



Digital Epidemiology: Understanding Epidemic Spread using Human Contact Networks

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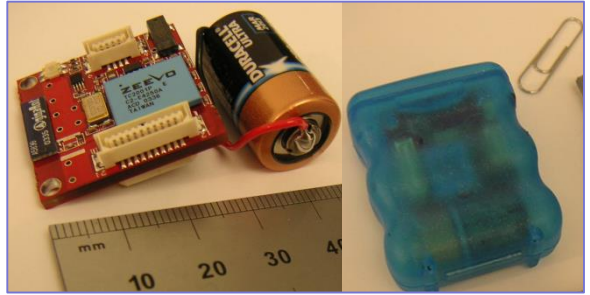
Opportunistic Networks

- **Pocket Switched Networks: Devices carried by people, thus 'do what users do'**
- Pocket switched networks involve consumer devices and incorporate social aspects and opportunistic communication




Measure Human Contact Networks

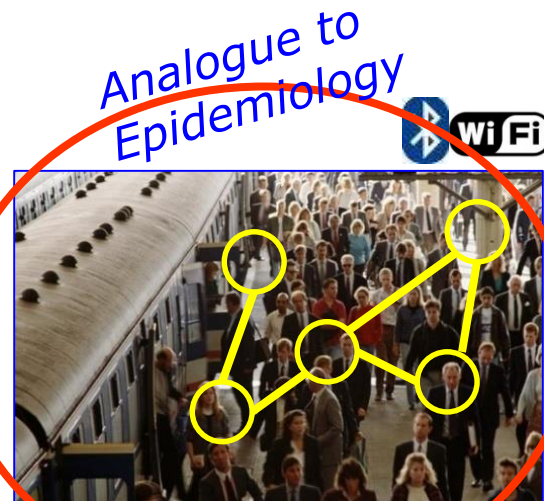
- Sensors
 - Bluetooth Intel iMote
 - 802.15.4 + (magnet, gyroscope)
- RFID Tags
 - UHF Tag Alien ALN-9640 - "Squiggle®" Inlay
 - OpenBeacon active RFID Tag
- Mobile Phones
 - Bluetooth
 - GPS, Google latitude
- GPS Logger
- Online Social Networks
 - Twitter, Facebook, Foursquare...





Spread of Infectious Diseases

- Thread to public health: e.g.,  SARS, Respiratory and other close-contact infectious diseases
- Current understanding of disease spread dynamics
 - Epidemiology: small scale empirical work
- Real-world networks are far more complex
 - Advantage of **real world data**
 - Emergence of wireless technology for proximity data
 - Post-facto analysis and modelling yield insight into human interactions

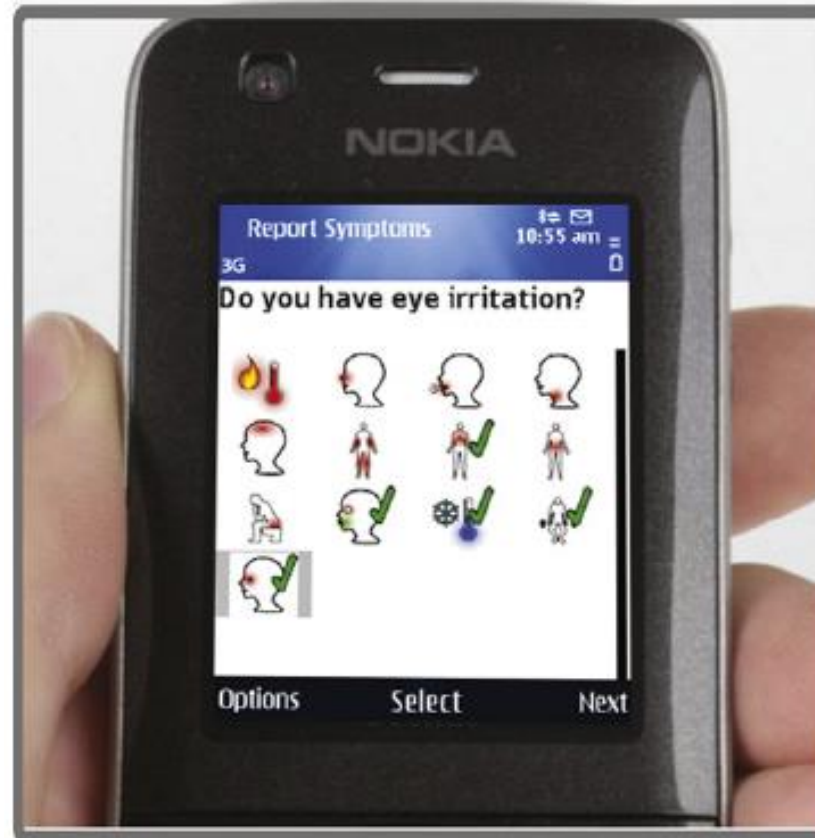


Modelling realistic infectious
disease spread/prediction

FluPhone Project



- Scan Bluetooth devices every 2 minutes
- Symptom Survey





FluPhone Project

- Understanding behavioural risk factors for disease outbreaks
- Proximity data collection using mobile phones in general public in Cambridge

<https://www.fluphone.org>



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FluPhone Study

This is the home page for the FluPhone study. A study to measure social encounters made between people using their mobile phones, to better understand how infectious diseases, like 'flu, can spread between people.

This study will record how often different people (who may not know each other) come close to each other as part of their everyday lives. To do this, we will ask volunteers to install a small piece of software on their mobile phones and to carry their phones with them during their normal day-to-day activities. The software will look for other nearby phones periodically using Bluetooth, record this information and send it to the research team via the cellular phone data service. This information will give us a much better understanding of how often people congregate into small groups or crowds, such as when commuting or through work and leisure activities. Also, by knowing which phones come close to one another, we will be able to work out how people actually are, and how fast diseases could spread within communities. We are also asking volunteers if they experience any influenza-like symptoms during the study period, so that we can compare the spread of 'flu to the underlying social network of encounters made between people.

BBC NEWS CAMBRIDGESHIRE

4 May 2011 Last updated at 17:49

FluPhone app 'helps track spread of infectious diseases'

A mobile phone application could help monitor the way infectious disease such as flu are spread.

The FluPhone app was developed by researchers at the University of Cambridge Computer Laboratory.

Volunteers' phones fitted with the app "talk" to each other, recording how many people each infected subject meets during an imaginary epidemic.

The FluPhone app tracks volunteer infected subjects' using Bluetooth technology

The university is one of seven institutions working on the study to reduce the impact of epidemics.

The FluPhone app uses Bluetooth technology to anonymously record interaction between volunteers involved in the study.

When mobile phones come into close proximity, that fact is recorded and data is sent automatically to the research team.

'Valuable insight'

Professor Jon Crowcroft and Dr Elio Yoneki, co-principal investigators of the study, said they believed the collected data could be used to simulate social interaction during a real epidemic or pandemic.

A three-month FluPhone pilot study, using a basic version of the app, was conducted in Cambridge in 2010.

Dr Yoneki said: "The data was a valuable insight into how human communities are formed, how much time people spend together, and how frequently they meet.

"Such data show complex network-like structures, which is very useful for understanding the spread of disease."

Prof Crowcroft explained epidemiologists traditionally monitor how a disease spreads by asking patients to keep diaries of their movements and social contacts.

"That's very time-consuming and people often forget to do it, or forget who they've met," he said.

The FluPhone app was, he explained, a more reliable way to record contact between "infected subjects".

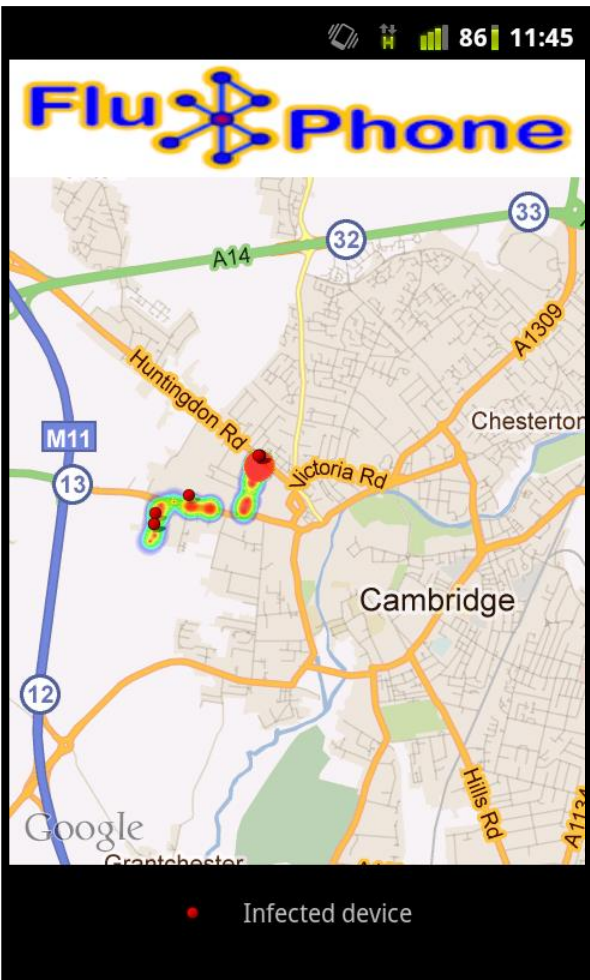
"Provided we have people's permission, we can upload the data, and medical researchers can see who met whom within the set of volunteers, without there being any missing encounters."

