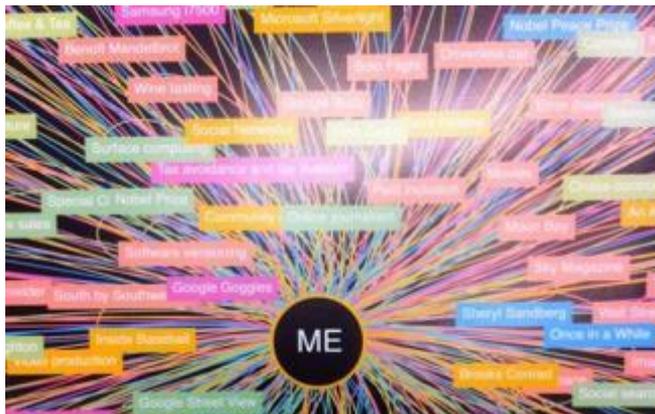


# Situation-awareness in social overlays

Hariton Efstathiades

# Research Focus

- ▣ **Situation-aware overlay:** Each individual is connected with the source of information, according to the current situation



# Scenario - Vision

## User



**Andres Ledesma**

Researcher at University of Cyprus  
Cyprus | Computer Software



**Andres Ledesma** @wizardseal

See you in #stockholm for #iSocial summer school (31st of May - 5th of June)

### Andres Ledesma's Skills & Expertise

- C
- C++
- Java
- Python
- Programming
- Computer Science
- Software Engineering
- Eclipse
- Linux
- Algorithms
- LaTeX
- C#
- SQL
- JavaScript
- Software Architectural Design

## Social web data



**Nick Thibieroz** @NThibieroz · May 6

AMD and Microsoft are presenting on advanced graphics topics in a free Stockholm developer event on Monday 2 June: [eventbrite.com/e/amd-microsof...](http://eventbrite.com/e/amd-microsof...)



**Anatoliy Gromov** @agromov · May 27

Hey crowd! I've got two ticket for Aerosmith in the 1st of June and John Mayer on 12th of June in Stockholm, giving out half-priced, DM

Introduction

LinkedIn Analysis

Twitter Analysis

Summary

Next steps



# Highlighted Related Research

- Berlingerio, Michele, et al., "**Safer City: A system for Detecting and Analyzing Incidents from Social Media**", IEEE International Conference on Data Mining, December, 2013.
- Gupta, Amarnath, and Ramesh Jain. "**Social life networks: a multimedia problem?**", *Proceedings of the 21st ACM international conference on Multimedia*, October 2013.
- Abel, Fabian, et al. "**Semantics+ filtering+ search= twitcident. exploring information in social web streams**", Proceedings of the 23rd ACM conference on Hypertext and social media, June 2012.
- Adriana Iamnitchi, et al., "**The Social Hourglass: An Infrastructure for Socially Aware Applications and Services**", IEEE Internet Computing, May-June, 2012

Introduction

LinkedIn Analysis

Twitter Analysis

Summary

Next steps



# Practical Issues

## ▣ Limited attention



© iStockphoto.com/Stephen Morris

- ▣ User is not able to follow everyone
- ▣ Not interested in the entire social stream of the publisher

