools/global-ev-data-explorer>

# Monopoly strategy

## Amortization of stations (not including depreciation)

Large : 1 year and 3 month

Medium : 1 year and 2 month

Small : 1 year

## Set importance on mapping to incentivize growth of H2 trucks fleet

Choose location to provide H2 across France so trucks can recharge as needed

### For 2030

**2024**

**2028**

Small : 25

Small : 24

Medium : 24

Medium : 24

Large : 24

Large : 25

Investment : 387 M€

Amortized in 2026,  
possible to build same  
amount in 2028 with no  
further investment

Investment : on payoff

Amortized in 2031

TOTAL 2030 : 148

### For 2040

Same reasoning with the following steps to keep up with demand :

	2031	2033	2036	2038
Small	40	37	51	70
Medium	40	37	50	69
Large	40	37	50	69

TOTAL 2040 : 740

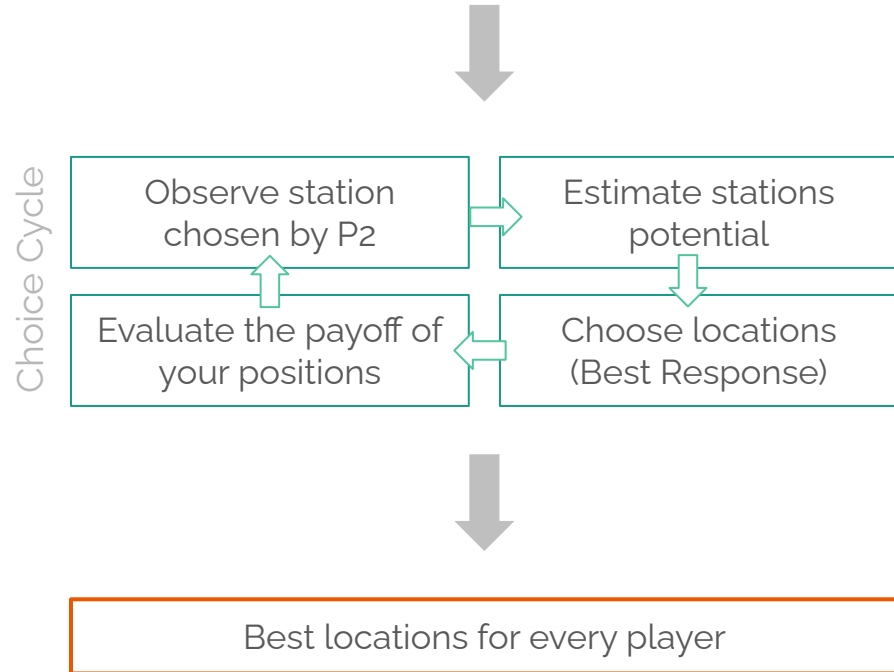
## Part 3 Duopoly & Entrant Scenarios

### Flexibility

Type of competition :

- Aggressive (reward if the other loses payoff)
- Pacific (reward if we are not exposed to competition)
- Number of stations chosen per round
- Learning rate

Multi Agent Reinforcement Learning (Game Theory model)



# Reinforcement Learning model

## Selection of stations

### Explore :

Try random picking of stations

### Exploit :

Select best stations according to score (best response)

### Learning rate :

Balance the number of explorations / number of exploitations

## Reward

### Payoff of each station:

- (storage x H2 price) if max of H2 needed is not covered.
- Weighted distributed payoff if H2 offered by both players in the region > H2 needed

### Final reward: payoff ...

- ... adjusted with a coeff of number of adverse stations in the area if strategy is safe.
- ... adjusted with payoff of other player if strategy is bold

## Update

Removing the stations already chosen from possibilities.

Updating the score of each station given its payoff relative to the general payoff taking closeness to opponent into account depending on strategy

## Training

Choose number of batch to train the model.

Play the game as many times as batch & refine the station score through training using the update & keeping previous update in memory.