Related Stories

Web surveillance map of global disease trends

Trajectory of Encountering Sick People

- Integration with GPS equipped Smartphones

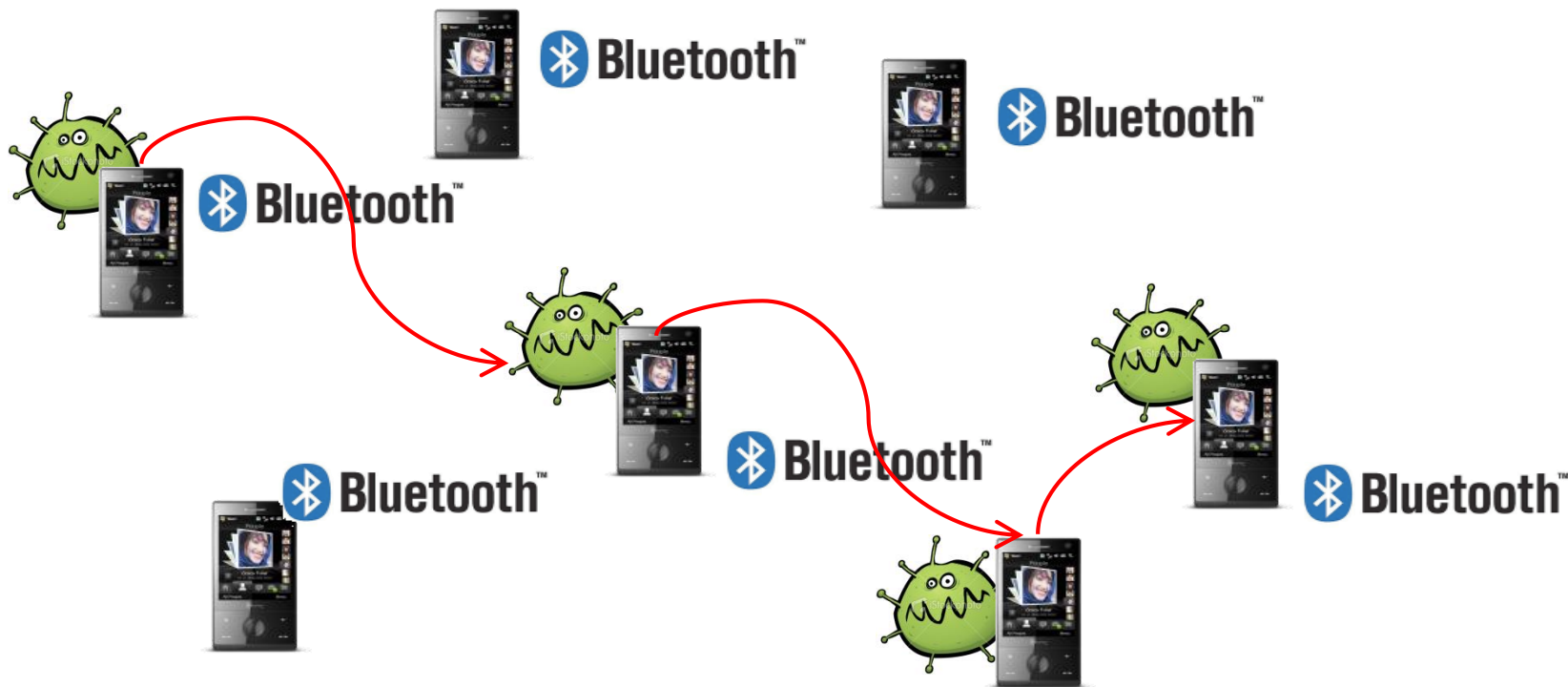




Virtual Disease

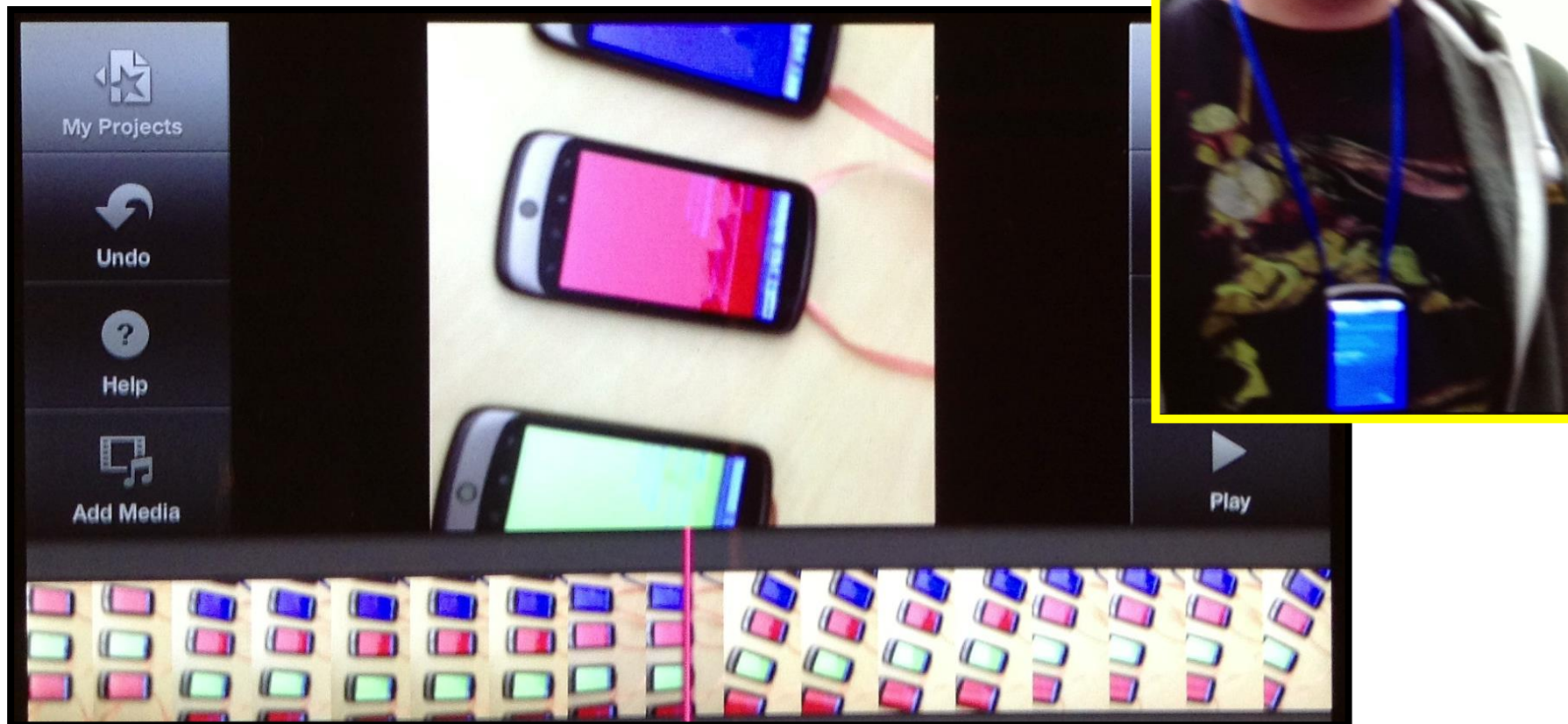
- Spread virtual disease via Bluetooth communication

Disease Name	Exposed Duration	Infectious Duration	Infection Probability
Base line	0	31536000000	1.0
SARS	86400000 <i>.5H</i>	108000000 <i>1H</i>	0.8
Flu	172800000 <i>1H</i>	216000000 <i>2H</i>	0.4
Cold	259200000 <i>2H</i>	432000000 <i>3H</i>	0.2



Virtual Disease: Infection State Coloring

- Spread of Disease (Green: Susceptible, Yellow: Exposed, Red: Infected, and Blue: Recovered)



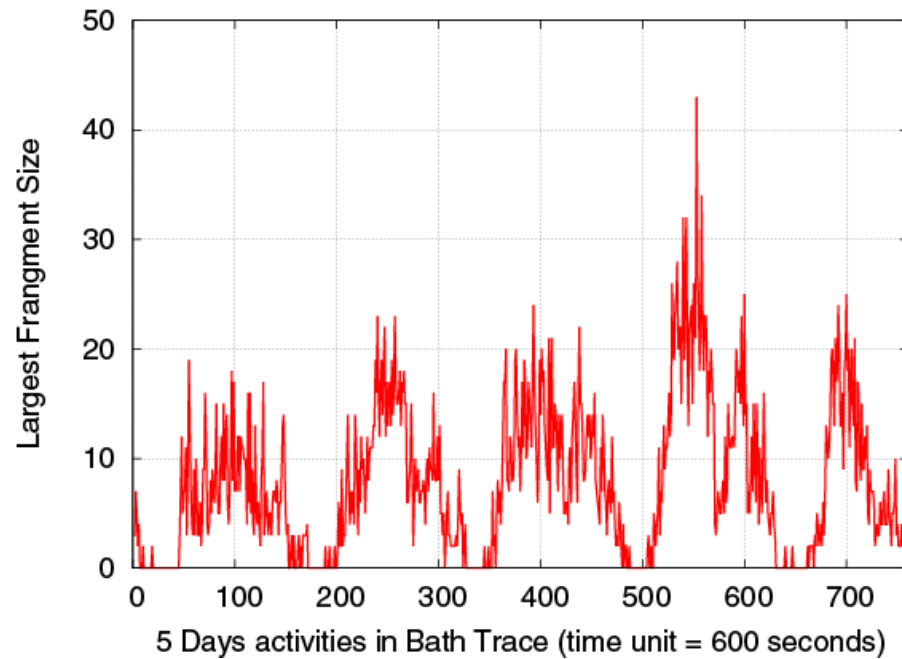
Trace Data of Contact Networks

- Analysis of **dynamic** network structure
 - How does community structure affect epidemic spread?
 - How do hubs influence temporal or spatial effects, and how does this affect the transmission characteristics of disease?

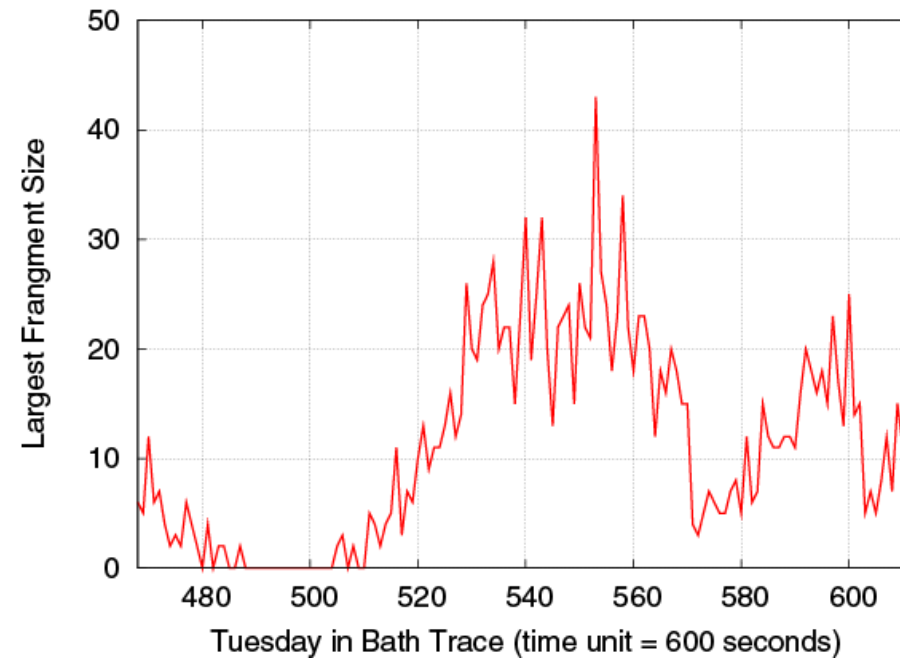
Experimental data set	Cambridge	Infocom06	MIT
Device	iMote	iMote	Phone
Network type	Bluetooth	Bluetooth	Bluetooth
Duration (days)	11	3	246
Granularity (seconds)	600	120	300
Number of Devices	36	78	97
Number of contacts	10,873	191,336	54,667
Average # Contacts/pair/day	0.345	6.7	0.024

Regularity of Network Activity

- Size of largest connected nodes shows network dynamics



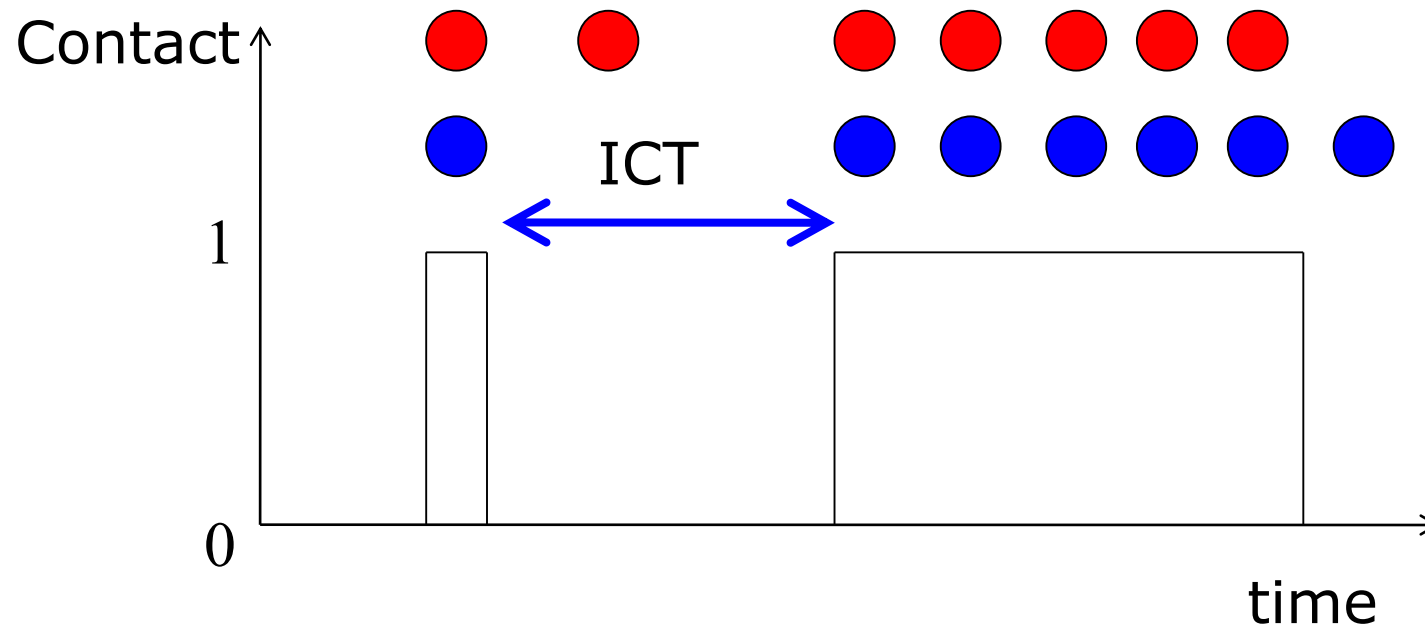
5 Days



Tuesday

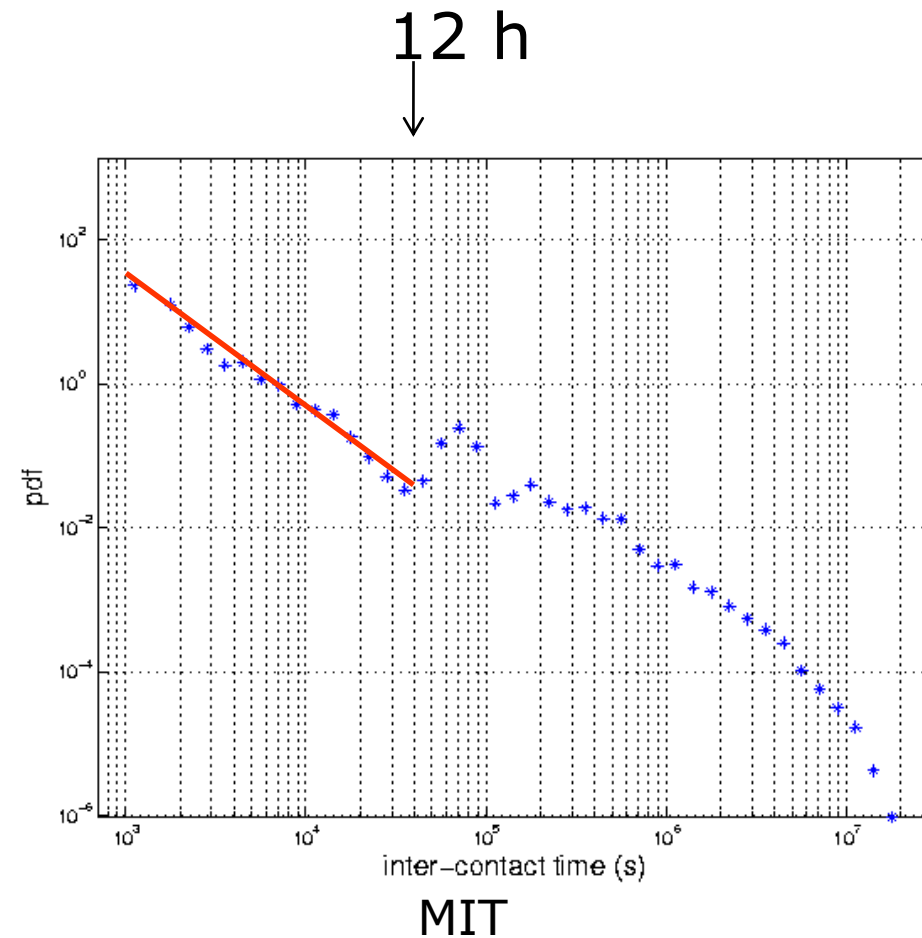
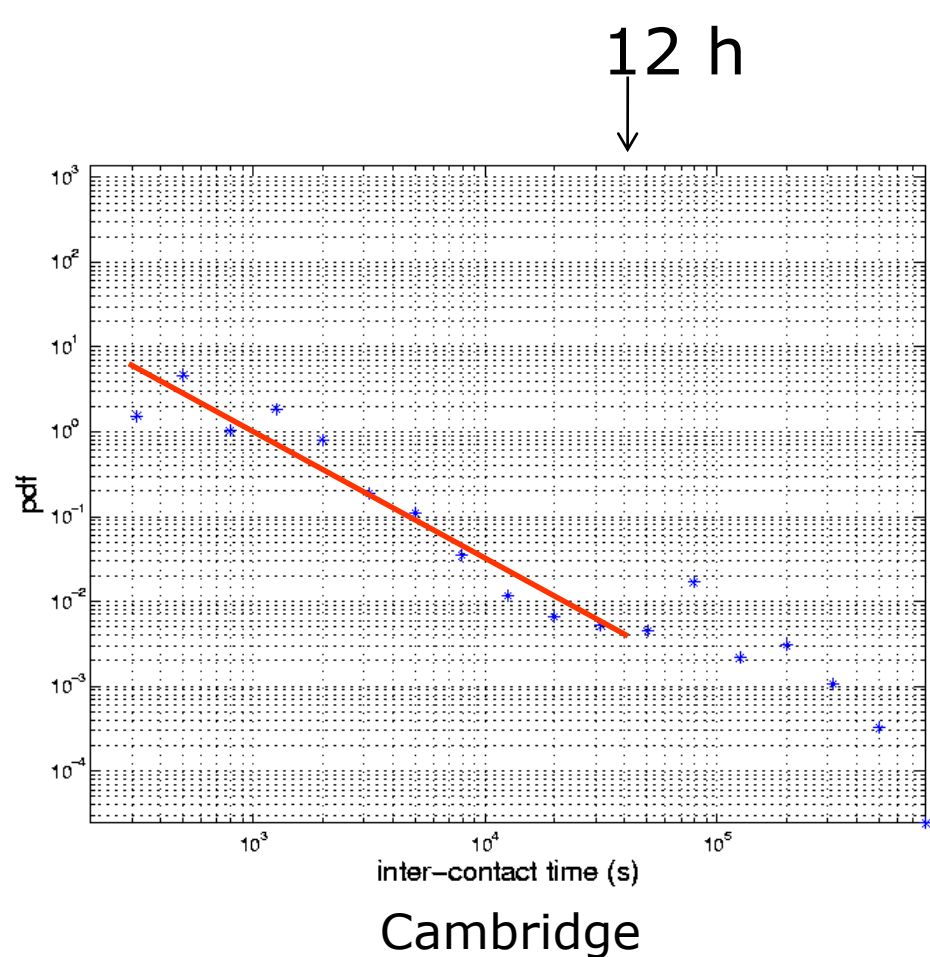
Inter-Contact Time (ICT)

