

# Around The World in 80 Days

Programming with Python
Master in Data science for economics,
business, finance

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## Aim of the project

Like a new Phileas Fogg you have the desire to travel around the world always moving east, could you do it in 80 days?

The aim of the project is to help you and show you the best way to fulfill your dream, with the method *travel* of the *AroundTheWorld* class.

The starting point is the city of London (GB), but it could be the one you want.



#### Parameters of AroundTheWorld

- dataframe : Dataset of all cities
- city\_start : Name of the starting city
- country\_start : Name of the starting country
- n\_min: Number of the closest cities to which it is possible to travel
- x\_size: Size of the longitudinal side of the grid used to search for the nearest cities
- y\_size: Size of the latitudinal side of the grid used to search for the nearest cities
- rise\_factor: Multiplication factor to increase the grid used to search for the nearest cities

# **Input Data**

#### A dataset with 26569 cities of the world

	city	city_ascii	lat	Ing	country	iso2	iso3	admin_name	capital	population	id
0	Tokyo	Tokyo	35.6897	139.6922	Japan	JP	JPN	Tōkyō	primary	37977000.0	1392685764
1	Jakarta	Jakarta	-6.2146	106.8451	Indonesia	ID	IDN	Jakarta	primary	34540000.0	1360771077
2	Delhi	Delhi	28.6600	77.2300	India	IN	IND	Delhi	admin	29617000.0	1356872604
3	Mumbai	Mumbai	18.9667	72.8333	India	IN	IND	Mahārāshtra	admin	23355000.0	1356226629
4	Manila	Manila	14.5958	120.9772	Philippines	PH	PHL	Manila	primary	23088000.0	1608618140
5	Shanghai	Shanghai	31.1667	121.4667	China	CN	CHN	Shanghai	admin	22120000.0	1156073548
6	São Paulo	Sao Paulo	-23.5504	-46.6339	Brazil	BR	BRA	São Paulo	admin	22046000.0	1076532519
7	Seoul	Seoul	37.5833	127.0000	Korea, South	KR	KOR	Seoul	primary	21794000.0	1410836482
8	Mexico City	Mexico City	19.4333	-99.1333	Mexico	MX	MEX	Ciudad de México	primary	20996000.0	1484247881
9	Guangzhou	Guangzhou	23.1288	113.2590	China	CN	CHN	Guangdong	admin	20902000.0	1156237133

#### **DataFrame definition**

city	lat	Ing	country	iso2	population	flg_pop	visited_city
Tokyo	35.6897	139.6922	Japan	JP	37977000.0	1	0
Jakarta	-6.2146	106.8451	Indonesia	ID	34540000.0	1	0
Delhi	28.6600	77.2300	India	IN	29617000.0	1	0
Mumbai	18.9667	72.8333	India	IN	23355000.0	1	0
Manila	14.5958	120.9772	Philippines	PH	23088000.0	1	0
Nord	81.7166	-17.8000	Greenland	GL	10.0	0	0
Timmiarmiut	62.5333	-42.2167	Greenland	GL	10.0	0	0
heremoshna	51.3894	30.0989	Ukraine	UA	0.0	0	0
Ambarchik	69.6510	162.3336	Russia	RU	0.0	0	0
Nordvik	74.0165	111.5100	Russia	RU	0.0	0	0
	Tokyo Jakarta Delhi Mumbai Manila Nord Timmiarmiut neremoshna Ambarchik	Tokyo 35.6897  Jakarta -6.2146  Delhi 28.6600  Mumbai 18.9667  Manila 14.5958   Nord 81.7166  Timmiarmiut 62.5333  neremoshna 51.3894  Ambarchik 69.6510	Tokyo 35.6897 139.6922  Jakarta -6.2146 106.8451  Delhi 28.6600 77.2300  Mumbai 18.9667 72.8333  Manila 14.5958 120.9772   Nord 81.7166 -17.8000  Timmiarmiut 62.5333 -42.2167  neremoshna 51.3894 30.0989  Ambarchik 69.6510 162.3336	Tokyo         35.6897         139.6922         Japan           Jakarta         -6.2146         106.8451         Indonesia           Delhi         28.6600         77.2300         India           Mumbai         18.9667         72.8333         India           Manila         14.5958         120.9772         Philippines                 Nord         81.7166         -17.8000         Greenland           Timmiarmiut         62.5333         -42.2167         Greenland           neremoshna         51.3894         30.0989         Ukraine           Ambarchik         69.6510         162.3336         Russia	Tokyo         35.6897         139.6922         Japan         JP           Jakarta         -6.2146         106.8451         Indonesia         ID           Delhi         28.6600         77.2300         India         IN           Mumbai         18.9667         72.8333         India         IN           Manila         14.5958         120.9772         Philippines         PH                  Nord         81.7166         -17.8000         Greenland         GL           Timmiarmiut         62.5333         -42.2167         Greenland         GL           neremoshna         51.3894         30.0989         Ukraine         UA           Ambarchik         69.6510         162.3336         Russia         RU	Tokyo         35.6897         139.6922         Japan         JP         37977000.0           Jakarta         -6.2146         106.8451         Indonesia         ID         34540000.0           Delhi         28.6600         77.2300         India         IN         29617000.0           Mumbai         18.9667         72.8333         India         IN         23355000.0           Manila         14.5958         120.9772         Philippines         PH         23088000.0                    Nord         81.7166         -17.8000         Greenland         GL         10.0           Timmiarmiut         62.5333         -42.2167         Greenland         GL         10.0           neremoshna         51.3894         30.0989         Ukraine         UA         0.0           Ambarchik         69.6510         162.3336         Russia         RU         0.0	Tokyo         35.6897         139.6922         Japan         JP         37977000.0         1           Jakarta         -6.2146         106.8451         Indonesia         ID         34540000.0         1           Delhi         28.6600         77.2300         India         IN         29617000.0         1           Mumbai         18.9667         72.8333         India         IN         23355000.0         1           Manila         14.5958         120.9772         Philippines         PH         23088000.0         1                    Nord         81.7166         -17.8000         Greenland         GL         10.0         0           Timmiarmiut         62.5333         -42.2167         Greenland         GL         10.0         0           neremoshna         51.3894         30.0989         Ukraine         UA         0.0         0           Ambarchik         69.6510         162.3336         Russia         RU         0.0         0