Introduction

LinkedIn Analysis

Twitter Analysis

Summary

Next steps



# Practical Issues

## □ Limited discovery



□ Only community can discover published info

Introduction

LinkedIn Analysis

Twitter Analysis

Summary

Next steps



# Practical Issues

## ■ Fast data



- Information gets obsolete fast
- User's environment change fast

Introduction

LinkedIn Analysis

Twitter Analysis

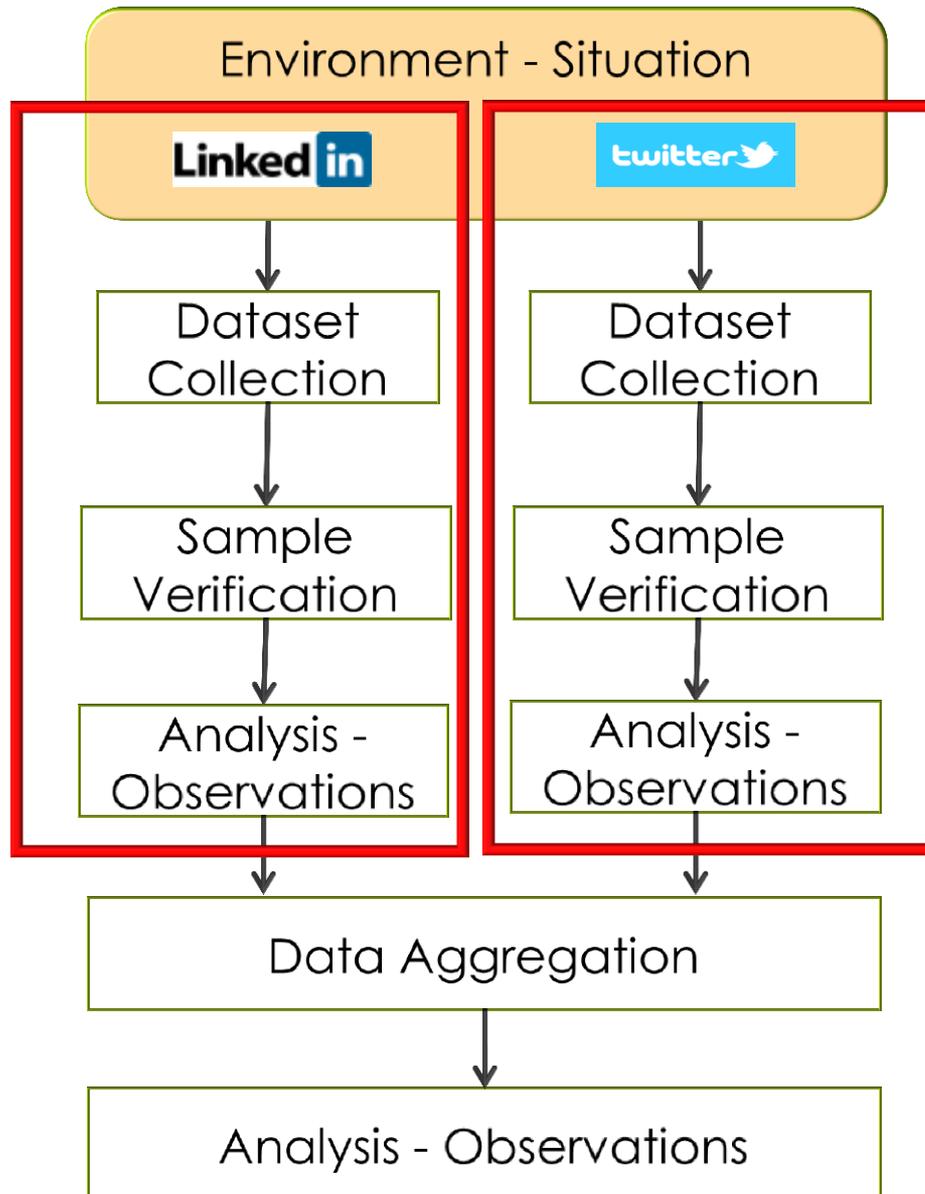
Summary

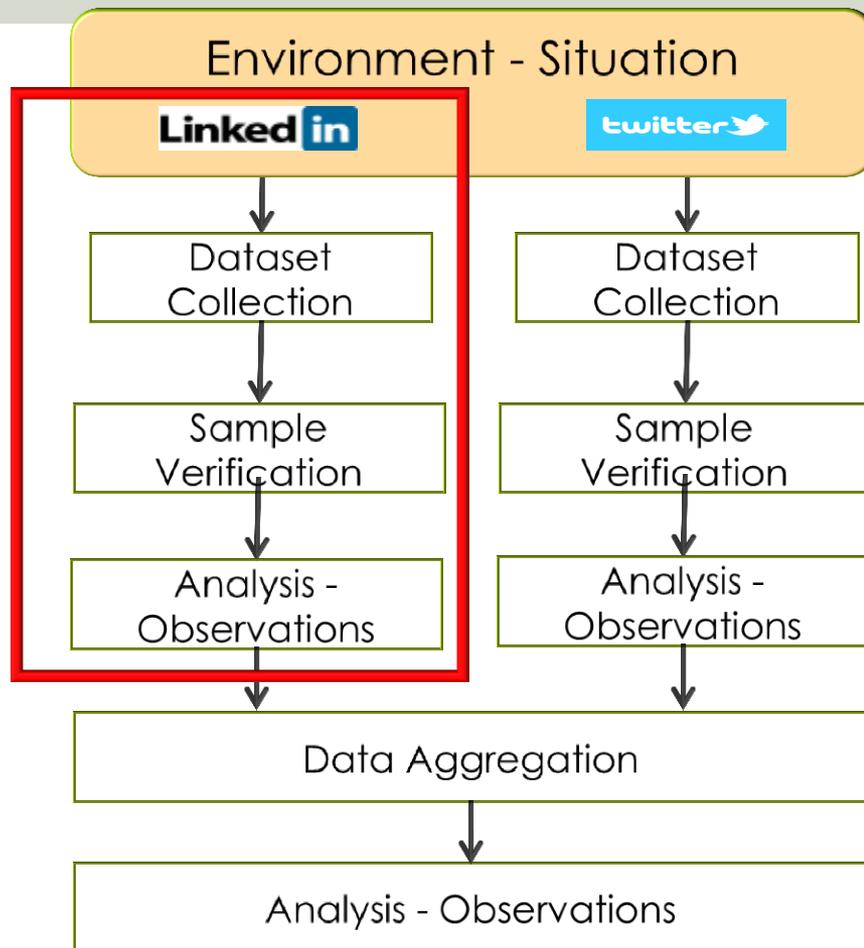
Next steps



# Today's presentation

# Today's presentation





# LinkedIn analysis

Business-market knowledge extraction from online social networks

# Dataset Collection

# LinkedIn Analysis: Problems



- Existing APIs have many limitations
- LinkedIn API limitations
  - User oriented: Need user to give permissions
- Without users' permissions
  - Access to basic profile fields (name, headline, community size)
  - Throttle limits
    - Search for user
    - Access resource
    - Access company profile
- **The public information that we can find on the website is much more than the information that we can retrieve from API!**

# LinkedIn Analysis: Solution



- Development of LinkedIn crawler
  - “Web Scraping” method
- Simulates visitor’s behavior
- Reads the tags and extracts the data



Introduction

LinkedIn Analysis

Twitter Analysis

Summary

Next steps



# LinkedIn Analysis: Crawler



- Two parts:
  - Directories collector
  - Users collector



Introduction

LinkedIn Analysis

Twitter Analysis

Summary

Next steps



# Dataset Collection tool



- Directories collector
  - OSNs have directories with the public URLs of their users
  - The tool is able to visit and retrieve LinkedIn's directories
    - Country-wise
    - Global directory

Introduction

LinkedIn Analysis

Twitter Analysis

Summary

Next steps



# Dataset Collection tool



## Users collector

- Each user has a public profile URL
- The tool is able to retrieve user's public profile information
  - Input: Public URL
  - Output:
    - Information about the User, experience, education, interests