- Calculated all possible inter-contact times between any two nodes, where ICT is defined as time between end of contact among two nodes and start of next contact among same two nodes



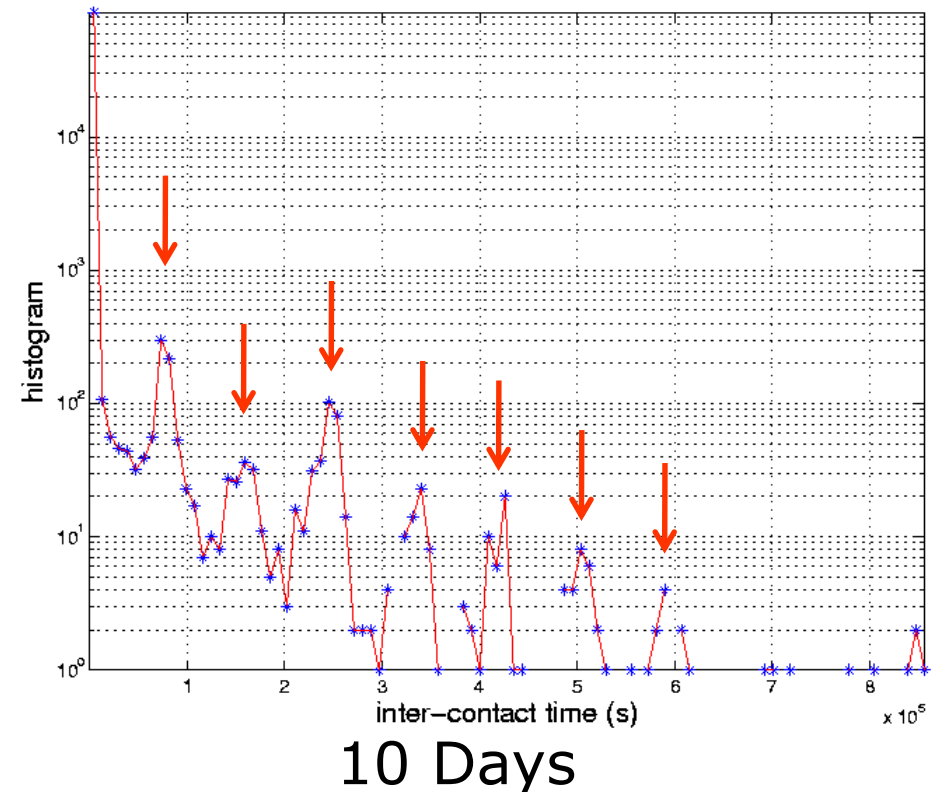
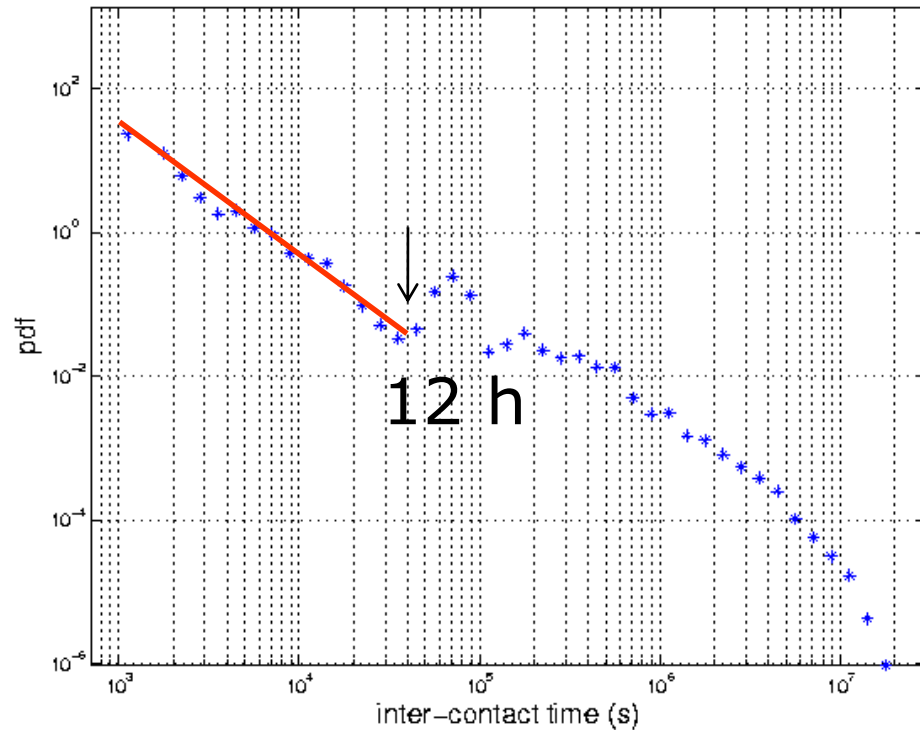
ICT: Random and Scale-free

- Sufficiently short time scales (<12 hours): ICT dist is approximated by power law



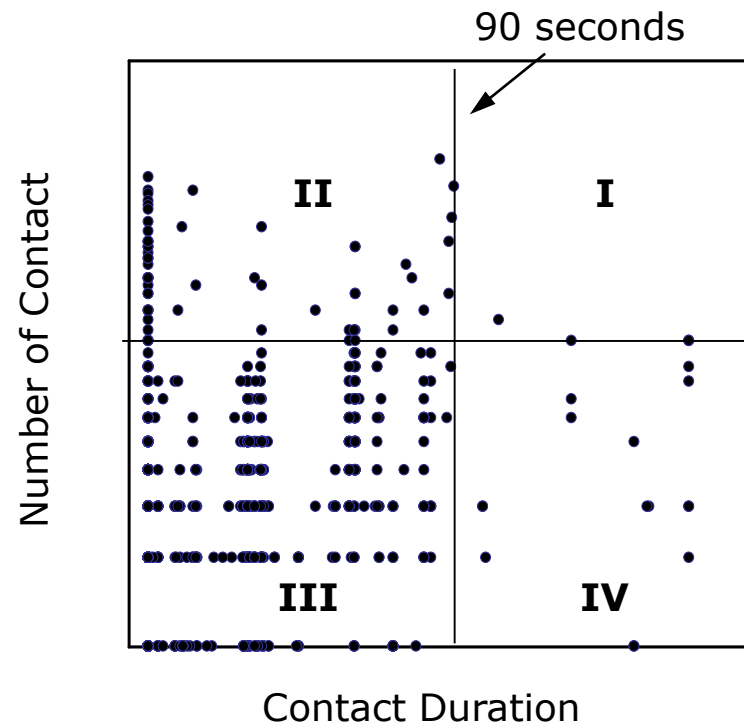
Inter Contact time: Random/Scale-free and Periodic

- Sufficiently short time scales (<12 hours): ICT distribution is approximated by power law
- Environmental, biological and social constraints show rhythms (ICT separated by 24 hours)



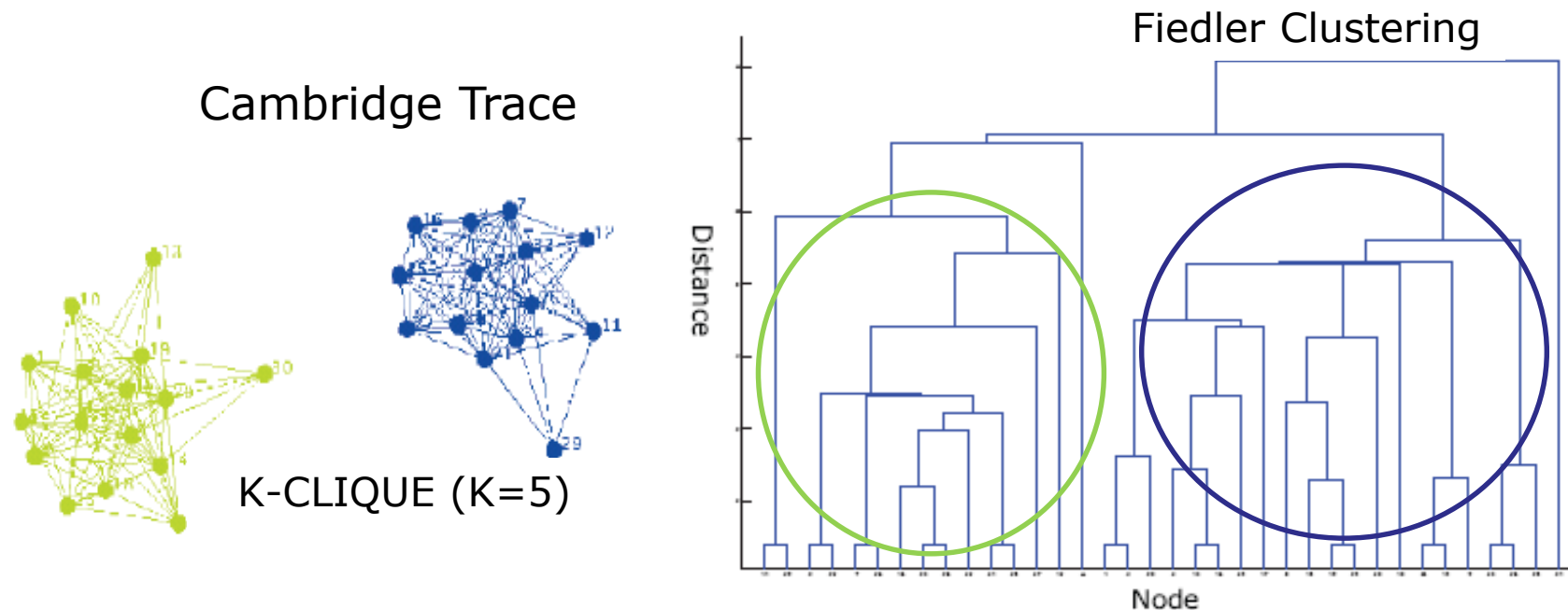
Edge Weight

- I.** High Contact N° - Long Duration: Community
- II.** High Contact N° - Short Duration: Familiar Stranger
- III.** Low Contact N° - Short Duration: Stranger
- IV.** Low Contact N° - Long Duration: Friend



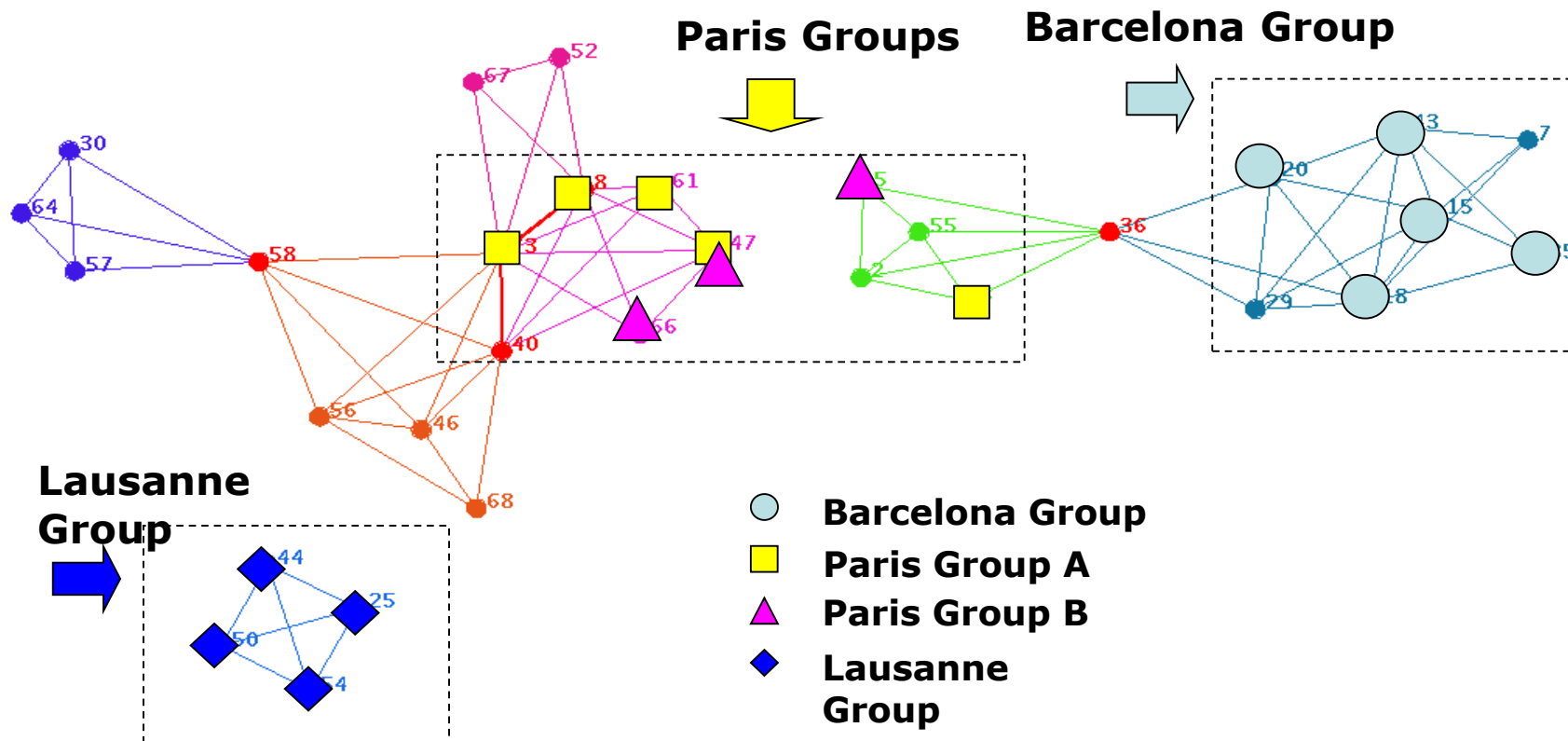
Uncovering Community

- Contact trace in form of weighted (multi) graphs
 - Contact Frequency and Duration
- Use community detection algorithms from complex network studies
 - K-clique [Palla04], Weighted network analysis [Newman05], Betweenness [Newman04], Modularity [Newman06], Fiedler Clustering etc.





