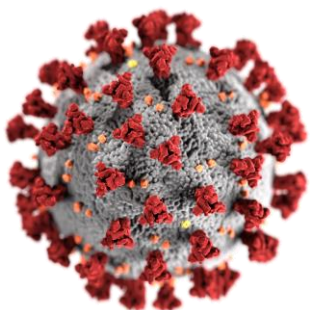


VU Visualisierung 2 (186.833)

Spread of Covid-19



Arlind Avdullahi 11813054

Princ Mullatahiri 11846033



- Our dataset consists of data of Covid19 spread across world.
- Source: Kaggle
- It contains number of cases per country from 22th of January until 16th of April.
 - First case in Europe 24th of January
 - As of 17th of March all countries in Europe have confirmed cases
 - Until 16th of April there are 965488 confirmed cases in Europe
- Questions we plan to answer:
 - How well does the technique in the paper visualize the spread trajectories of Covid-19? Can we optimize it?



Technique Summary

- Their technique approximates the underlying distribution of non-directional data over time through the application of a 2D kernel density estimation.

- $$f_{2D}(x, y) = \frac{1}{N} \sum_{i=1}^N \frac{1}{h^2} K\left(\frac{x-x_i}{h}, \frac{y-y_i}{h}\right),$$

- N – number of samples, h- bandwidth of kernel, (x,y) – location longitude and latitude.

1. Functional representation of spatiotemporal data (using KDE)
2. Flow map extraction (gravity-based flow extraction model)
3. Visualization

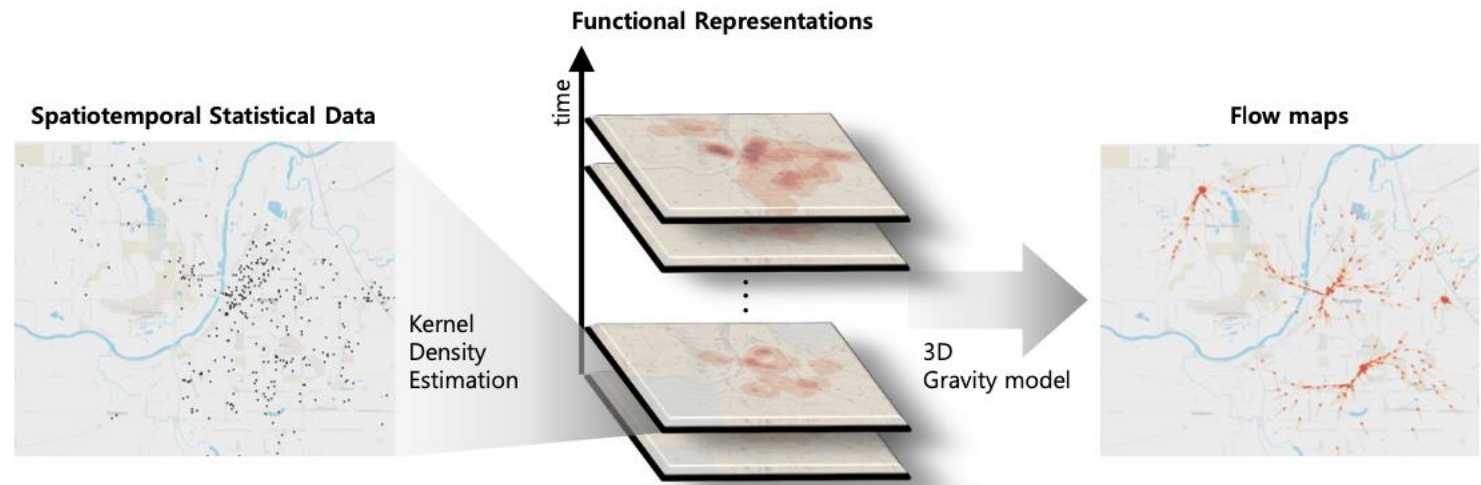


Fig 1: Visual overview of the system



- We focus our implementation in Europe and ignore the data we have on other countries.
- For visualization we use plotly library.
- As for 2D KDE we need to reimplement a weighted 2D KDE.
 - We reimplement `scipy.stat.gaussian_kde` to weighted kde
 - For comparison we also use `kdeplot` from seaborn package
- We might need to use Gaussian Kernel instead of Triweight Kernel because there are much more information for 2D Gaussian KDE.



Seokyeon Kim, Seongmin Jeong, Insoo Woo, Yun Jang, Ross
Maciejewski, David S. Ebert

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