## Tested on cy.linkedin.com

- 119,817 profiles

## Collection in progress

- Netherlands, Portugal

Introduction

LinkedIn Analysis

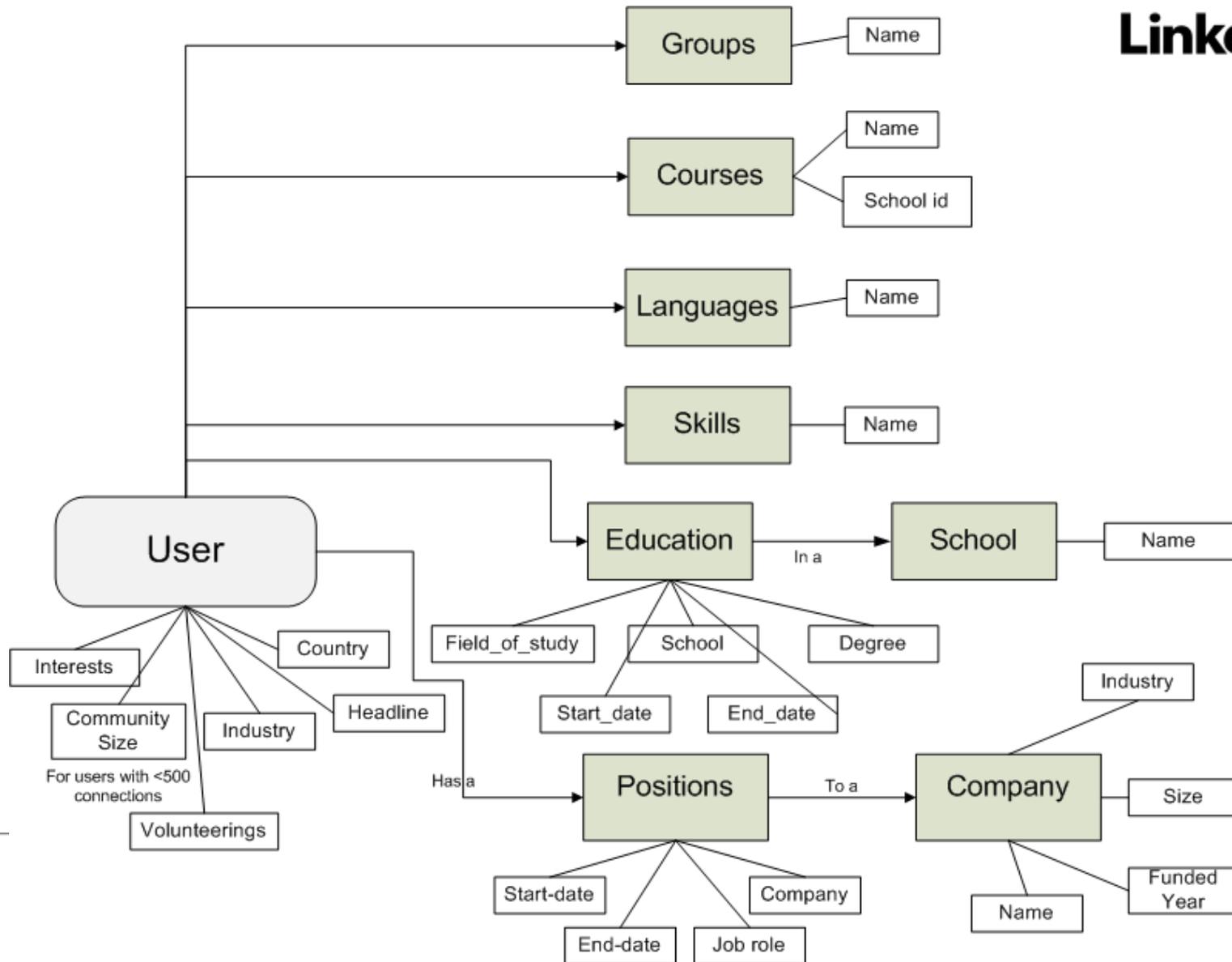
Twitter Analysis

Summary

Next steps



# Data Description





Data analysis

# Data analysis



## □ Purpose

- Infer implicit connections between users
- Identify market-business trends
  - Collaboration with Department of Management, University of Cyprus
- Anomalies detection for early warning system
- Personalized warnings based on user's background