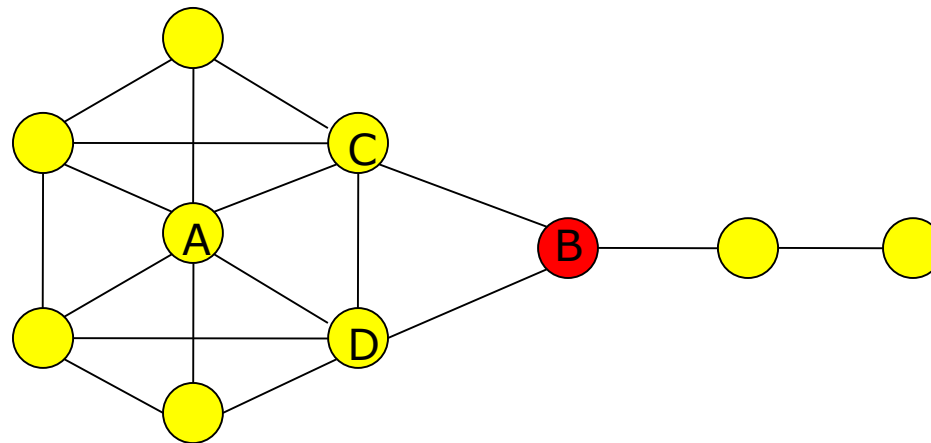
K-CLIQUE Communities with INFOCOM Conference



K=4

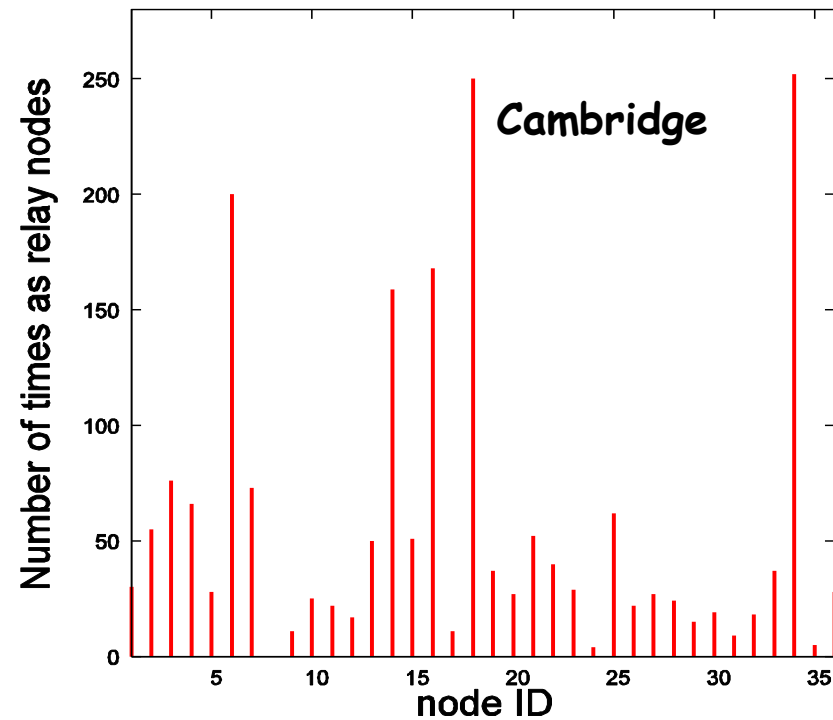
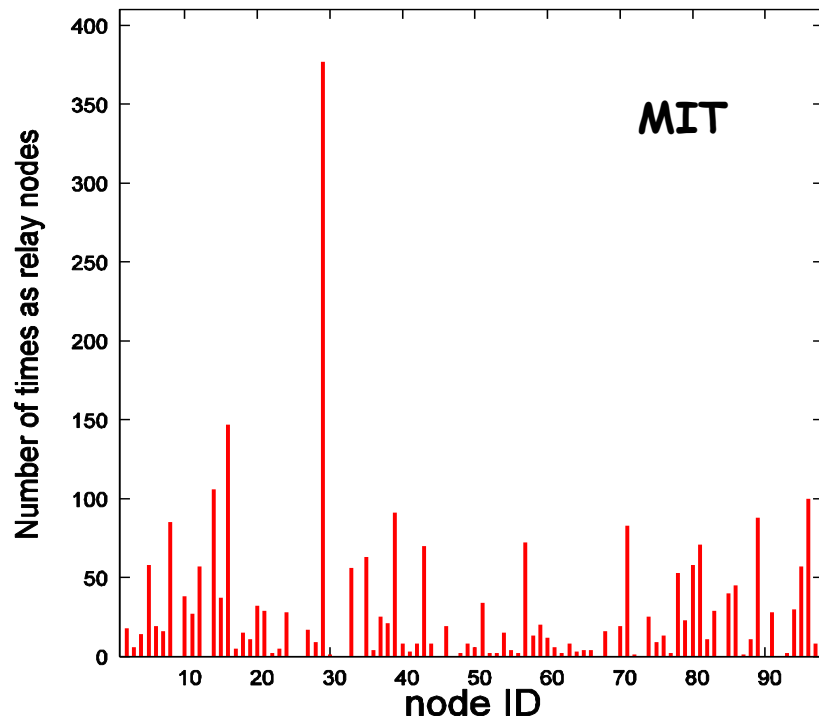
Centrality

- **Betweenness Centrality:** Control over information flowing between others
 - High betweenness node is important as a relay node
 - Large number of unlimited flooding, number of times on shortest delay deliveries → Analogue to Freeman centrality



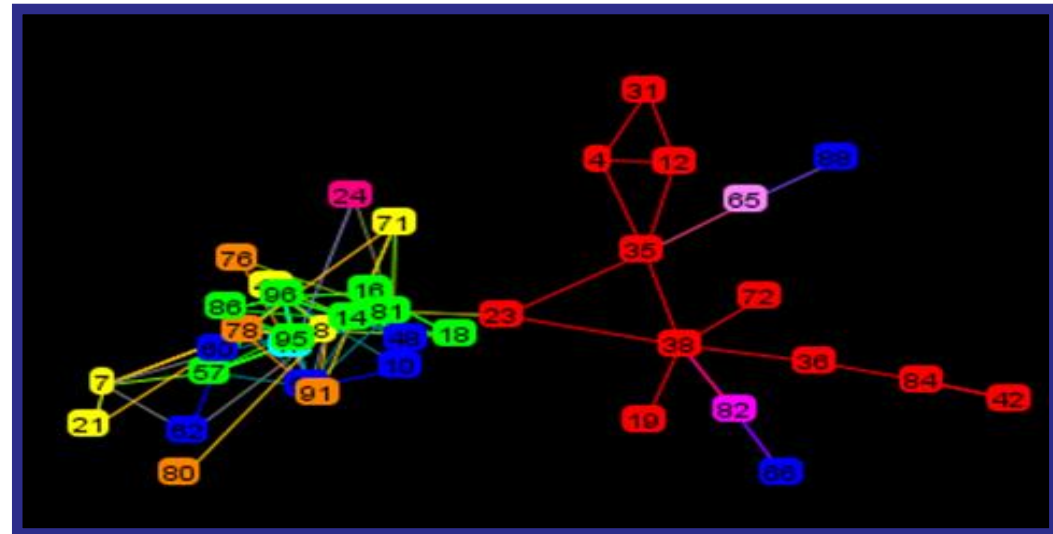
Betweenness Centrality

- Frequency of a node that falls on the shortest path between two other nodes
- High ranking nodes \sim Popular nodes



Contact Networks

- Contact networks: **time dependent** contacts
 - $A \rightarrow B \rightarrow C \rightarrow A$
- Think about spread of messages, infectious disease, and gossip
 - Physical contacts, virtual online contacts
- Understand flow and control: key to uncover centrality nodes



