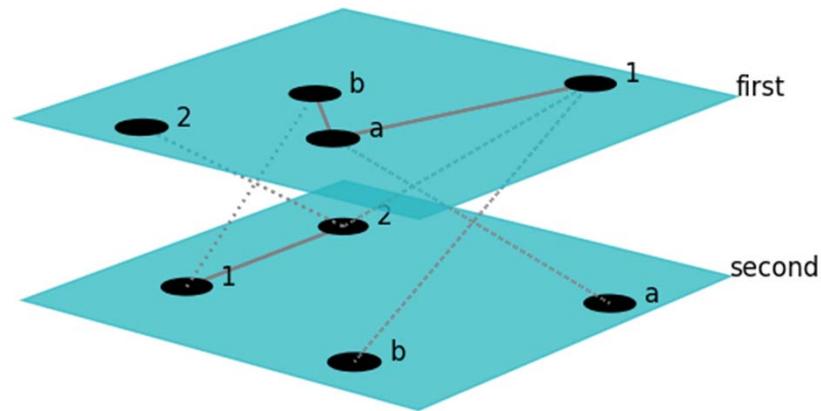


Multilayer Networks



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Multilayer networks

- Networks that are composed of multiple layers (tautological...)
- A number of different names:
 - Multilayer networks
 - Multiplex networks
 - Interconnected networks
 - Interdependent networks
 - Networks of networks
 - etc...

Heads up

- This is even newer than temporal network research
- Tools for multilayer network research still under active development
 - But there are some tools now available 😊
- More math/theory-driven, and involve more dynamical models, than temporal network research

Examples

- **Social network that involve different types of connections**
- **Network of airports connected by different air carriers**
- **Multiple infrastructures of a nation that are connected to each other**
- **Species interaction patterns that involve distinct layers of habitat**

Fundamentals of Multilayer Networks

General representation

- To represent a multilayer network, define the following function:

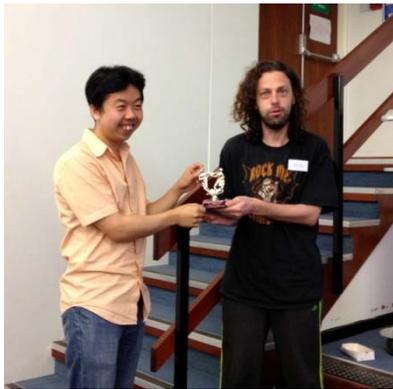
$$a(i, \alpha, j, \beta) = 0 \text{ or } 1 \text{ (or weight)}$$

- $a(i, \alpha, j, \beta) = 1$ if node i in layer α is connected to node j in layer β , otherwise 0
(This can be represented mathematically by an adjacency tensor)