Introduction

LinkedIn Analysis

Twitter Analysis

Summary

Next steps





Early results

# Industry "Leaves" per Year

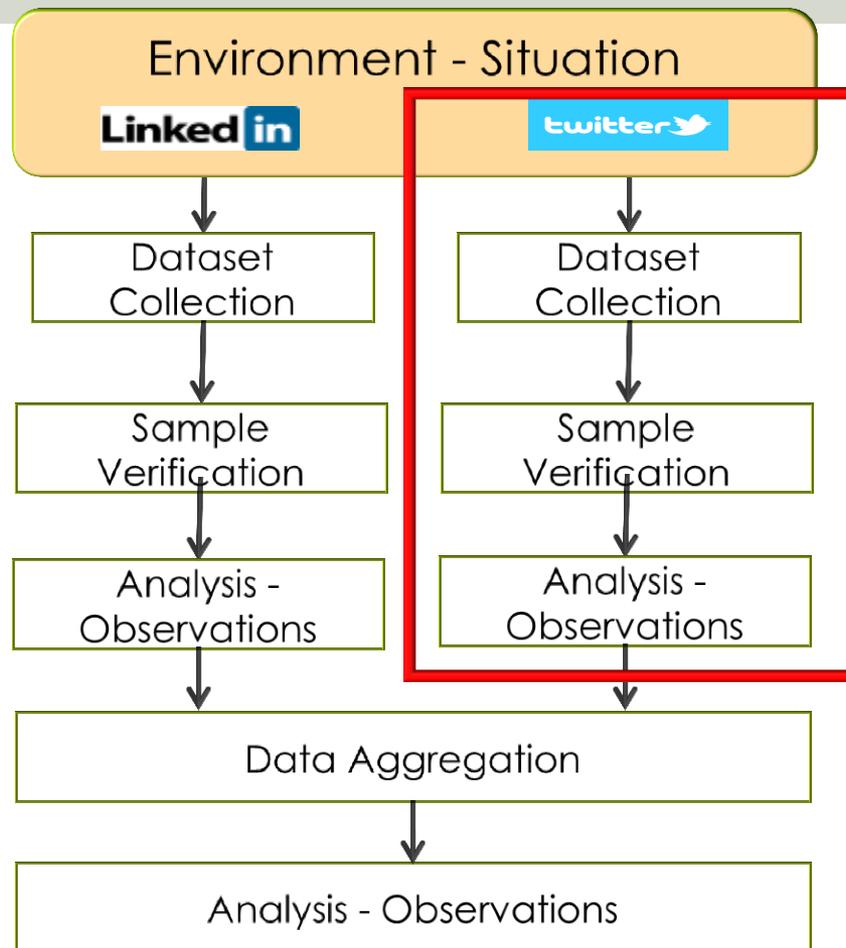


# Industry Leaves for 2013 (per month)



- Banking sector:
  - About 30% of total year leaves left in August
  - Deadline to get a reimbursement if they decide to leave
- Education:
  - **Government Policy:**
    - Temporal school teachers' contracts expire in June.
    - Renew in September





# Twitter analysis

Identification of transportation patterns based on online social networks' meta-data

# Dataset Collection

# Dataset Collection



## ▣ Problems:

- ▣ Twitter throttle limits
- ▣ We can retrieve only 1% of total Twitter stream

## ▣ Solution:

- ▣ Development of Twitter stream listeners
- ▣ Distribute different listeners in the areas that we want to collect

Introduction

LinkedIn Analysis

Twitter Analysis

Summary

Next steps



# Dataset Collection



- Collection of Twitter users from city of Amsterdam
  - Tweets history ( $\leq 3200$  tweets) for each user
  - Meta-data of each Tweet
    - Timestamp
    - Location
  - Dataset:
    - ~10,000 users
    - ~6 millions tweets
      - ~1,5 millions geo-tagged

Introduction

LinkedIn Analysis

Twitter Analysis

Summary

Next steps



# Dataset Collection



- Ground truth data from the city of Amsterdam
  - Amsterdam zones in polygons
  - Information about each zone:
    - #habitants
    - #jobs



Introduction

LinkedIn Analysis

Twitter Analysis

Summary

Next steps





# Data analysis

# Data analysis



- Purpose (Steps in progress...)
  - Infer implicit connections between users from meta-data
    - E.g. same working/leisure area
  - Identification of transportation patterns
  - Analysis of the influence of home or working location to the transportation pattern of the user
  - Events detection based on meta-data

Introduction

LinkedIn Analysis

Twitter Analysis

Summary

Next steps





# Early results

One example

# Data analysis



- Identification of Home/Working Location
  - Divide day-hours in parts (case of Netherlands)
    - Working: 09:00 – 17:30
      - Most popular area: marked as Job area
    - Home: 22:00 – 09:00
      - Most popular area
    - Leisure:
      - Most popular area except Home