# Results Duopoly for 2030

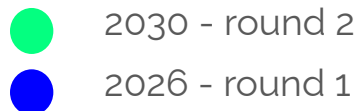
2 steps of selection

Strategy : safe

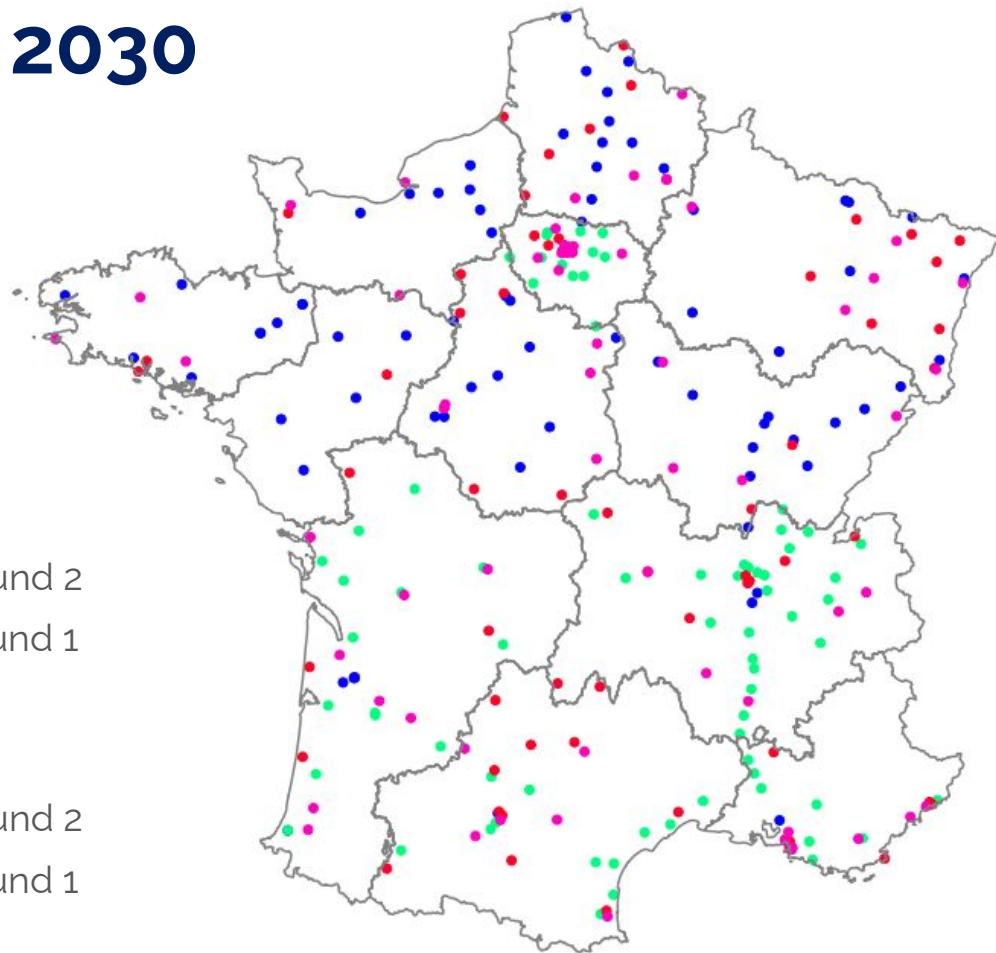
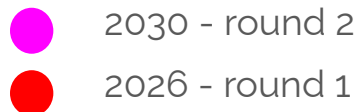
74 stations chosen  
per round

Batch = 10

## AirLiquide



## Competitor





# Entrant Scenario

## compute the remaining capacity to provide

Modelise the deployment of the competitor given an exponential following the deployment of trucks

Compute H2 quantity offered per region

Subtract from total needed quantity modelised per year per region (exponential too)

## Find advantage threshold

When the remaining H2 need per region is too low to allow for new stations plan to be profitable → threshold to enter on the market.

After it will not be profitable to enter the market.

## Deploy according to strategy game of duopoly

Given the remaining quantity to provide : compute possible number of stations per region.