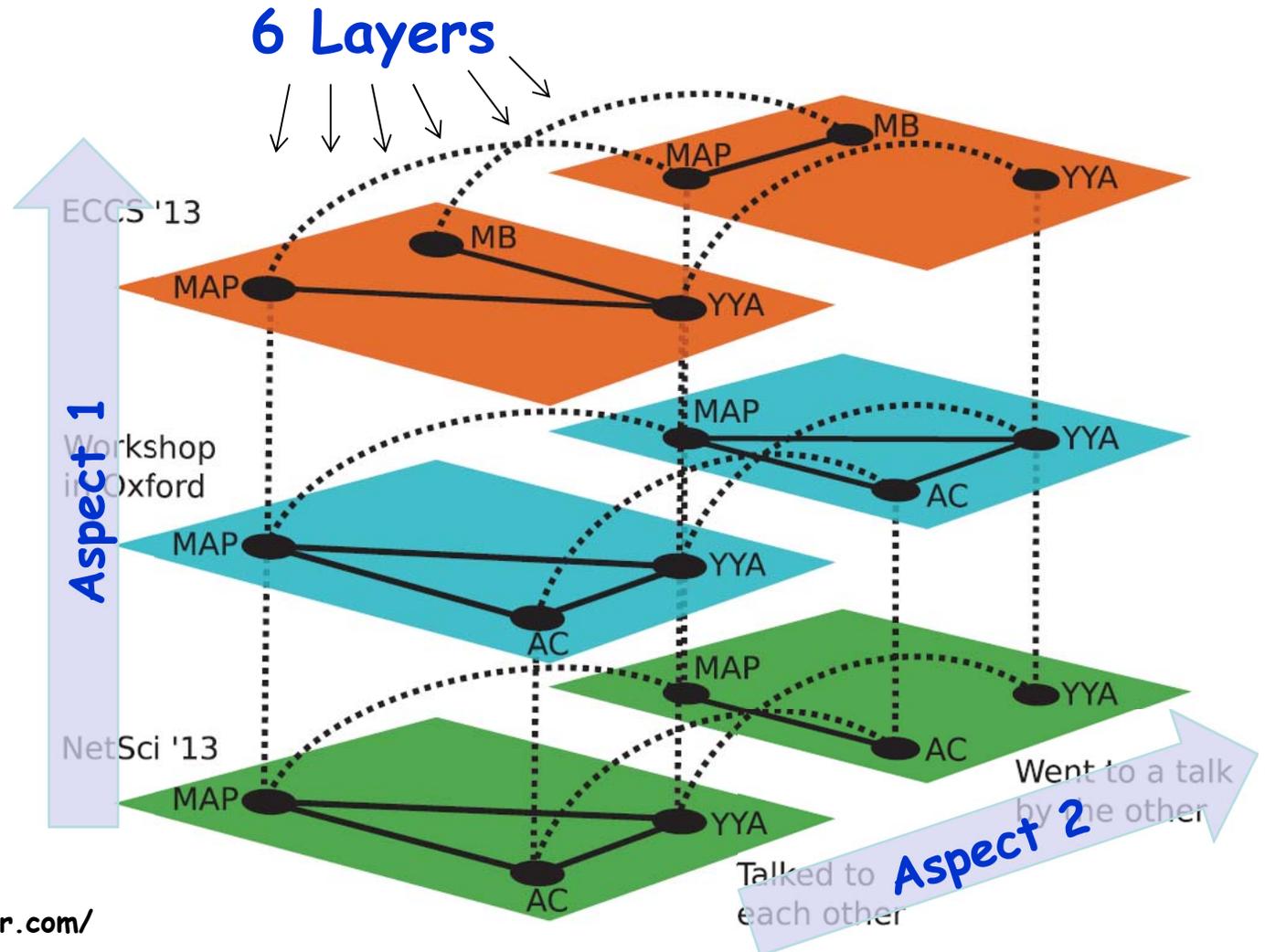
Aspects

- Layers α , β can be mapped to a space made of multiple “aspects”
 - Time, location, type of connection, etc.
 - This means α , β can be vectors
- Aspects help organize the relationships between different layers

Example: Zachary's Karate Club Club (ZKCC) network