Introduction

LinkedIn Analysis

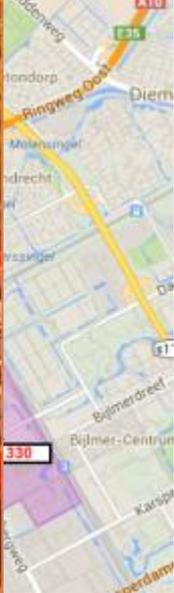
Twitter Analysis

Summary

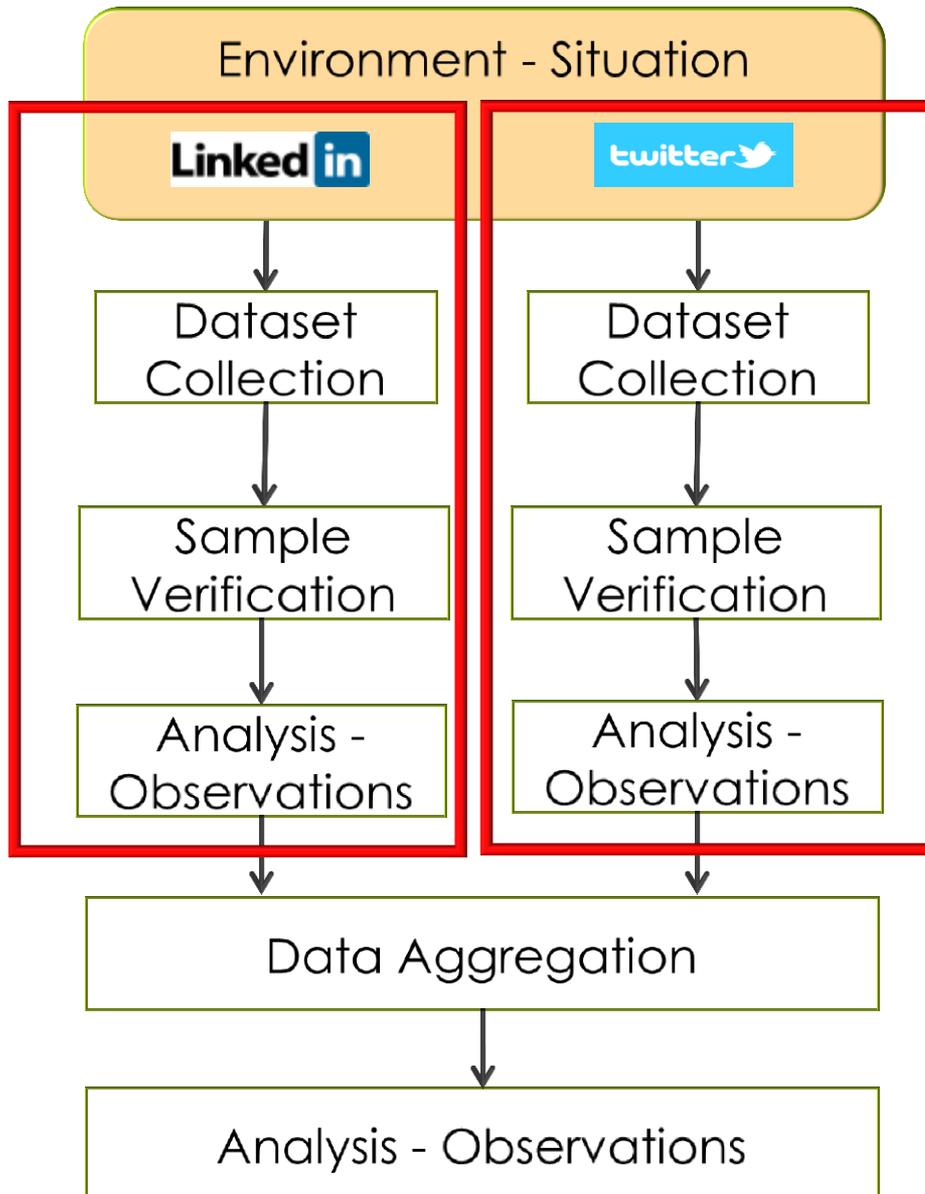
Next steps



# Data Analysis



# Summary



Next steps

# Next steps

- Extend datasets (In progress)
  - LinkedIn: Different countries
  - Twitter: More users
  - Other SNs
- Enrich users' profiles based on knowledge from social networks analysis (e.g. Unemployed, based on transportation pattern, working location etc.) (In progress)
- Identify the same user in different platforms (Early stage)
- Investigate the correlation between data that user publishes (*Twitter*) with his background (*LinkedIn*) during the reported period
- Investigate if we can model the behavior of the user based on the situation