Deploy these station in the same optimized framework than for the duopoly

Can be simplified by following the priority order previously defined

# Deployment road map (Part 3)

## Monopoly

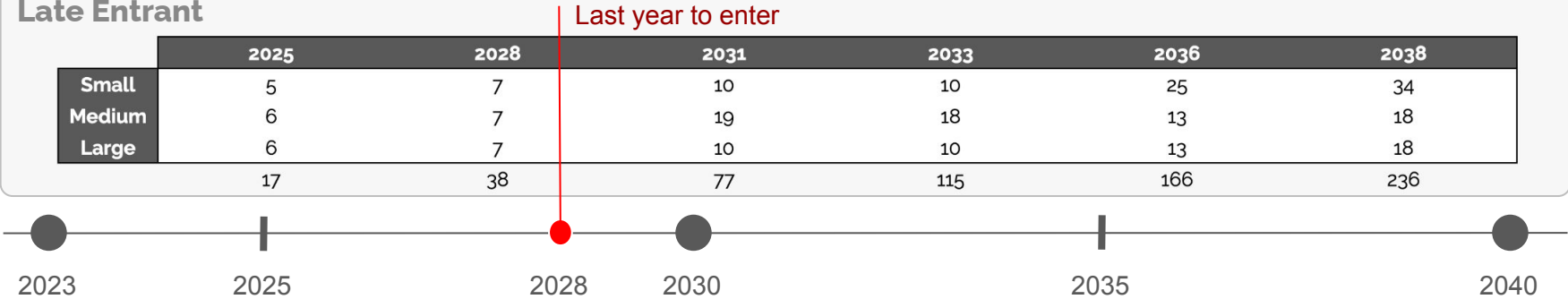
	2025	2028	2031	2033	2036	2038
Small	25	25	40	37	51	70
Medium	24	25	40	38	50	70
Large	24	25	40	37	50	69
	73	148	268	380	531	740

## Duopoly

	2025	2028	2031	2033	2036	2038
Small	13	13	20	19	25	34
Medium	12	13	19	18	25	35
Large	12	13	20	19	25	35
	37	76	135	191	266	370

## Late Entrant

	2025	2028	2031	2033	2036	2038
Small	5	7	10	10	25	34
Medium	6	7	19	18	13	18
Large	6	7	10	10	13	18
	17	38	77	115	166	236



## Part 4 - H2 production facilities

- Size the need for H2 production plants based on stations data
- Identify what makes a good location for building a H2 production plant
- Find best locations according to our strategy

The data

Output of Part 2 on region-linked target number of different stations



Building and transport costs

Number of necessary plants

Potential locations dataset

Production and capacity



Identified locations over the whole territory

# Defining our model parameters

## Potential locations

- Need an expansive dataset for permutations
- Across whole territory
- Coherent with production plants



Industrial areas dataset



## Sizing the need (2030 example)

Number of different stations needed

region	quantity_h2_m	max_quantity_h2_proposal	small	medium	large	total
0 Auvergne-Rhône-Alpes	41881	42000	4	6	2	16
1 Bourgogne-Franche-Comté	17826	18000	4	2	1	7
2 Bretagne	14179	8000	1	2	0	3
3 Centre-Val de Loire	7985	8000	1	2	0	3
4 Grand Est	11291	18000	4	2	1	7
5 Hauts-de-France	9722	10000	2	2	0	4
6 Normandie	9338	10000	2	2	0	4
7 Nouvelle-Aquitaine	20752	21000	4	3	1	8
8 Occitanie	19783	18000	3	2	1	6
9 Pays de la Loire	6961	7500	2	1	0	3
10 Provence-Alpes-Côte d'Azur	8162	10000	2	2	0	4
11 Île-de-France	227183	23400	3	3	1	7
12 Total	186217	191000	38	29	7	74

H2 requirements per station (at profitability threshold)

		Small - 1 tpd	Medium - 2 tpd	Large - 4 tpd
Capex (installed equipment)	m€	3	5	8
Depreciation	Y	15	15	15
Yearly Opex	% of Capex	10	8	7
Storage Onsite	m3	2	3	4
Construction time	Y	1	1	1
Footprint	m2	650	900	1200
Profitability threshold	load (%)	90	80	60