<http://networkkarate.tumblr.com/>



Source: Kivelä et al. (2014)

Exercise

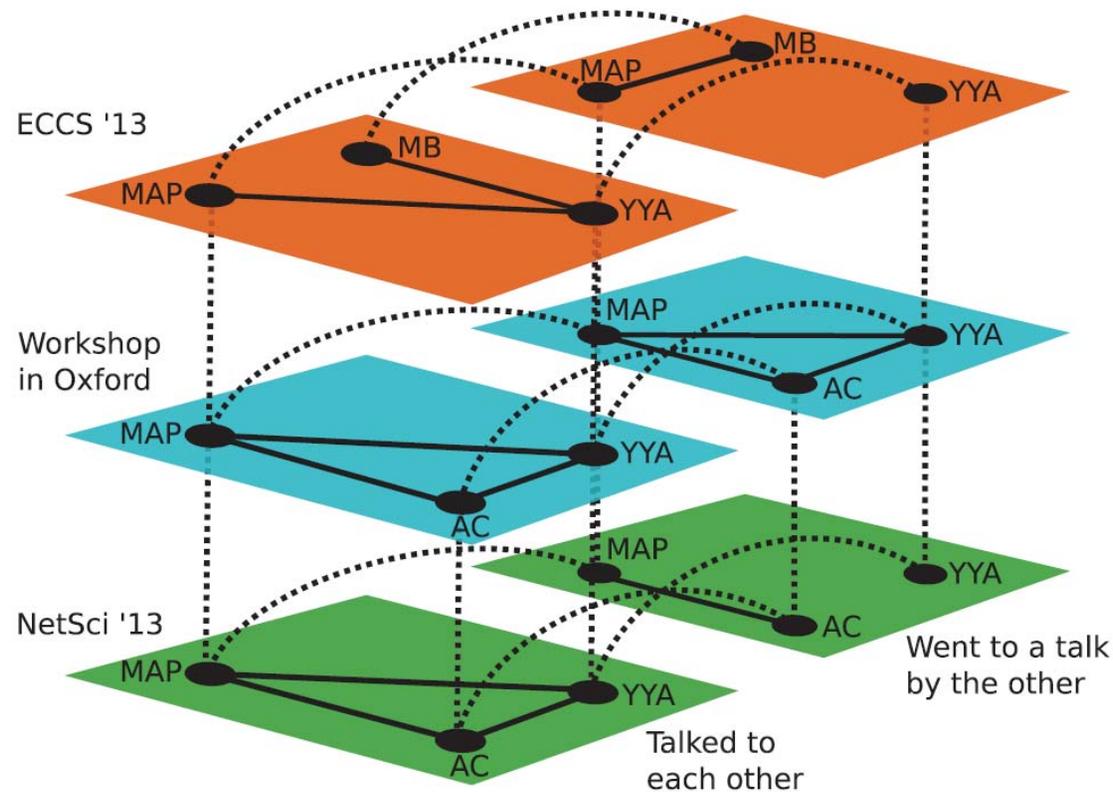
- Write down the values of the adjacency function $a(i, \alpha, j, \beta)$ for some edges in the previous ZKCC network
- Represent the same information in an edge list

