

# Spatio-temporal Public Health Analysis and its Ethical Concerns

Debjyoti Paul  
University of Utah  
deb@cs.utah.edu

## INTRODUCTION

Many research has revealed that analyzing tweets in volume can measure different population characteristics, including public health measures [1–4, 6, 7]. Research analysis like correlating influenza rates w.r.t geography (spatial) and time [8], state level food and health behavior analysis [5], predicting heart disease rate mortality rate based on twitter information [2]; are motivating examples to carry out such analysis for improving and create for public health. All these above adhoc analysis inspire us to build a general system for comprehensive analysis. In this work, I will present overview of architecture and desired features to build such system or tools.

## 1 PART A: SPATIO-TEMPORAL SYSTEM ARCHITECTURE FOR HEALTH ANALYSIS:

A comprehensive system for spatio-temporal analysis requires the following components which can be broadly categorized based on their operations:

- **Data Ingestion**
  - *Data Collection Module*
- **Data Enrichment**
  - *Data Cleaning Module*
  - *Location Extraction Module*
- **AI/ML Models**
  - *Tweet/Document Classification Module*
  - *Sentiment Analysis Module*
  - *Image Classification Module (optional)*
- **Data Storage**
- **Analytics Processing Engine:**
  - *Realtime Data Aggregation Support*
  - *Spatio-temporal Query Support*
- **Visualization**
  - *Realtime Dashboard*
  - *Dynamic Data Visualization Module*

## 2 PART B. PUBLIC HEALTH ANALYSIS AND ETHICAL CONCERNS:

[? ]

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