# Learning Networks of Places and People Using Location Data 

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# Sense Networks 

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## WWW




A network of online places
facebook



## From online to real networks?

What's next?

a network of real places
a network of real people

Online data is easy to get, what about the real world?

## From online to real networks?

What's next?


Online data is easy to get, what about the real world?

## GPS and location data



## GPS and location data



## Sense Networks



## CitySense: where is everyone

-Citysense: real-time density of users at every street corner -Poisson models find most active bars/restaurants


## Next: where's everyone like me

Need to have a network of people

Each dot is a user

Dot's color is user's social cluster


## Network of People

who is like whom? who colocates with whom?


## Network of People

Hard to say if User A is like User B...
User A

... don't just see if they colocate physically
... do they overlap semantically (network of places)

## Network of Places

Is place $A$ like place $B$ ?
Look at each place's Flow, Commerce \& Demographics


## Network of Places: Flow

Look at flow A to B

Markov transition

Minimum Volume
Embedding (MVE)
Color code clusters


## Network of Places: Commerce

Get each block's SIC (standard industrial categorization) code \& cluster


## Network of Places: Demographics

Get each block's census demographic data \& cluster


## Encoding people

For each user, convert GPS trail into matrix of probabilities for week hour probability of being in


1) flow cluster
2) sic cluster
3) demographic cluster

| Week <br> Hour | FLO <br> 1 | FLO <br> 2 | $\ldots$ | FLO <br> 20 | SIC <br> 1 | SIC <br> 2 | $\ldots$ | SIC <br> 97 | DEM <br> 1 | DEM <br> 2 | $\ldots$ | DEM <br> 78 |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 1 | .03 | .31 |  | .14 | .03 | .05 |  | .41 | .11 | .04 |  | .01 |
| 2 | .14 | .34 |  | .02 | .04 | .05 |  | .52 | .01 | .01 |  | .00 |
| $\ldots$ |  |  |  |  |  |  |  |  |  |  |  |  |
| 168 | .07 | .34 |  | .51 | .02 | .06 |  | .48 | .02 | .01 |  | .00 |

## Encoding people

9 example users
compute pair-wise overlap from weekly exposure
 matrices real friends should colocate!


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## Connecting Similar People



## Network of People



## Network of People $\rightarrow$ Tribes



## Network of People Predictions <br> Percent Improvement



## The Next Net

## Senses MetWOrkS Macrosense Citysense Technology Principles Media Center About Us