Different classes of edges

- Intra-layer edge $[(i, \alpha), (j, \alpha)]$
- Inter-layer edge $[(i, \alpha), (j, \beta)] \quad \alpha \neq \beta$
 - In particular:
Coupling edge $[(i, \alpha), (i, \beta)] \quad \alpha \neq \beta$

Exercise

- Identify (a) intra-layer edges, (b) inter-layer edges and (c) coupling edges in this multilayer network

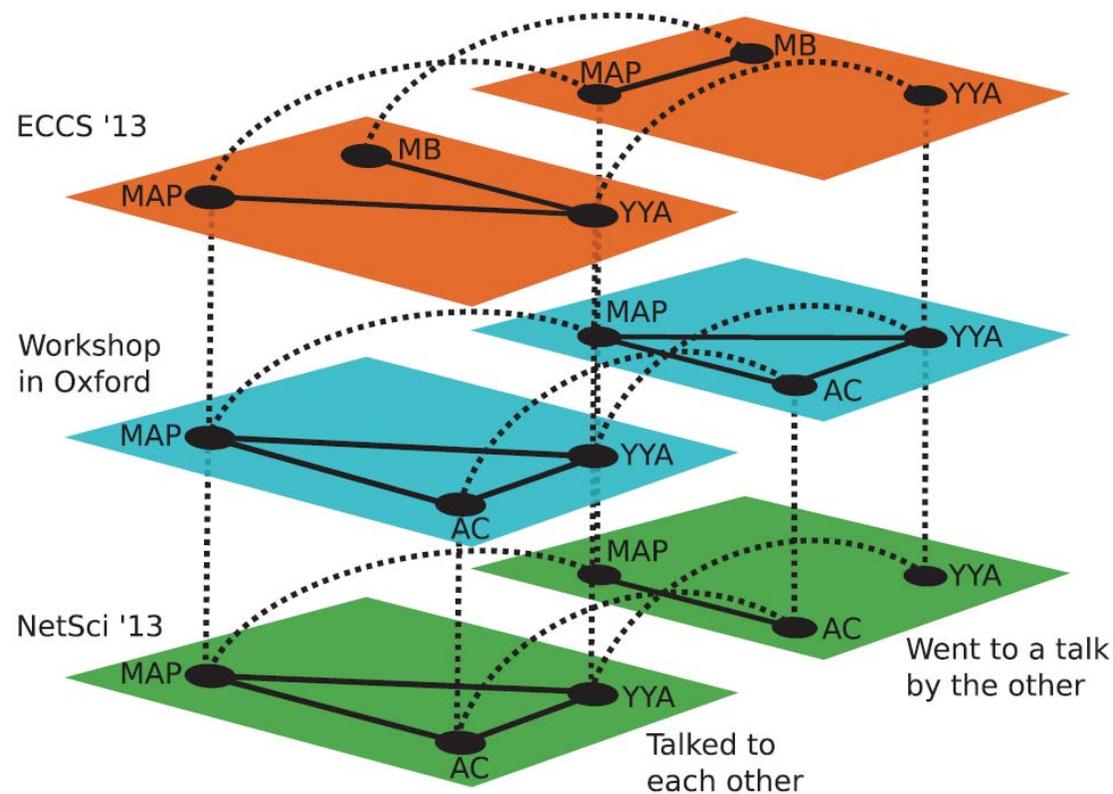


Properties of multilayer networks

- Do all the layers contain the same set of nodes? → **node-aligned**
- Or, does each node appear only in one layer? → **layer-disjoint**
- Are all the inter-layer edges coupling ones? → **diagonally coupled**
 - Are the diagonal coupling edges independent of nodes? → **layer-coupled**

Exercise

- Is this network (a) node-aligned or layer-disjoint?
(b) diagonally coupled? (c) layer-coupled?

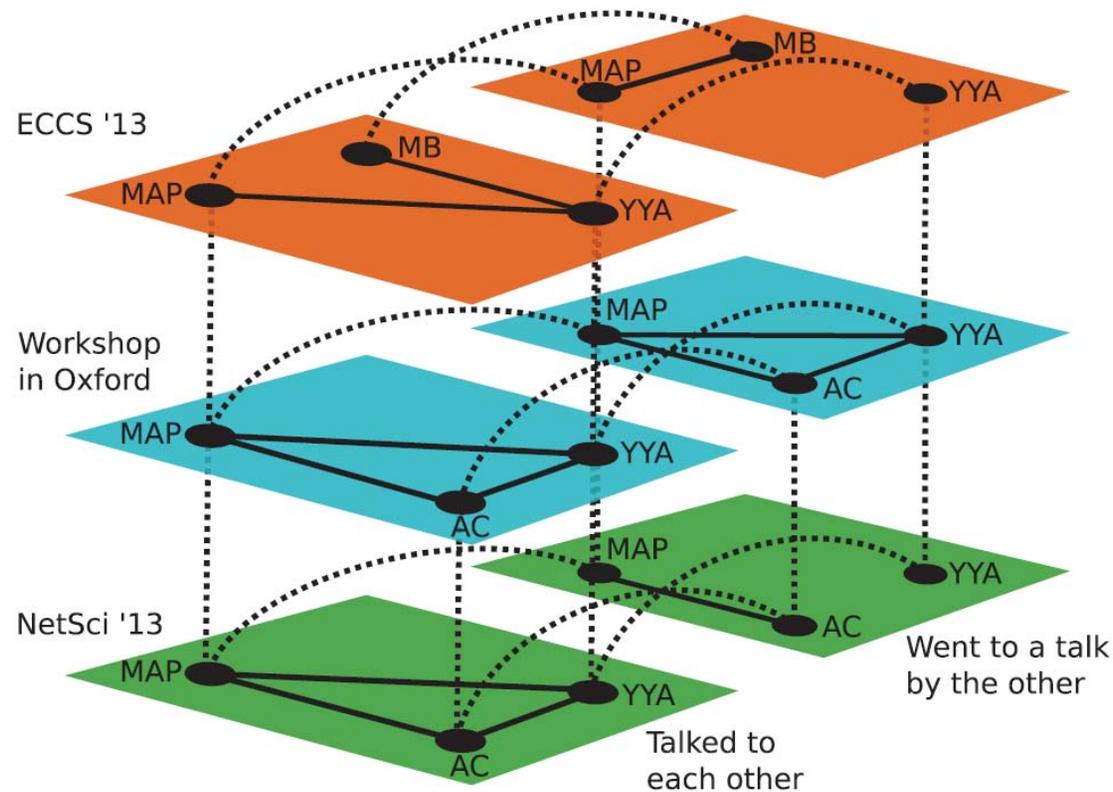


Different kinds of diagonal coupling along an aspect

- Do the diagonal edges connect a node to its counterparts in all other layers along an aspect? → **categorical**
- Or, do the diagonal edges connect a node to its counterparts only in “nearby” layers along an aspect?
→ **ordinal**

