French Salaries [Aka companies in french cities]

Angela Salgado and Jaime Sastre Ironhack Paris 11/2022 Project week 4







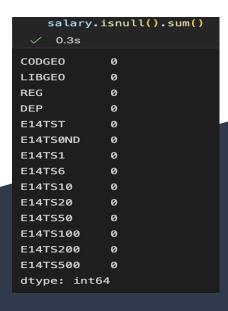
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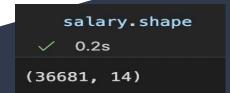
- 1. Dataset description
- 2. Challenges
- 3 Project process
- 4. Learnings
- 5. Improvements
- 6. Comparison of the initial dataset and the final one
- 7. Highlights



1. Dataset Description

	salary.head()														
	′ 0.2s														
	CODGEO	LIBGEO	REG	DEP	E14TST	E14TSOND	E14TS1	E14TS6	E14TS10	E14TS20	E14TS50	E14TS100	E14TS200	E14TS500	
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1	01002	L'Abergement-de-Varey	82	01	10	9	1	0	0	0	0	0	0	0	





	REG	E14TST	E14TSOND	E14TS1	E14TS6	E14TS10	E14TS20	E14TS50	E14TS100	E14TS20) E14TS	500							7
REG	1.000000	-0.002858	-0.002906	-0.001119	-0.005153	-0.005035	-0.007237	-0.008729	-0.015250	-0.02046	0 -0.023	3357							-
E14TST	-0.002858	1.000000	0.999489	0.996461	0.995297	0.993956	0.984406	0.969626	0.962369	0.94397	0.92	4913				1			
E14TS0ND	-0.002906	0.999489	1.000000	0.993396	0.991979	0.990579	0.979331	0.963099	0.956575	0.93824	0.92 ⁻	1855							
E14TS1	-0.001119	0.996461	0.993396	1.000000	0.998088	0.996207	0.990042	0.978388	0.968586	0.94923	0.92	3146					7		
E14TS6	-0.005153	0.995297	0.991979	0.998088	1.000000	0.998597	0.994095	0.983196	0.973983	0.95595	1 0.929	809					· \		
E14TS10	-0.005035	0.993956	0.990579	0.996207	0.998597	1.000000	0.995787	0.985659	0.977239	0.95989	1 0.934	1023					4	مر	
E14TS20	-0.007237	0.984406	0.979331	0.990042	0.994095	0.995787	1.000000	0.992583	0.985338	0.97029	0.940)543						~~\	
E14TS50	-0.008729	0.969626	0.963099	0.978388	0.983196	0.985659	0.992583	1.000000	0.987325	0.97496	0.94	3851							
E14TS100	-0.015250	0.962369	0.956575	0.968586	0.973983	0.977239	0.985338	0.987325	1.000000	0.97984	0.955	5842							
E14TS200	-0.020460	0.943975	0.938243	0.949235	0.955951	0.959891	0.970295	0.974962	0.979842	1.00000	0.96	1836							
E14TS500	-0.023357	0.924913	0.921855	0.923146	0.929608	0.934023	0.940543	0.943851	0.955842	0.96183	1.000	0000							
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2. Challenges



Find what to analyse in our charts

Look for the proper charts to display what we want

Create a column of Latitude and Longitude from the name of the town

3. Project process

- Planning on Trello
- 2. Python Code
 - a. Download dataset
 - b. Examine the dataset for the possible issues. (Too beauty to be true)
 - c. Pivot tables, groupby
 - d. Charts to display insights (tough, because we didn't have a lot of info)
- 3. Conclusions in the Readme file



4. Learnings

-We tried to draw as many inferences as possible even though the data was quite sterile.

Work with the different Python libraries for data visualization

5. Improvements

Improve data visualization. It takes a long time to build the charts, and we didn't manage to customize them to everything we wanted: colors, shapes, legend etc.