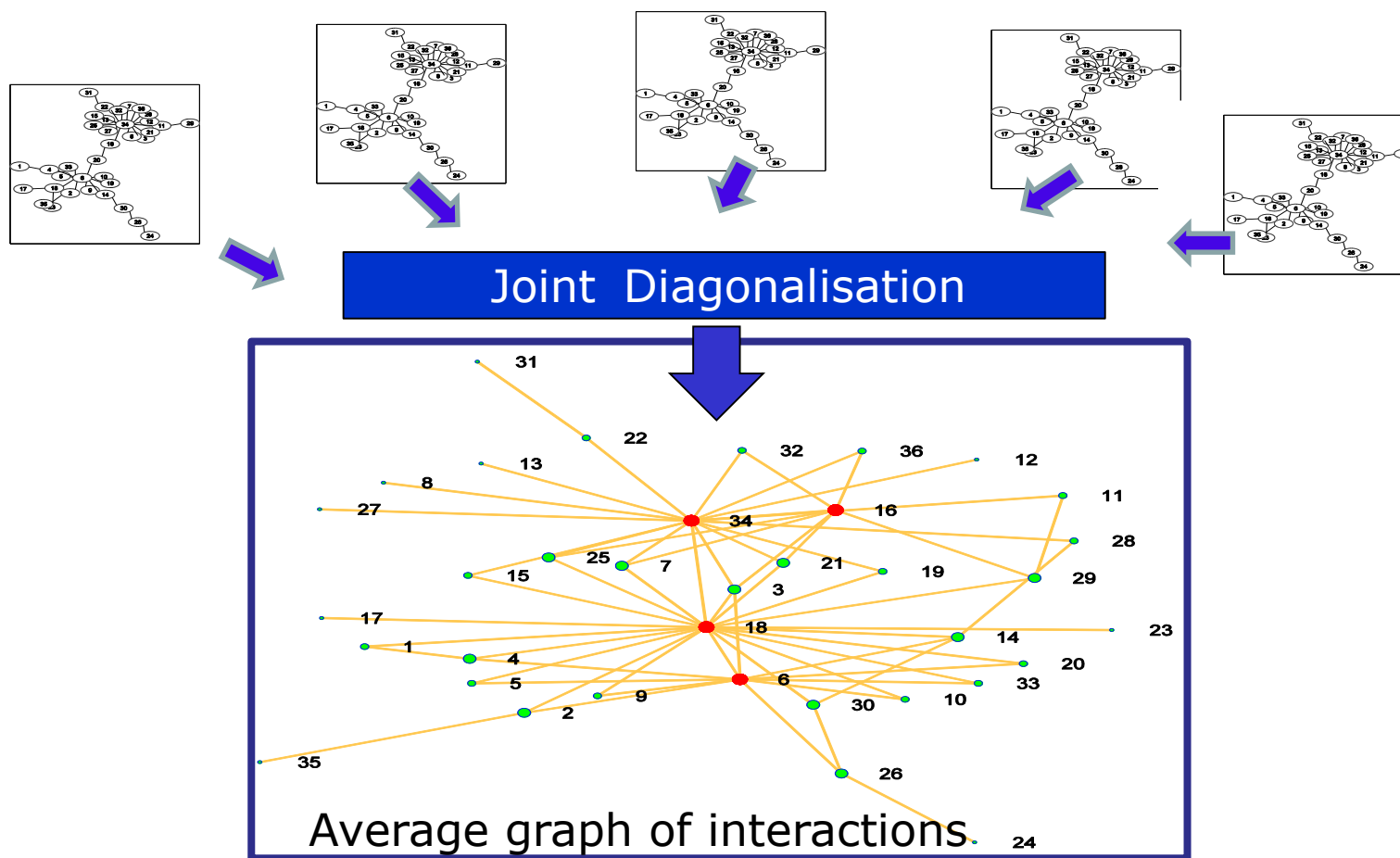


Multiple Spread Modes

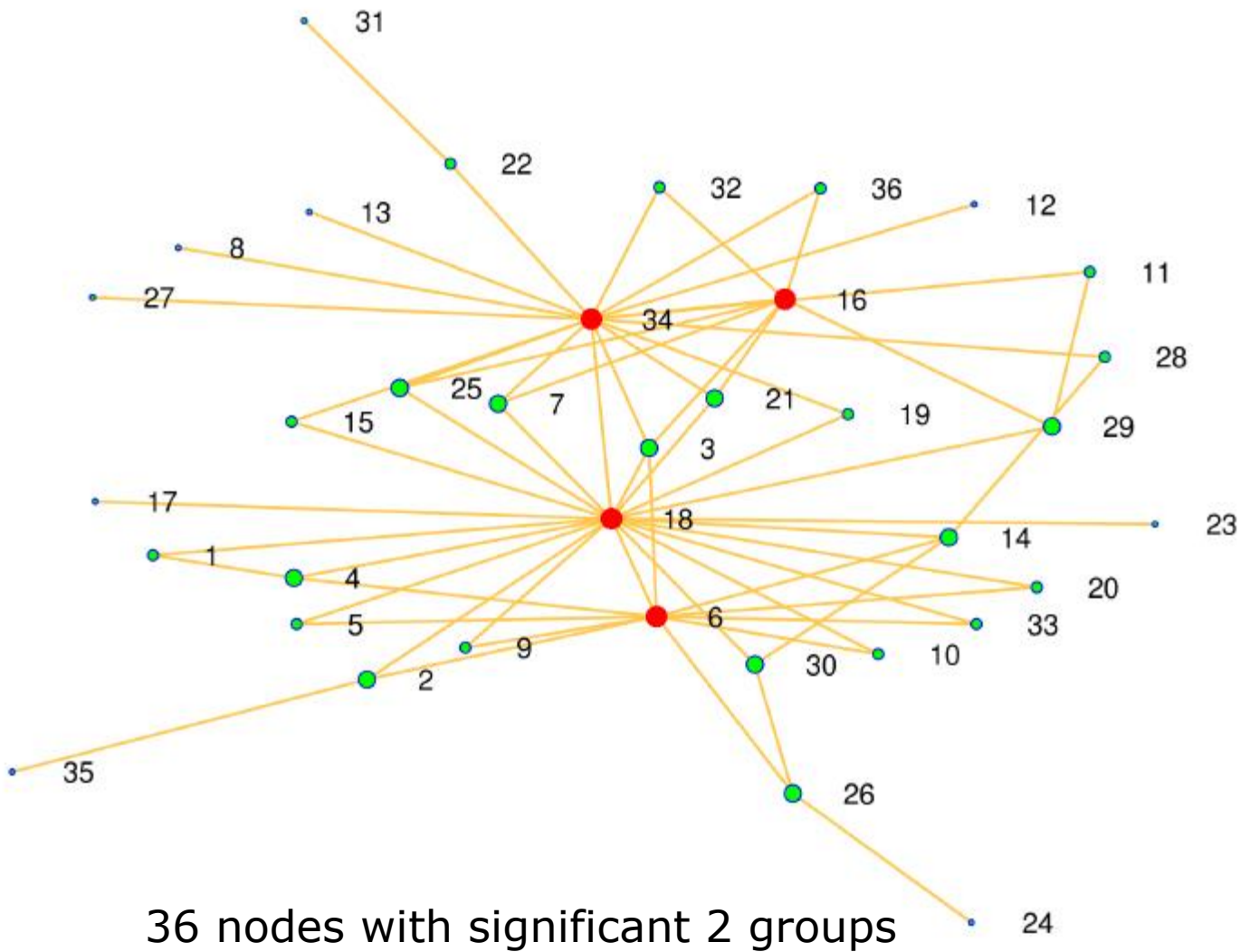
- Typical approach: Cluster nodes to build single network or multiple networks within the sliced time windows
 - Aggregate the number of contacts
 - Ignores time
 - Ignores correlation between links
- Solution: Use spanning tree based samples of a network
 - Akin to spreading a disease in the population and recording the order of infection
 - Define an eigen-space average across these trees
 - Distribution of deviations gives the required groups

Joint Diagonalisation

- Build by combining many of spanning tree based samples of a network using Joint Diagonalisation → Average Interaction Network



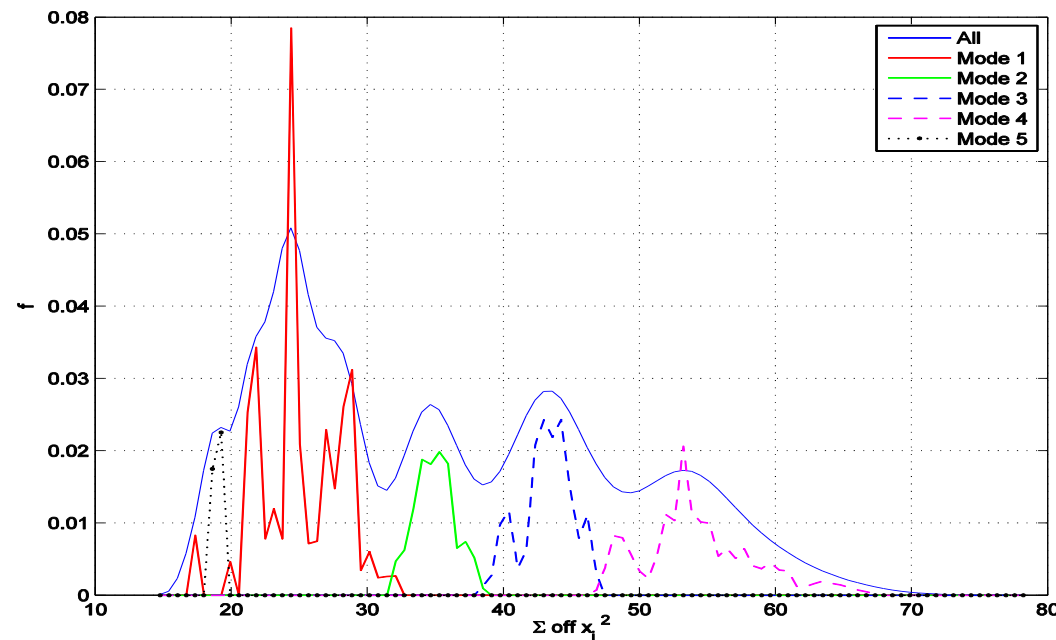
Average Graph of Interactions



36 nodes with significant 2 groups

Multiple Network Modes

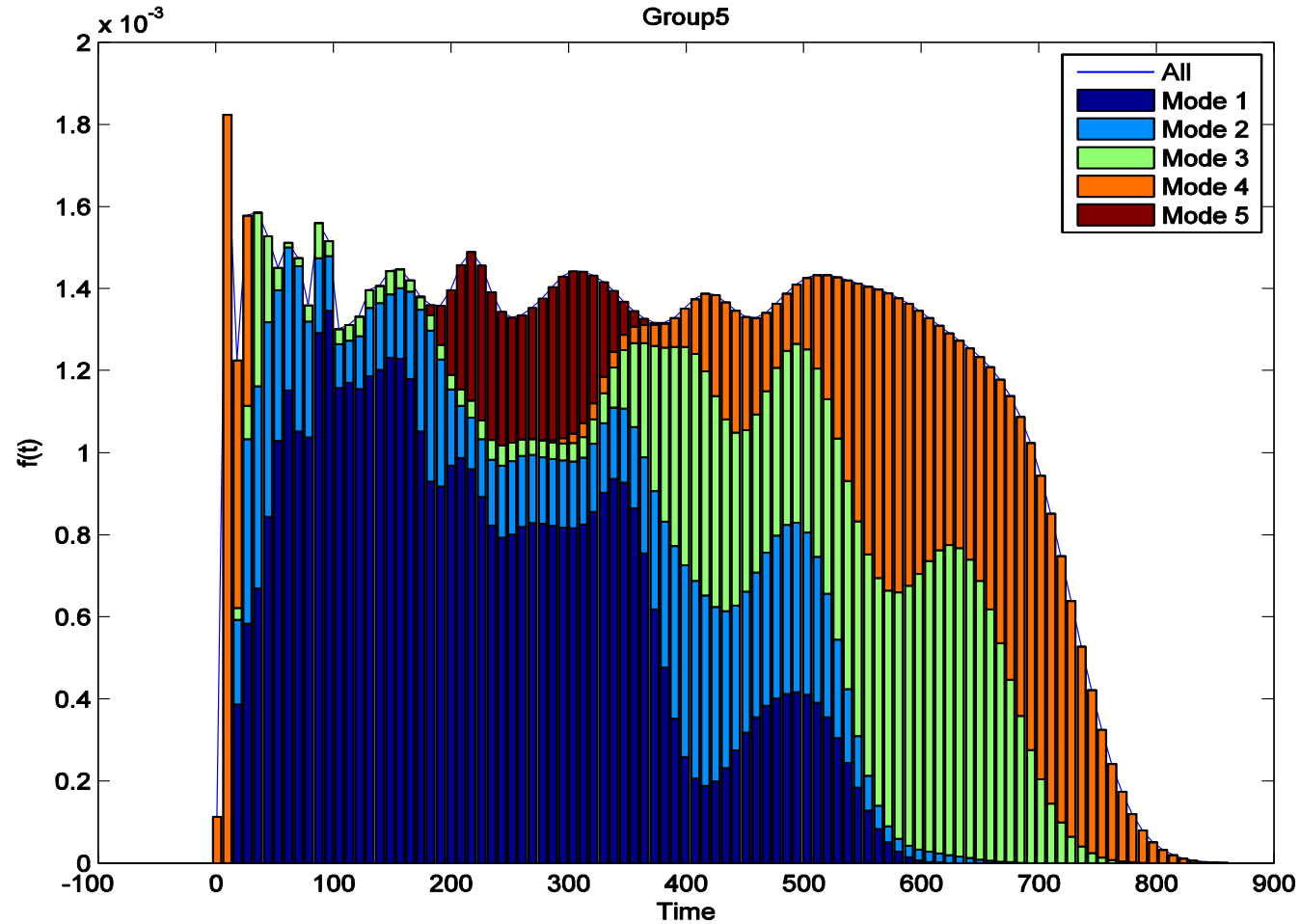
- Build average network with Joint Diagonalisation method
- Define deviation from mean eigen-space as sum of off-diagonal elements
- Use Gaussian mixture model for mode determination
- Distribution of deviation \rightarrow different behaviour of network



Distribution of deviation: random network shows one mode

Extract Spread Modes

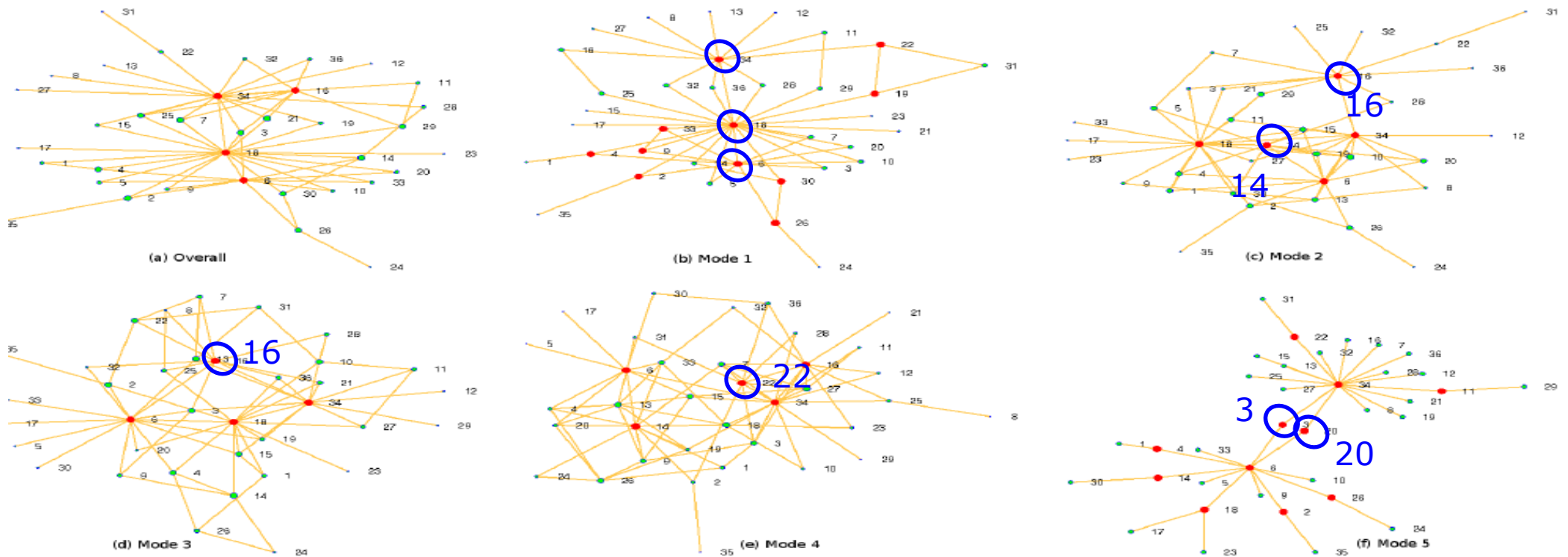
- Change of mode corresponds with state transition



Distribution of times by mode

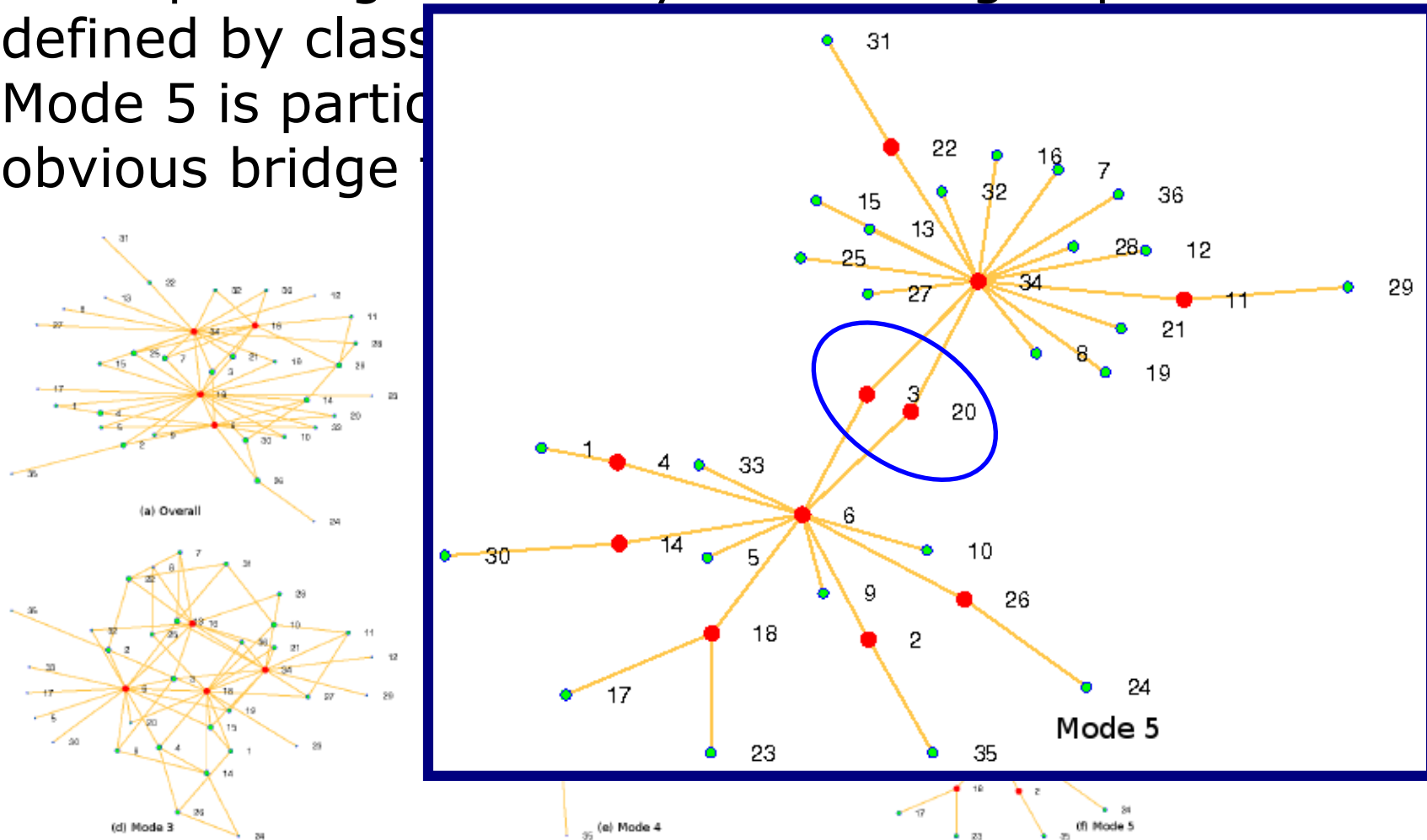
Network Structure of Each Mode

- Mode 1 shows highly structured network corresponding to day when groups are well defined by group dependent activity
- Mode 5 is particularly interesting as there is obvious bridge formed by nodes 3 and 20