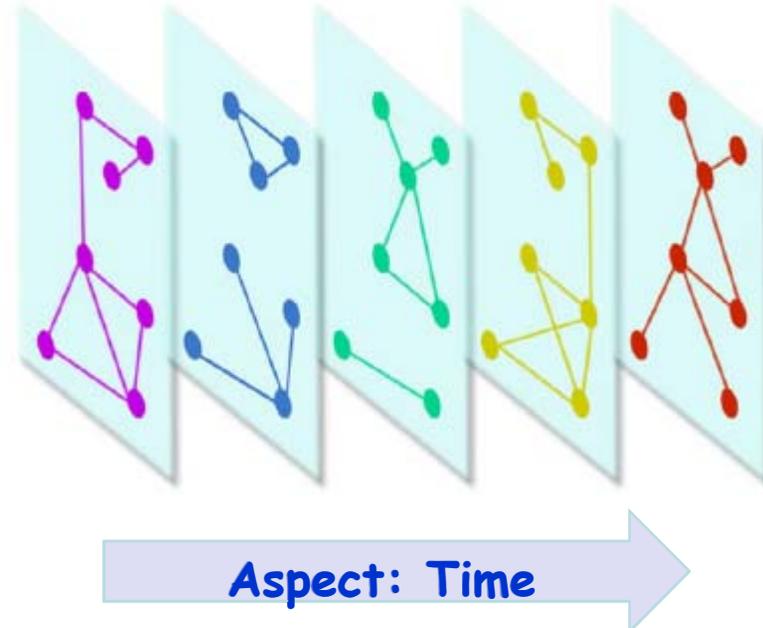
Exercise

- Does each of the two aspects of this network (a) categorical or (b) ordinal?



Temporal networks as multilayer networks

- Temporal networks can also be considered a special case of multilayer networks that have only one ordinal aspect: **time**



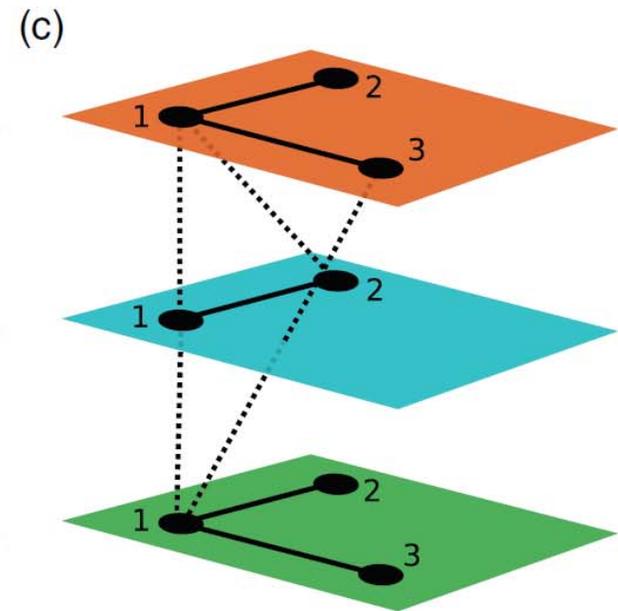
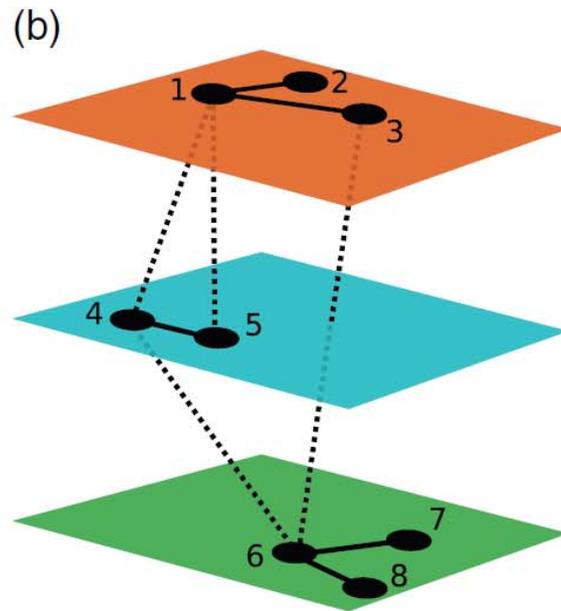
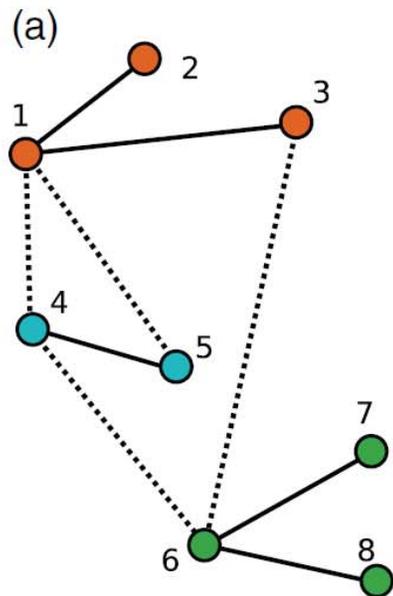
Source: De Domenico
et al. (2013)