We would like to be able to incorporate data on salaries, which was initially the name of the dataset



6. Comparison of the initial dataset and the final one

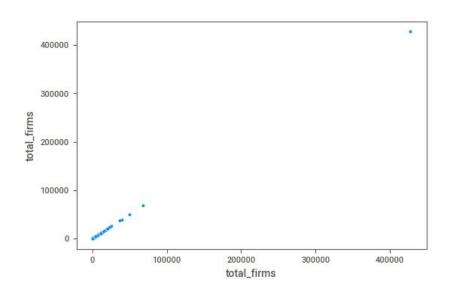
We add three columns in order to group the number of employees of the companies into three groups: small, medium and large companies

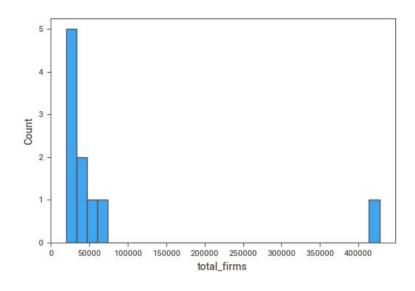
7. Highlights



- -Number of companies
- -Number of employees

Number of companies in centralized





Boxplot grouped by reg total_firms

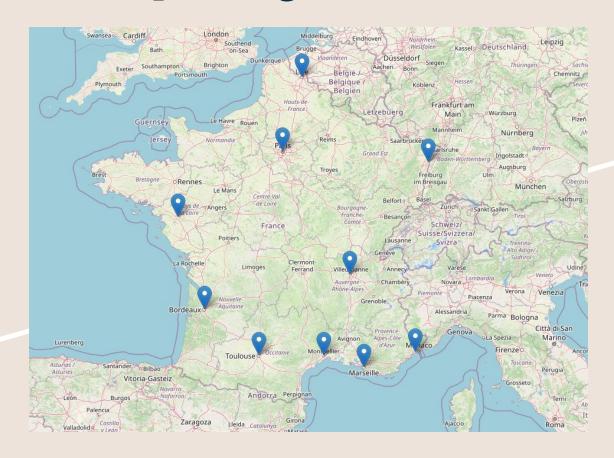


1 2 3 4 6 11 21 22 23 24 25 26 31 41 42 43 52 53 54 72 73 74 82 83 91 93 94 reg

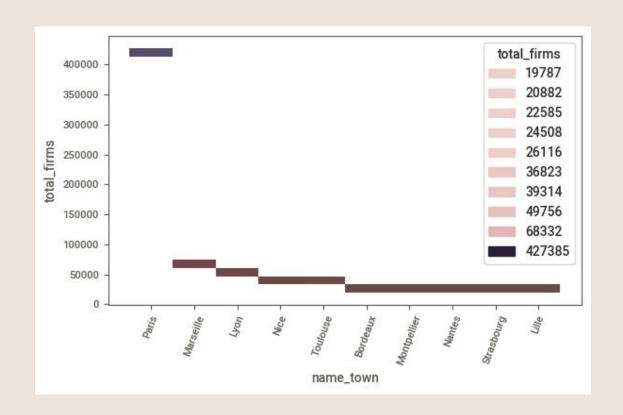
The location of companies is dramatically centralized in Paris.



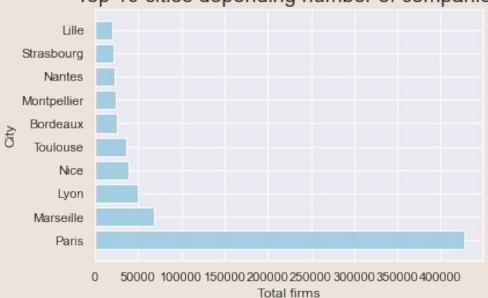
Top 10 cities depending of numbers of companies

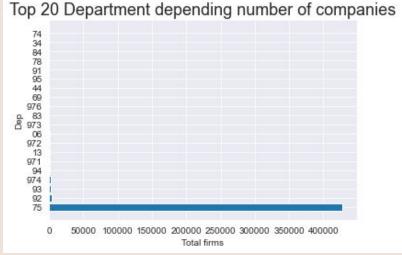


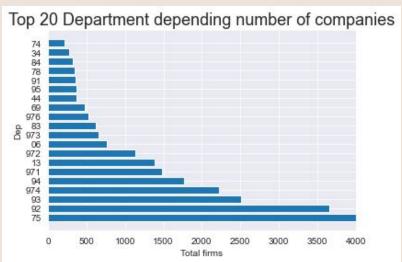
Paris has 7 times more companies than the second city with the most companies in France



Top 10 cities depending number of companies







To the centralization of the number of companies in Paris, we have also added the dynamics of 93 and 92 departments, which concentrate high numbers of companies.