Network Structure of Different Modes

- Mode 1 shows a highly structured network corresponding to the day when the groups are well defined by class
- Mode 5 is particularly interesting as it shows a very obvious bridge



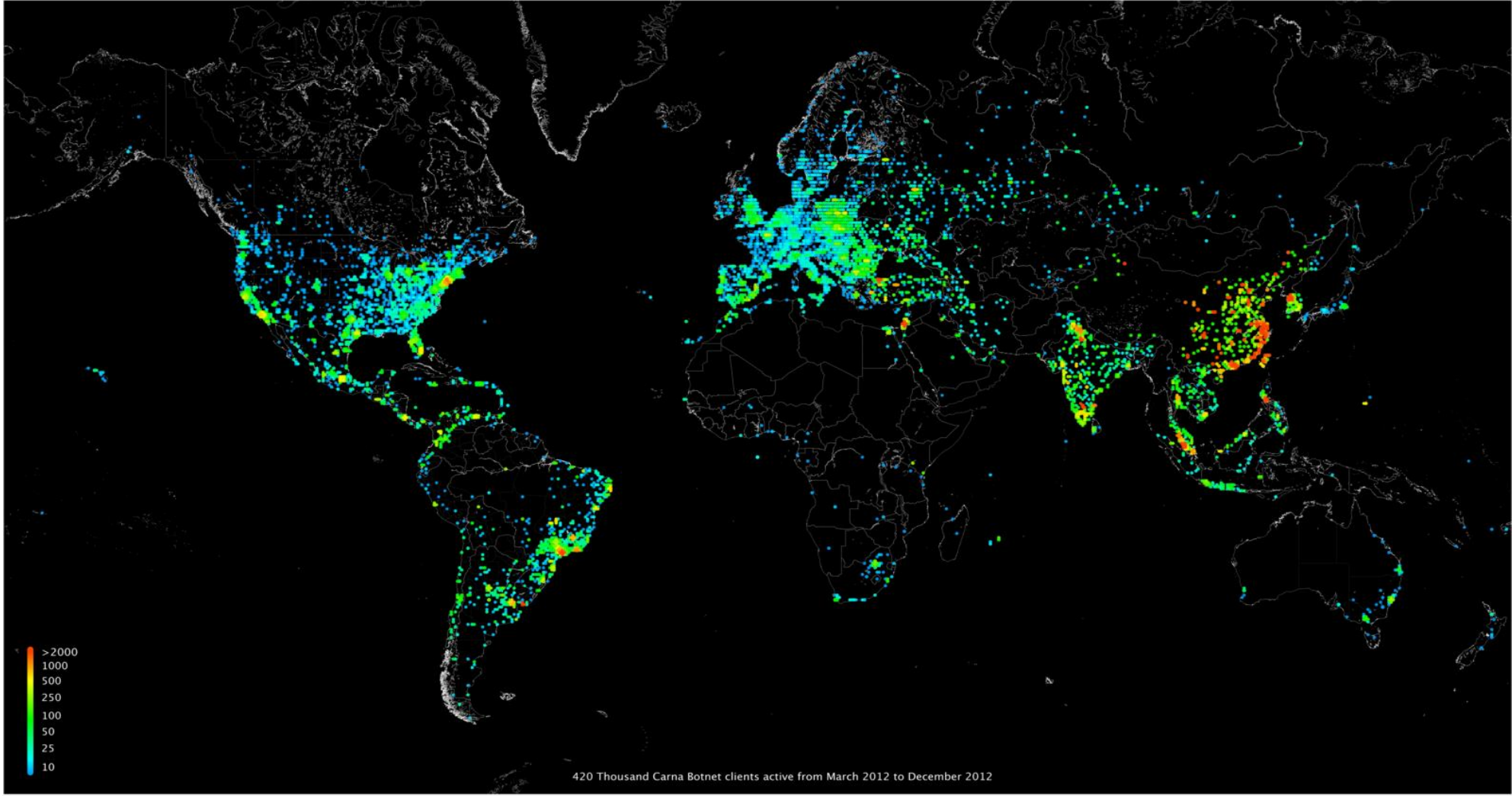
Data Collection in Developing Countries

- Studies in Africa and South America are desired
- Target diseases: Measles, tuberculosis, meningococcal, respiratory syncytial virus and influenza

Sensing Platform in Remote Region

- Build a platform for sensing and collecting data in developing countries
 - Build a standalone network for data collection and communication using Raspberry Pi → **RasPiNET**
 - Inexpensive network setting
 - Support streaming model
 - In-network partial data processing

Internet Coverage



OpenBeacon RFID Tags

- OpenBeacon Active RFID Tags
- Bluetooth has an omnidirectional range of $\sim 10\text{m}$
- OpenBeacon active RFID tags: Range $\sim 1.5\text{m}$ and only detect other tags are in front of them
- Low Cost $\sim = 10\text{GBP}$
- Face-to-Face detection
- Temporal resolution 5-20 seconds
- On-board storage (up to ~ 4 logs)
- Battery life $\sim 2\text{-}3$ weeks

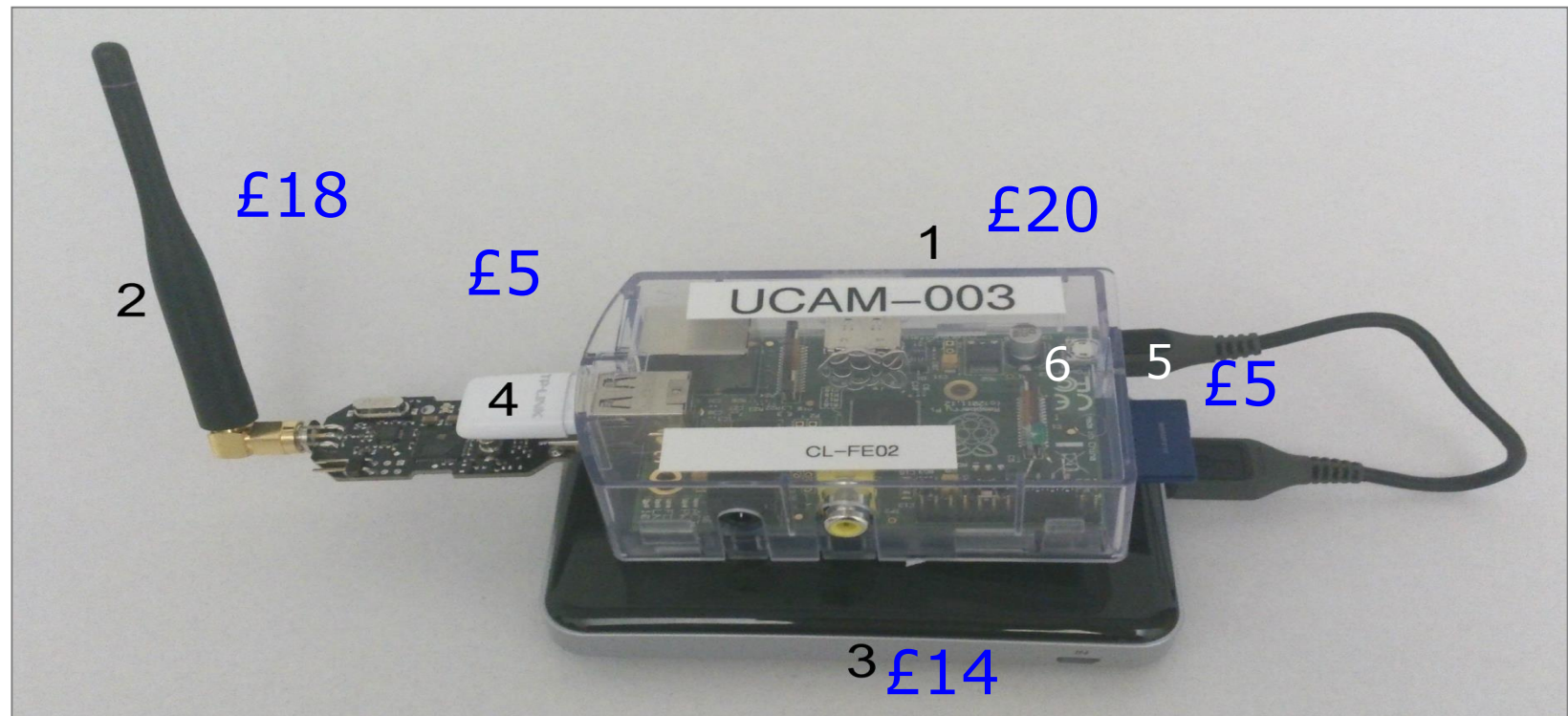


An OpenBeacon
RFID tag



OpenBeacon
Ethernet EasyReader

Raspberry Pi Open Beacon RFID Tag/Reader

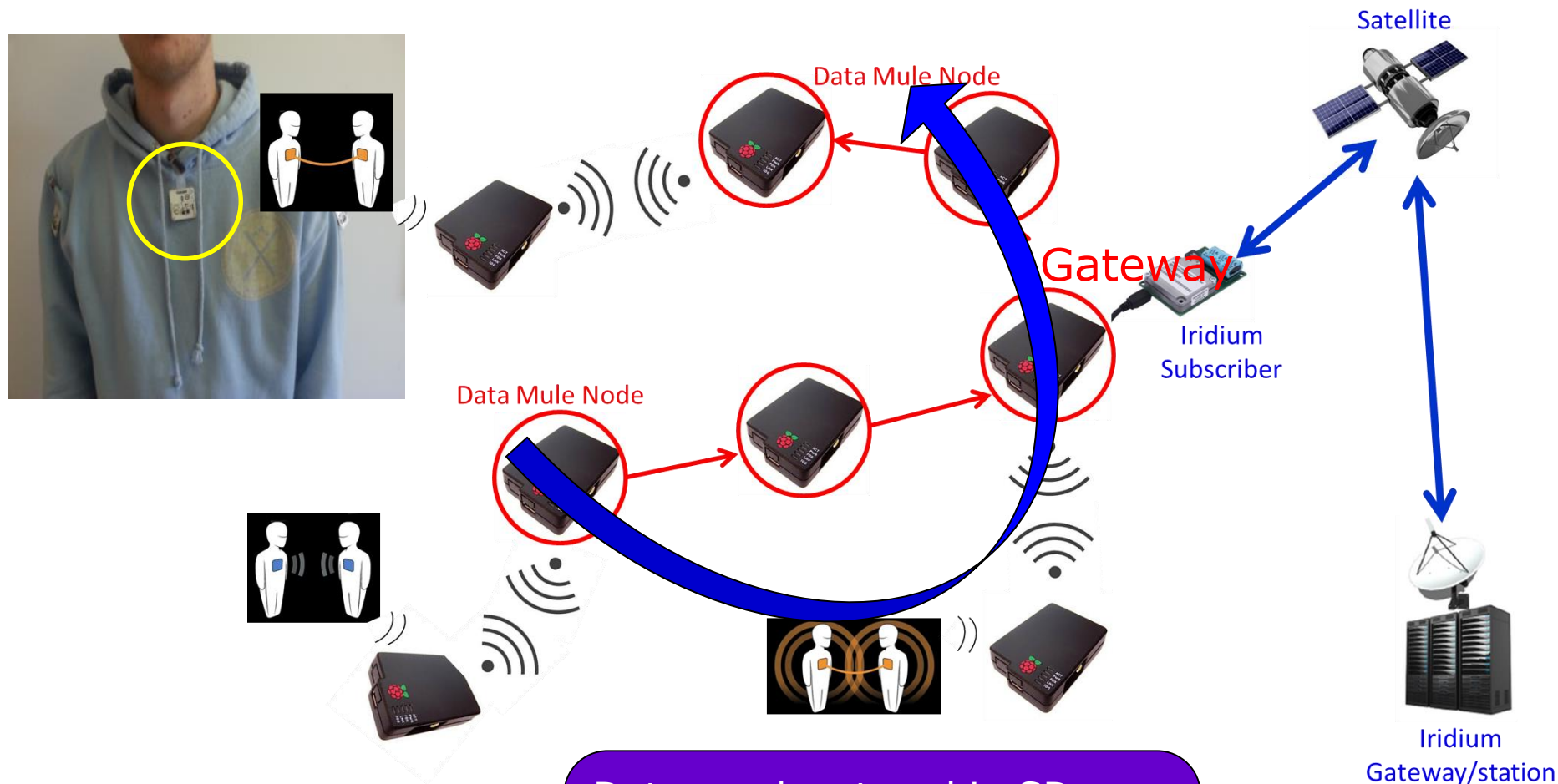


Open Beacon RFID Tag
Range ~1.5m
Low cost 10GBP

- 1. Raspberry Pi
- 2. OpenBeacon USB reader
- 3. Battery Pack (7000mAh)
- 4. WiFi dongle
- 5. SD Card
- 6. LED