Environment - Situation

LinkedIn

twitter



Introduction

LinkedIn Analysis

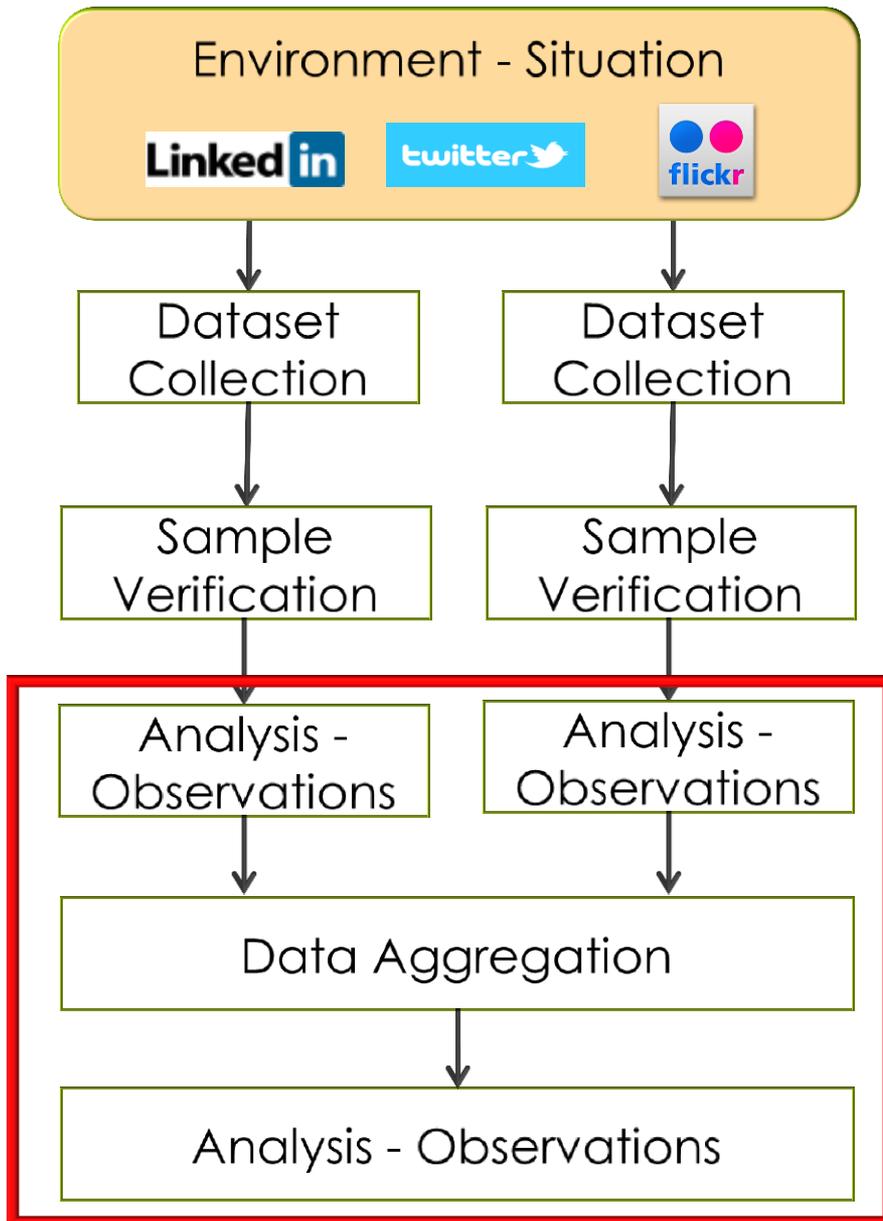
Twitter Analysis

