Data and slides

http://bit.ly/4ms5lci



Is my house at risk of flooding? A data-driven investigation to track buildings in flood-prone zones

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- 1. Summary of our investigation
- 2. What is spatial data and why we use it
- 3. How did we find the data
- 4. Difficulties
- 5. How can be **replicated** (for any research)
- 6. The process
- 7. Hands on



We found 200,000 buildings constructed in areas with a high risk of flooding in Spain

Why did we decide to do this project?

How did we carry out this data-driven investigation?

What kind of data did we use?

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How did we find and analyze the data?





Try to make the most of our data

With the same data we have released a special with 8 reports



Why did we decide to do this project?





Why did we decide to do this project?



October 29, 2024

Isolated Depression at High Levels.

One of the greatest natural disasters in Spain.

227 deceased.

Thousands of homes affected.



Deaths in Spain by type of natural disaster

Data between 1995 and 2015

Floods	410
High temperatures	347
Coastal storms	284
Forest fires	157
Strong winds	145
Avalanches	62
Storms and lightning	59
Landslides	51
Snow and cold episodes	50
Earthquakes	9

Gráfico: Newtral.es • Fuente: Compilation of data from Civil Protection • Creado con Datawrapper



Our hypothesis



Floods in Spain are recurrent and are the natural disaster that affects the country the most.

Just like in Valencia, there must be many homes at risk of flooding throughout the country. Quantify the number of homes at risk of flooding



Information which describes objects with a specific location on Earth

- Points
- Lines
- Polygons



It allows you to verify data, uncover relevant information and tell engaging stories about the environment.



How did we find the data?



Flood prone zones

Ministry for Ecological Transition and the Demographic Challenge

Cadastral data

General Directorate of Cadastre of Spain



Difficulties we face



Difficulties we face

- 1. Unify the data
- 2. Handle large volumes of data
- 3. Optimize data to be able to visualize it
- 4. Reduce costs



Low: over 60m/pixel Medium: 10 – 30m/pixel High: 3m – 5m/pixel Very high: less than 1m/pixel



How can it be **replicated**?



How can it be **replicated**?

Spatial join



Different from traditional 'joins'.

It compares two geospatial files based on the spatial relationship of their geometries.













1 2 3 4 5 7 8

parcels_flooded = gpd.sjoin(parcels, flood_zones, how='inner', predicate='intersects')
log_message(f"Parcels in flood zones found in {parcel_path}: {len(parcels_flooded)}")



What tools are needed?



What tools are needed?







The process

Transform

Analyse and cleaning

Spatial Join

Optimize

Visualize

Transform the raw files downloaded from official sources, which are typically in Shapefile or GML formats, into more manageable formats such as GeoJSON. Python is used in this step. Once we have the files in the desired format and organized into folders, we explore and clean the files (removing unwanted columns or omitting errors).

In this way, we can reduce the file size by half. In this step, we also use Python. With the clean and streamlined data, we proceed to perform a spatial join between two geospatial data files.

This allows us to identify and extract the matching geometries.

Python is also used in this step.

The results are combined into one, two, or three files (depending on the size). Both the initial cadastral data and the geometries isolated from the spatial join need to be processed again.

This process is done with Tippecanoe, and the result is files in Mbtiles format. These Mbtiles files are now ready to be visualized in an application like Mapbox. We upload them from Mapbox Studio, and that's it!

For a more professional touch, the map can be created using HTML and JavaScript code.



Our objective



https://goo.su/D2vOTb