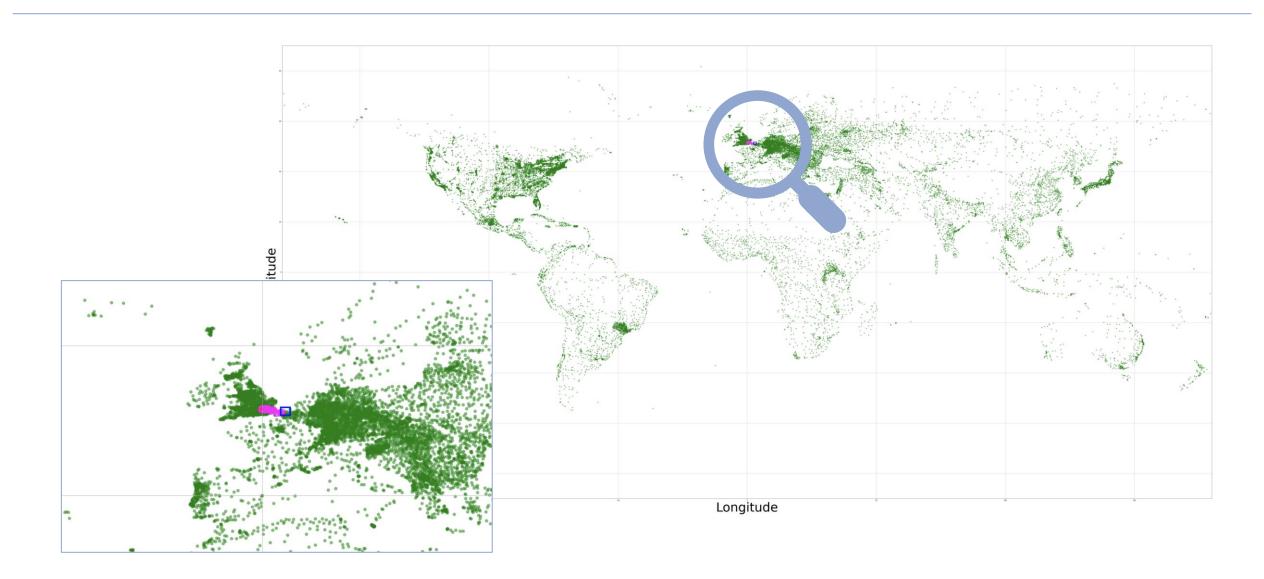
#### **Grid definition**

#### Move always towards the east

At each step, starting from the city in which it is located, the algorithm calculates a rectangle:

- **base**: the distance between the longitude of the current city and a subsequent point at a variable distance given as input (x\_size).
- **height**: changes depending on whether the current city is north (latitude is greater) or south (latitude is lower) of the starting city:
  - North: height is the distance generated between the latitude of the current city, adding a
    quantity (y\_size/2) and the latitude of the starting city, subtracting a quantity (y\_size/2)
  - South: height is the distance generated between the latitude of the starting city, adding a quantity (y\_size/2) and the latitude of the current city, subtracting a quantity (y\_size/2)

#### **Grid visualization**



### Weight assignment criteria

- **Distance:** at each step, according to the increasing Euclidean distance, the 3 closest cities are assigned values 2, 4, 8 respectively.
- **Population**: a weight of value 2 is added if the city has a population greater than or equal to 200 thousand inhabitants.
- **Country**: a weight of value 2 is added if the city in the next step is located in a different country than the previous one.