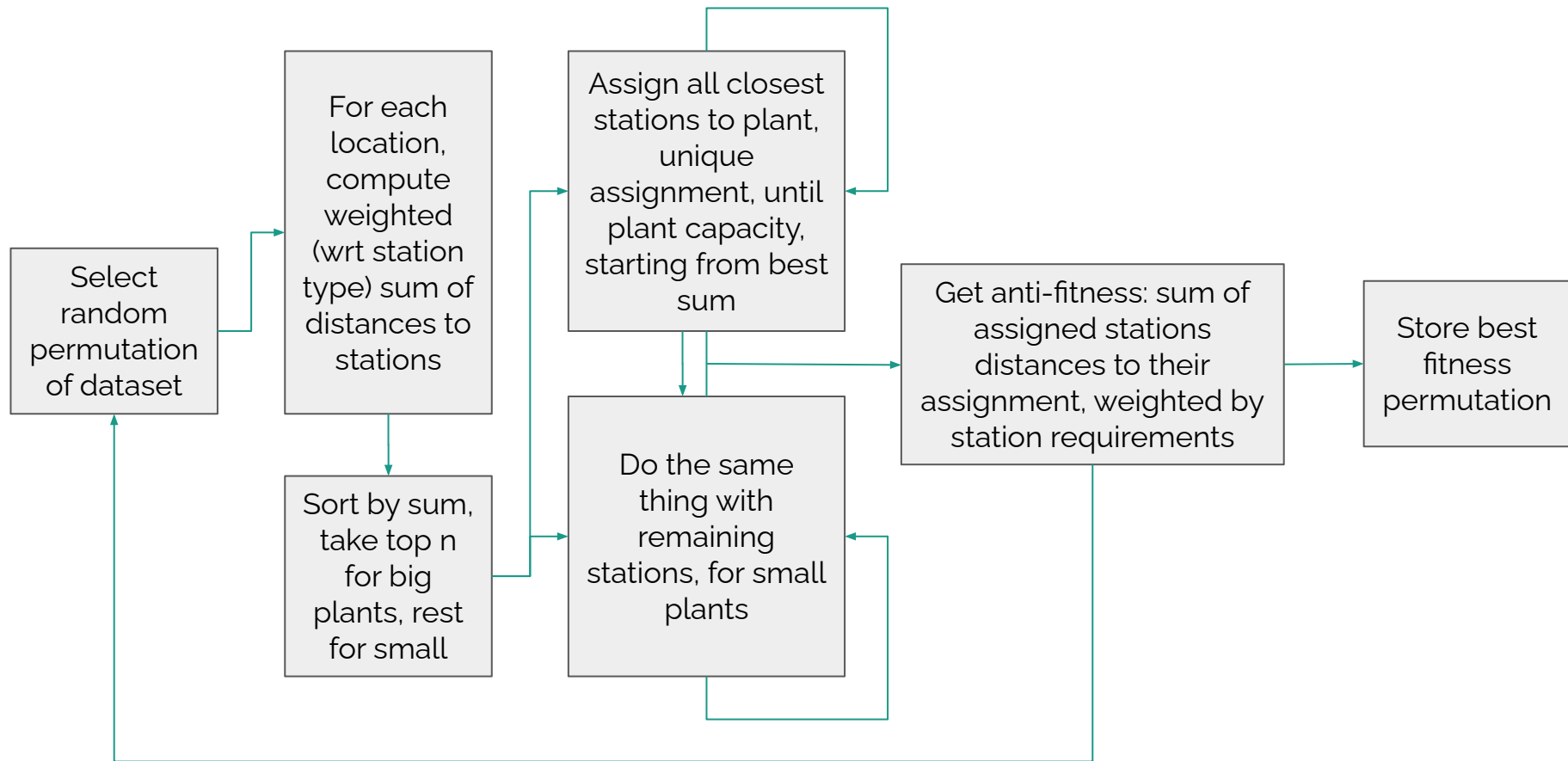
Costs and production per H2 plant

		Small - 5MW	Large - 100MW
Capex (installed equipment)	m€	20	120
Depreciation	Y	15	15
Opex (O&M, etc)	% of Capex	3	3
Power usage	kWh/kgH2	55	50
Water consumption	L/kgH2	10	10



Optimal Composition: **3 big and 5 small stations**

# A more complex permutation-based model



# Results and analysis



Red are large production plants, green are small

- ~ 600 potential locations
- Performed on 2030 stations dataset  
→ 74 stations across the country
- Find best locations according to our strategy

	e1	center	big	lon	lat
0	595.0	(476894.29758616723, 6852370.098878635)	1.0	-0.034219	48.732871
1	2431.0	(348110.12465595786, 6785919.327932795)	1.0	-1.728165	48.078805
2	1553.0	(459195.33880018943, 6706610.073236429)	1.0	-0.194512	47.415993
3	1171.0	(759470.3392490665, 6259114.507366363)	0.0	3.734119	43.428653
4	693.0	(409641.850507698, 6260182.298579458)	0.0	-0.582506	43.381479
5	888.0	(693016.0716674337, 7061847.219311408)	0.0	2.901414	50.653500
6	139.0	(879871.0807835811, 6259419.916361741)	0.0	5.219965	43.411138
7	856.0	(207907.98133333336, 6772254.596)	0.0	-3.589191	47.865792

Cost-efficiency based analysis

Station - plant mapping

Coherence with industrial locations

Very Computationally expensive : could only study ~4000 permutations in ~2h

Would require optimization or larger computational resources