Number of employees

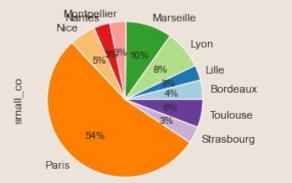
Small companies = 1-19 employees

Medium companies = 20 - 199

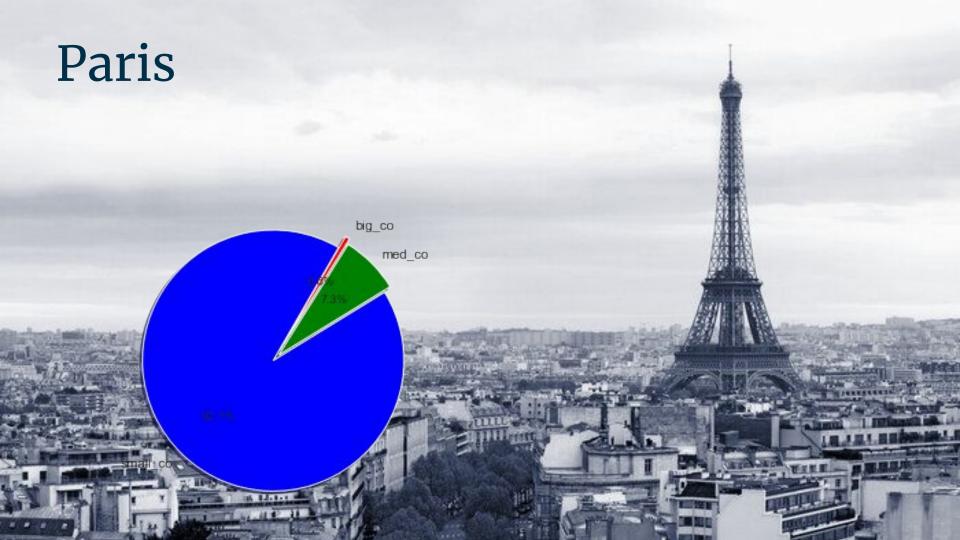
Big companies = 200



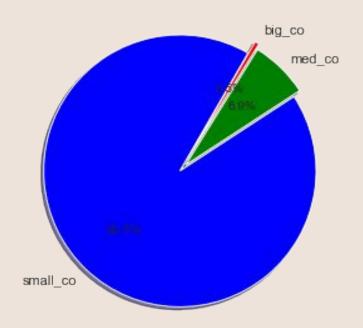




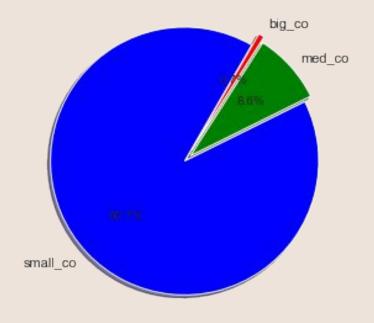




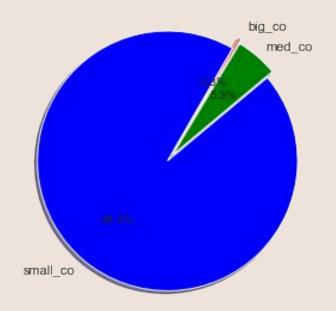
Marseille



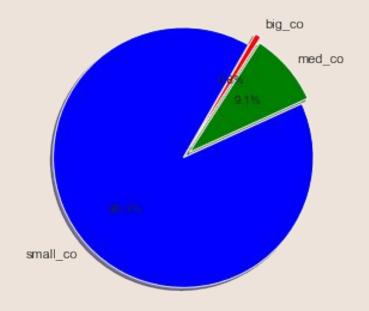
Lyon

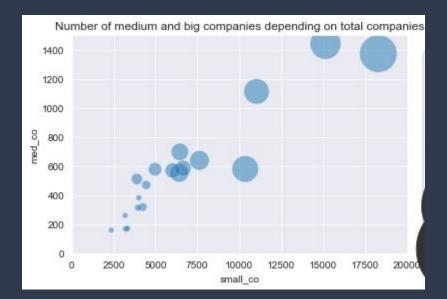


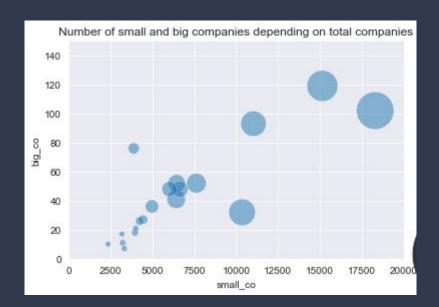
Nice



Toulouse







Q&A:)