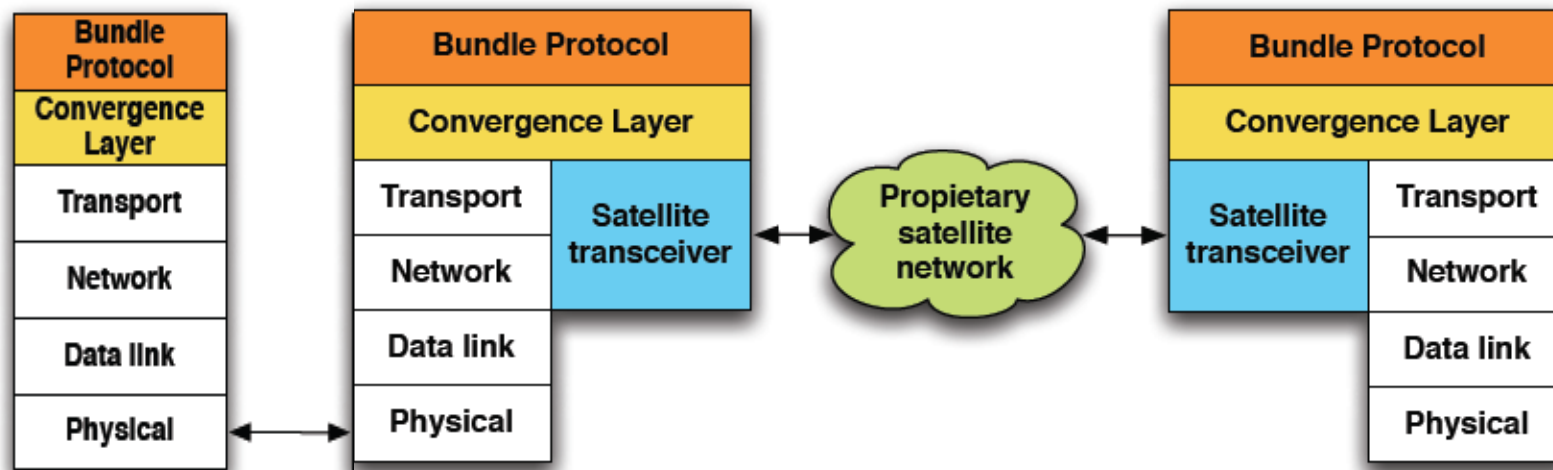
Raspberry Pi based Sensing Platform



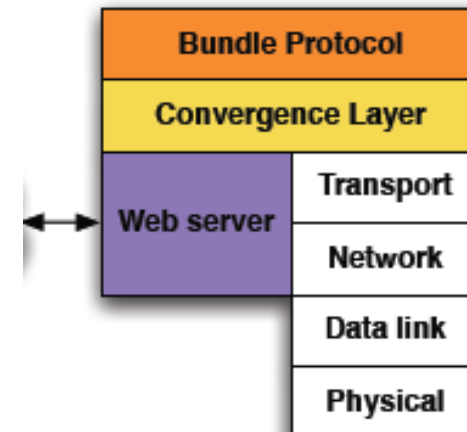
Data can be stored in SD card, transmitted to Data Mule node, or use of WiFi AdHoc mode transmission to Gateway

Communication Protocol

- Protocol for communication between devices with satellite transceiver



- Rockblock provides Web Service Interface
- also Email Interface





Pilot Study in Computer Laboratory

- 15 RasPi OpenBeacon Readers around Computer Laboratory
- 30 participants (4 groups)
- 3 days of data collection

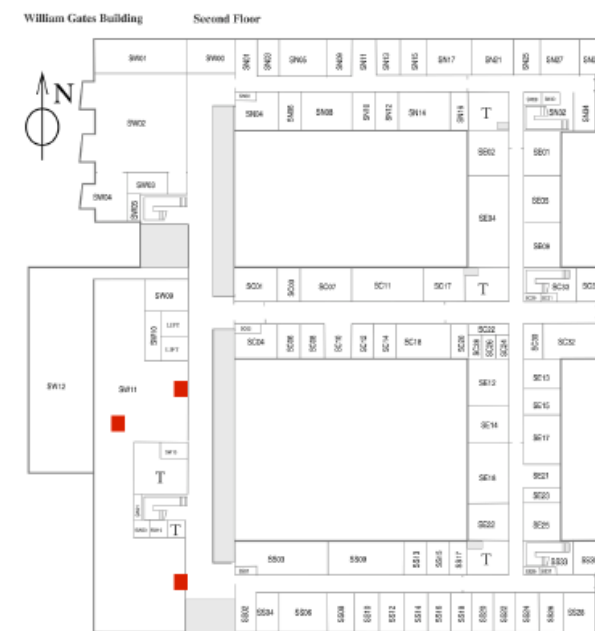
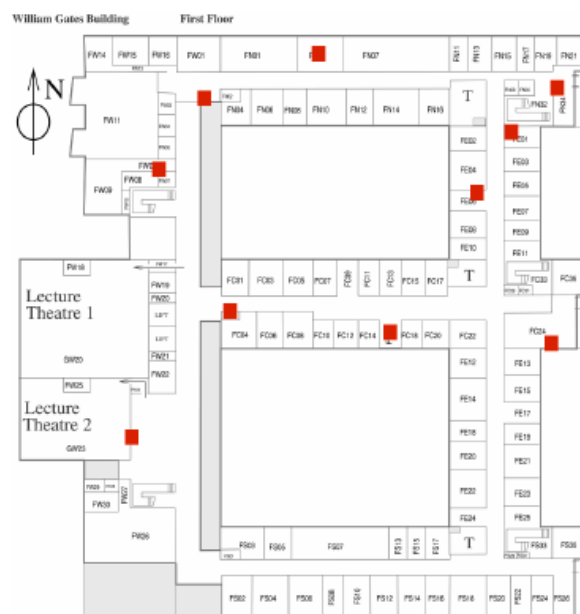
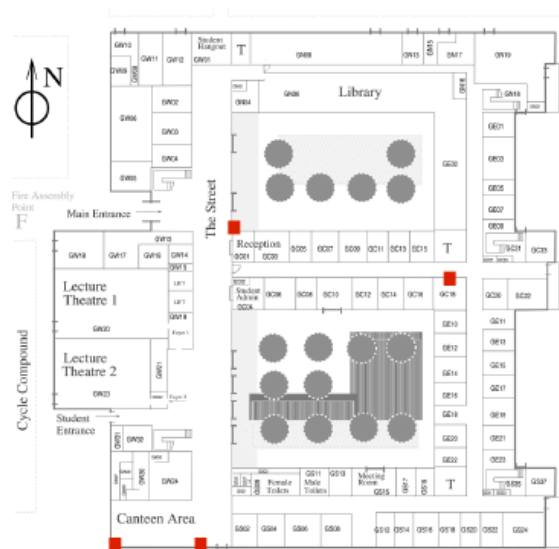


A participant wearing three RFID tags



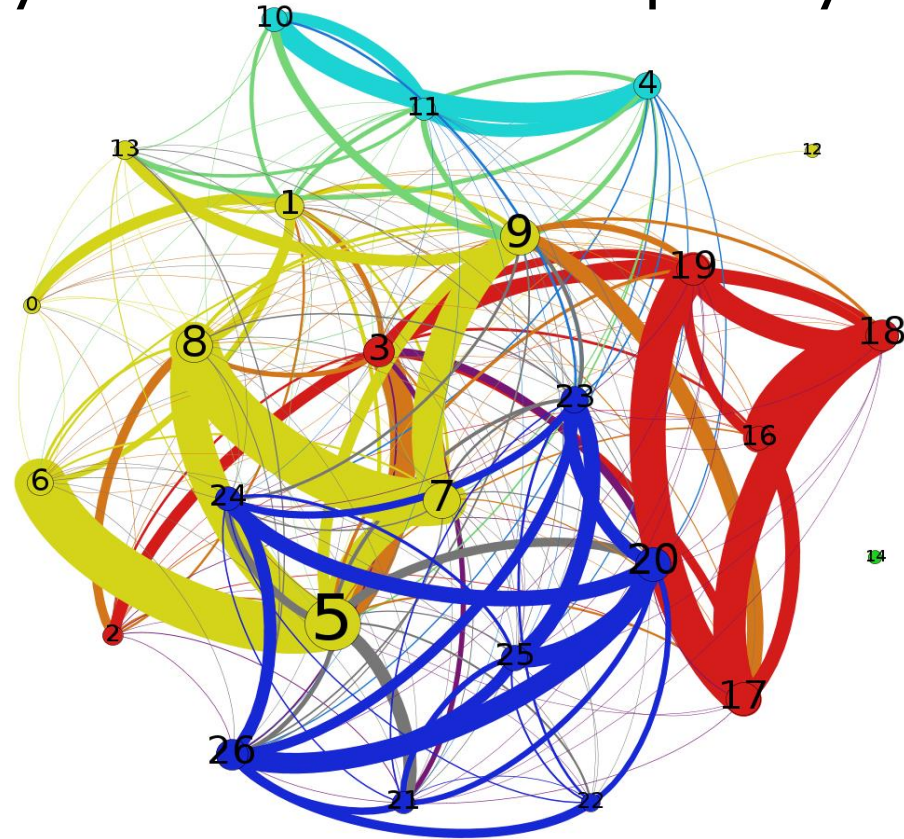
Setting RasPiNET on 3 Floors

- Use of Data Mule approach for Data Collection
- Satellite Communication for sending statistics and changing sensing rate



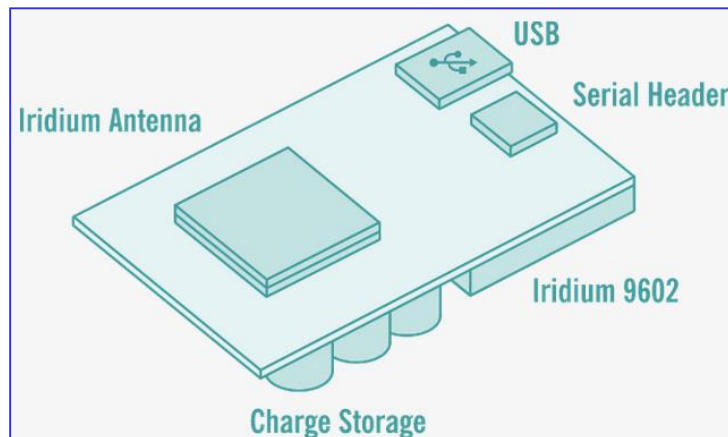
Post Data Analysis on Pilot Study

- Community Detection (4 groups and bridging nodes can be identified)
- No in-depth traffic analysis or network capacity evaluation yet
- One simulator based Simulator (w and w/o satellite connectivity)



Satellite Communication

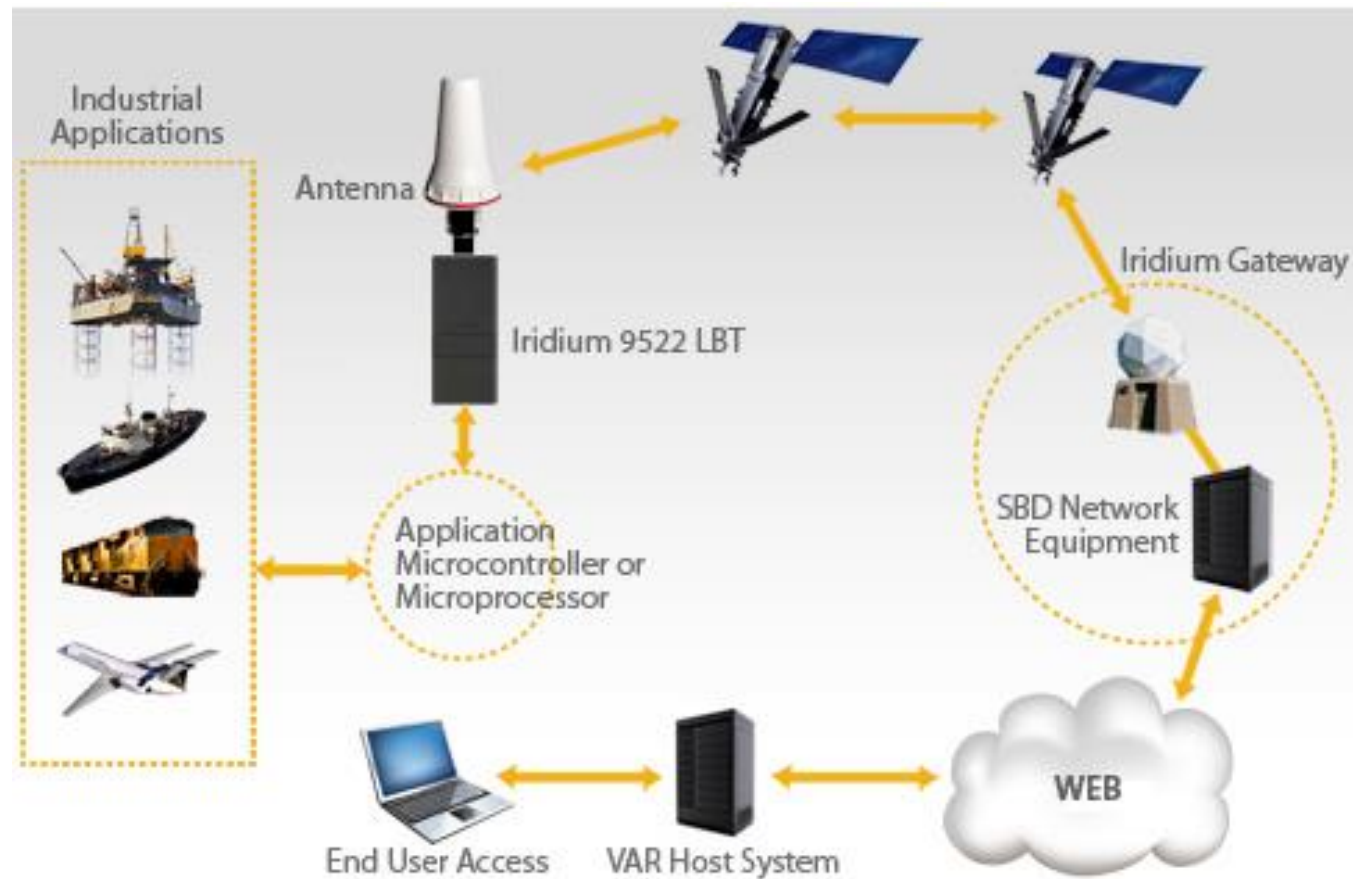
- Satellite module integration in Raspberry Pi
 - RockBLOCK Satellite Module (~=£120)
 - Uses Iridium Satellite Network: Short Burst Data(SBD)
 - Iridium SBD session roughly every 10 seconds
 - To email address, or own web service (i.e. HTTP POST)
 - pay-as-you-go – 34 bytes per message (Hex encoded)
 - 50 credits - 12p/message
 - 20000 Credits – 4p/message