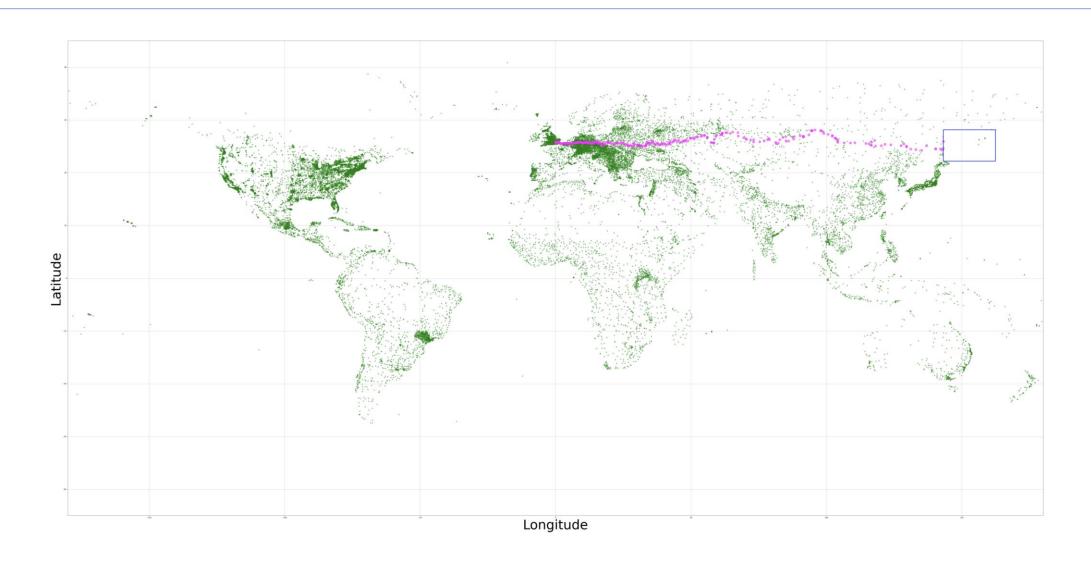
#### **Grid variability**

What happens when Phileas must cross oceans?

The size of the polygon considered at each step is variable to ensure the functionality of the algorithm with a minimum number of at least **3** cities.

The absence in the rectangle of at least 3 cities makes these dimensions vary by a multiplicative **rise\_factor**.

# **Grid variability visualization**



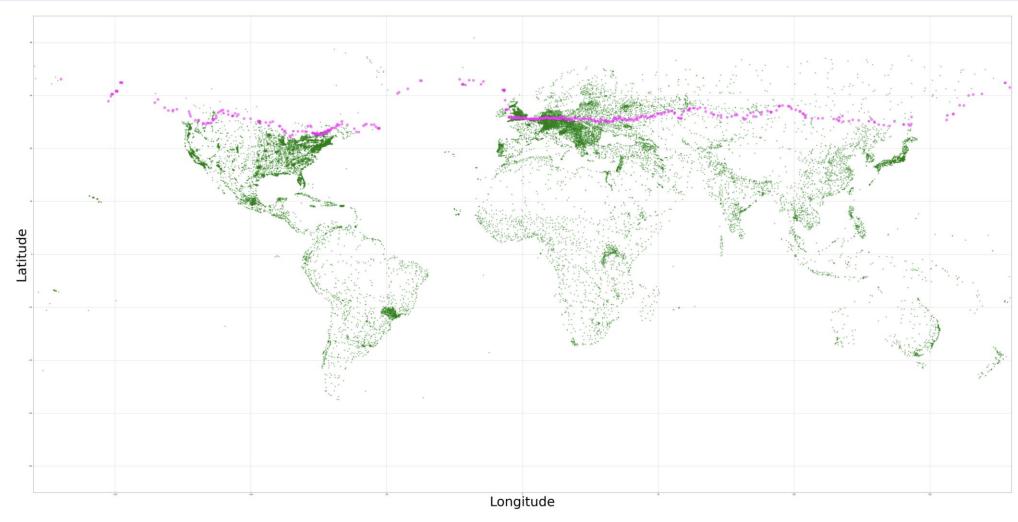
### **Stop criteria**

If among the 3 nearest cities the starting city is present, the algorithm chooses that one, adding up the weight it has.

Once Phileas returns to London, the algorithm returns:

- the number of hours taken to complete the journey (sum of the weights of all the steps)
- the total number of steps
- the dataframe of all the cities visited

#### **Path**



Completed the journey starting from London (GB) in 62.42 days (1498 hours) after visited 666 cities. Done in 20.59 seconds.

# **Visited City DataFrame**

	city	lat	Ing	country	iso2
6681	Holborn	51.5172	-0.1182	United Kingdom	GB
5977	Highbury	51.5520	-0.0970	United Kingdom	GB
7342	Spitalfields	51.5166	-0.0750	United Kingdom	GB
6407	Stepney	51.5152	-0.0462	United Kingdom	GB
6559	Hackney	51.5414	-0.0266	United Kingdom	GB
7531	Raynes Park	51.4033	-0.2321	United Kingdom	GB
5045	Wimbledon	51.4220	-0.2080	United Kingdom	GB
5446	Morden	51.4015	-0.1949	United Kingdom	GB
5080	Mitcham	51.4009	-0.1517	United Kingdom	GB
34	London	51.5072	-0.1275	United Kingdom	GB

666 rows × 5 columns

# Thanks for your attention