Summary

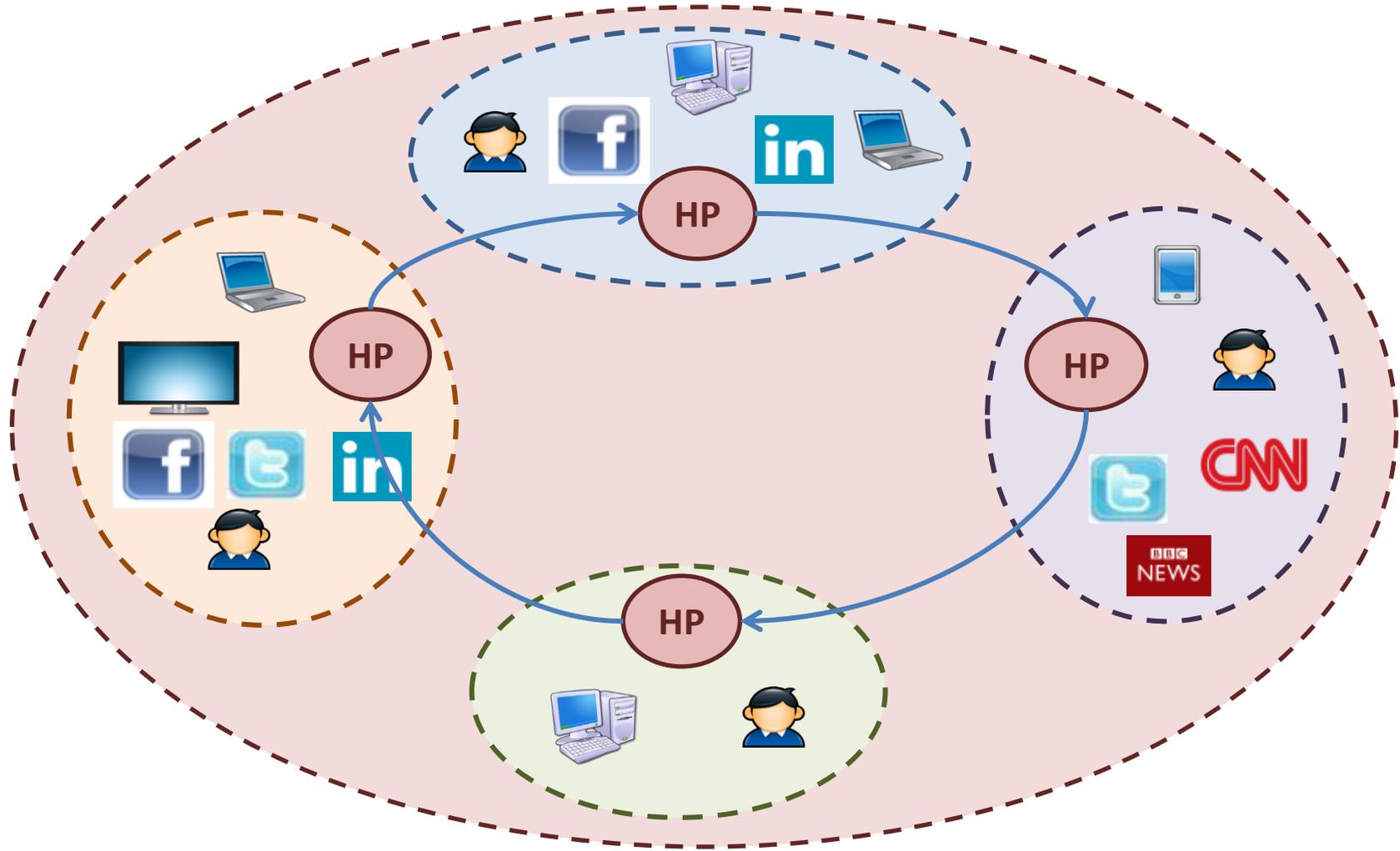
Next steps



# Next step focus



# Decentralized Vision



Questions?