RockBLOCK satellite module

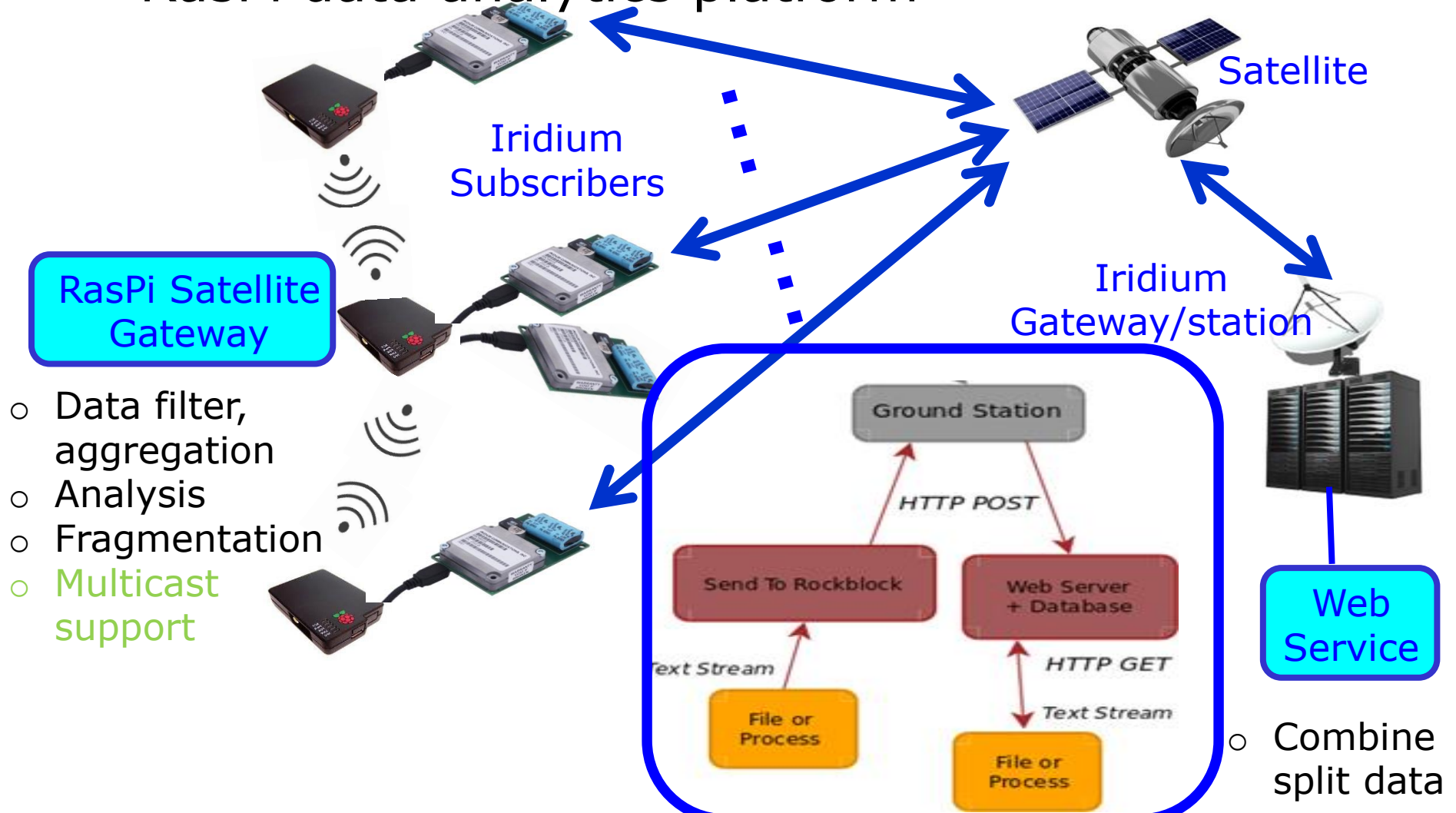
RasPiNET with Satellite Communication

- Satellite module is integrated
- Useful in developing country



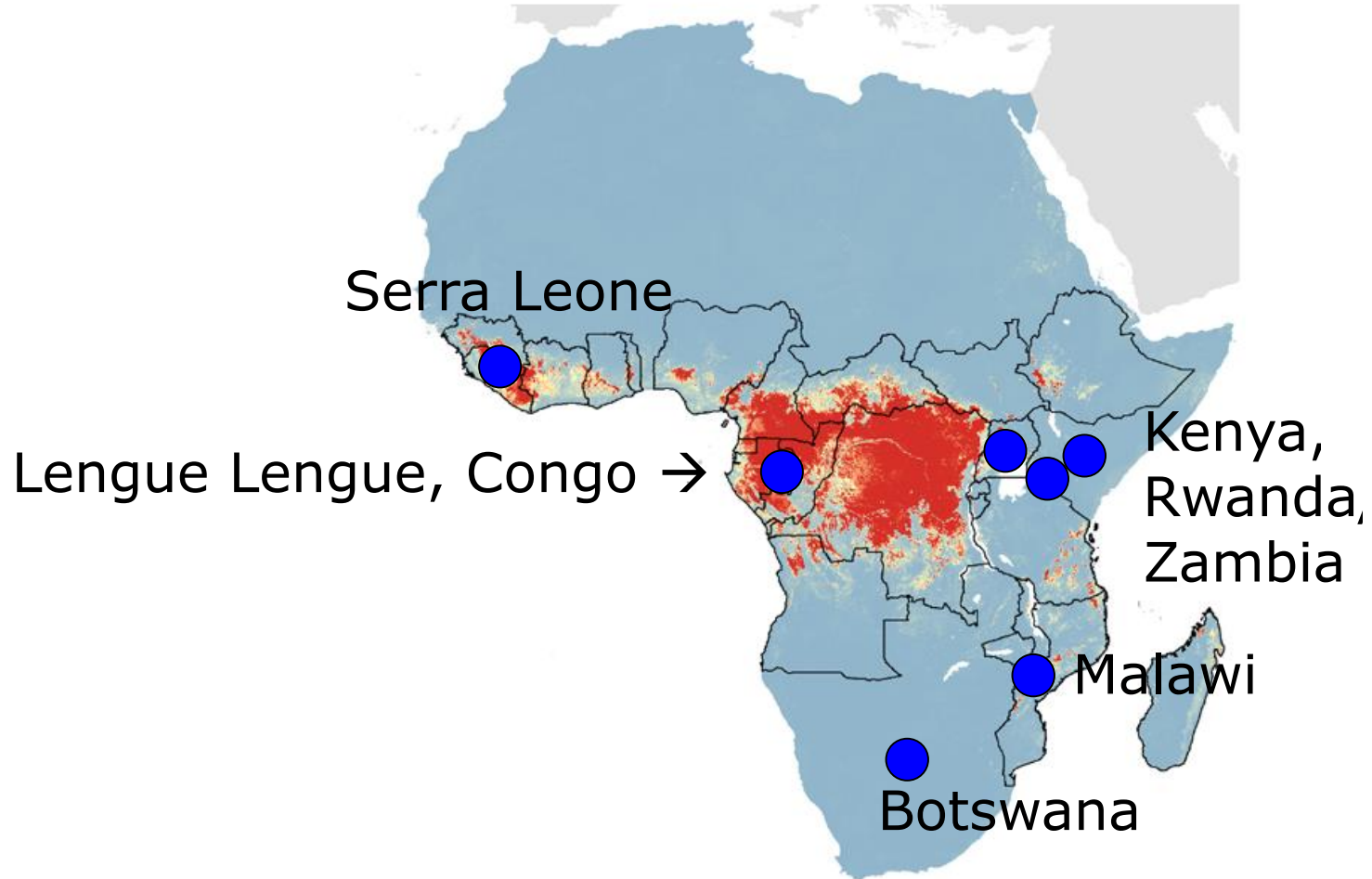
RasPi Satellite Gateway

- Build stream processing paradigm
- RasPi data analytics platform



In Africa, Several Projects Planned...

- Complicated **ethical approval** → killed the projects

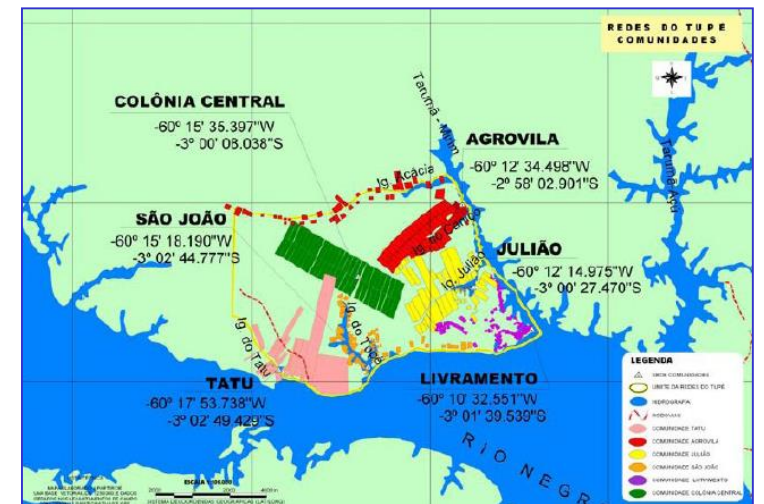


2015 Brazilian Amazon

- Local help by Federal University of Amazonas
- Small scale data collection from few fragmented rural communities along Rio Negro (~90 People, ~70 Raspberry Pis)

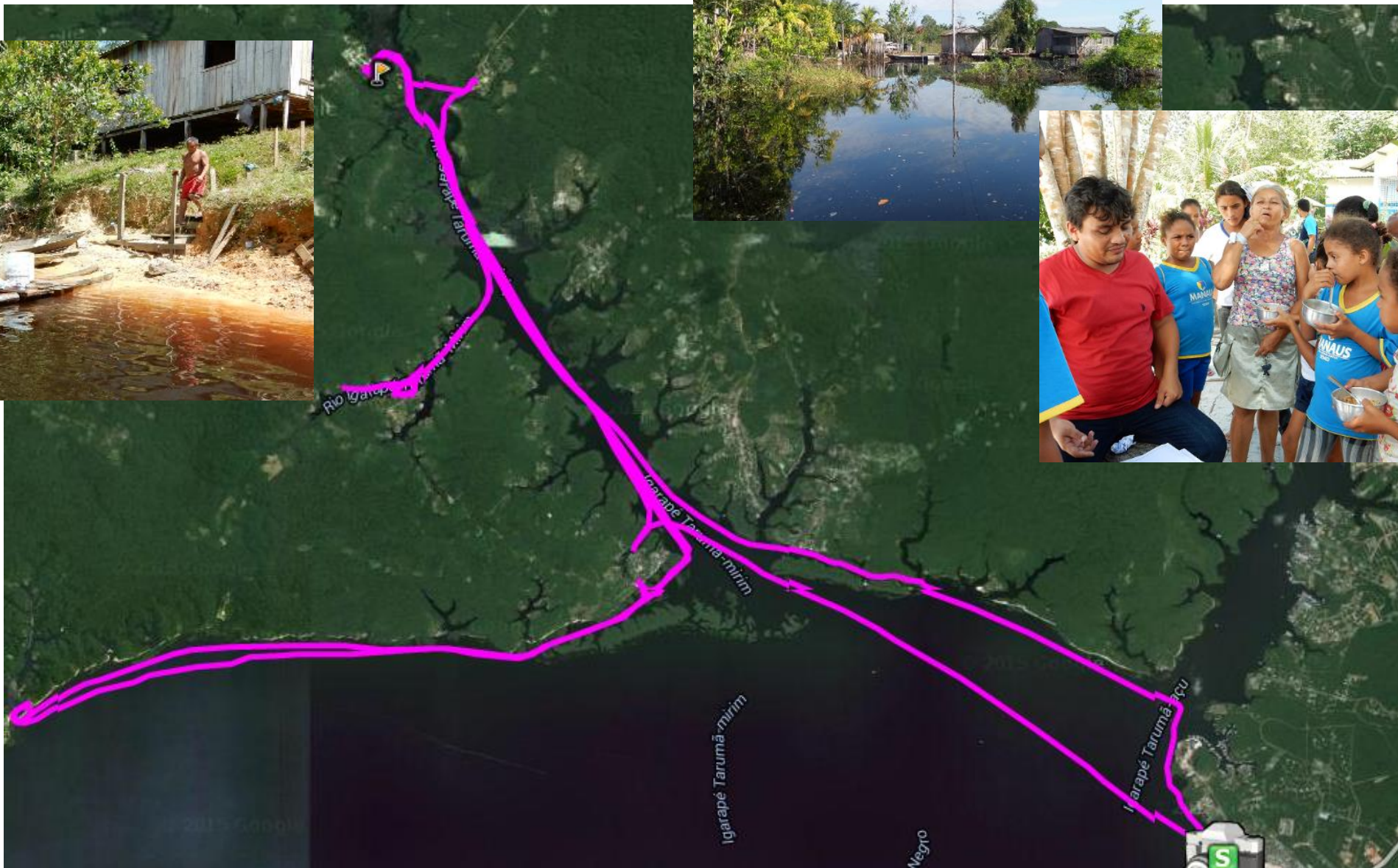
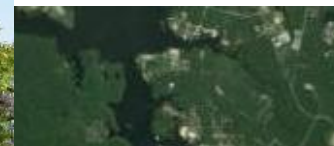


Tupé Sustainable Development Reserve





Rio Negro Communities



Digital Epidemiology with RasPiNet

- Rhythm and Randomness in Human Contact:
<http://arxiv.org/abs/1009.3980>, 2010
- On Joint Diagonalisation for Dynamic Network Analysis:
<http://arxiv.org/abs/1110.1198>, 2011
- EpiMap: Towards Quantifying Contact Networks for Understanding
Epidemiology in Developing Countries: Elsevier Ad Hoc Networks Journal:
Special Issue on Wireless Technology for Humanitarian Relief, 2014

**RasPiNET: Decentralised Network for Data
Collection and Communication with Raspberry Pi**

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<http://www.cl.cam.ac.uk/~ey204>