Two major subclasses of multilayer networks

- **Node-colored networks**
(a.k.a. “interconnected networks”, “interdependent networks”, “networks of networks”)
- **Edge-colored networks**
(a.k.a. “multiplex networks”, “multirelational networks”)

Node-colored networks

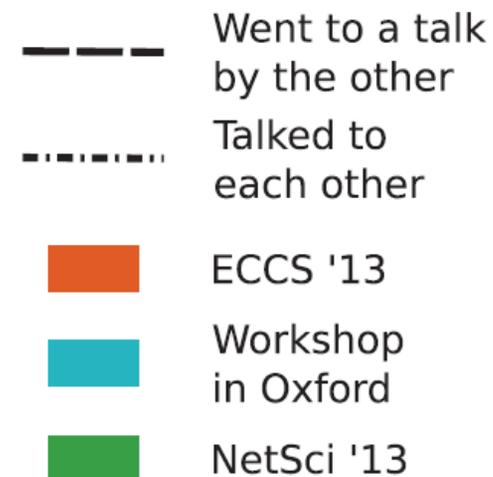
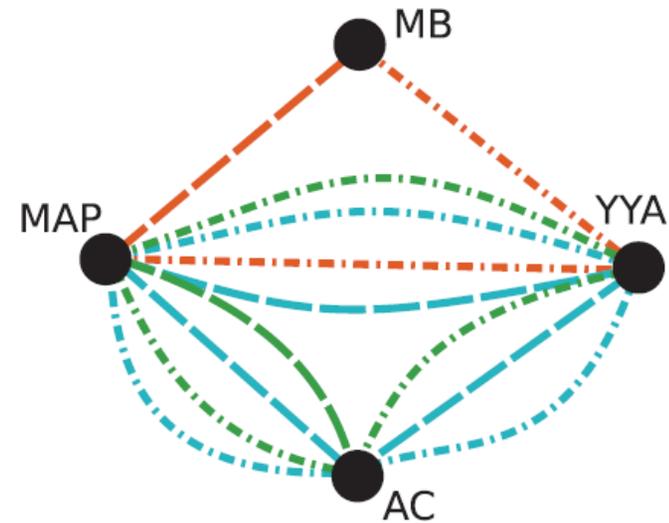
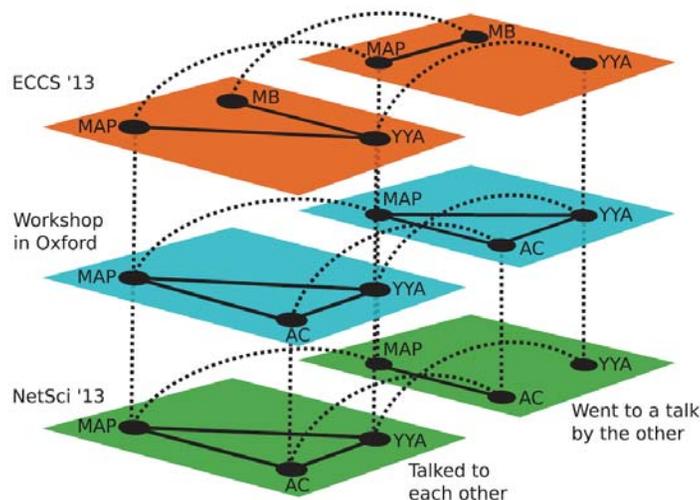
- Interconnected networks, interdependent networks, networks of networks, etc.



Source: Kivela et al. (2014)

Edge-colored networks

- Multiplex networks, multirelational networks, etc.



Source: Kivelä et al. (2014)

Exercise

- Give a few real-world examples of (a) node-colored and (b) edge-colored multilayer networks
- Discuss properties of those networks, especially the types of their inter-layer connections

Computational Modeling and Analysis of Multilayer Networks

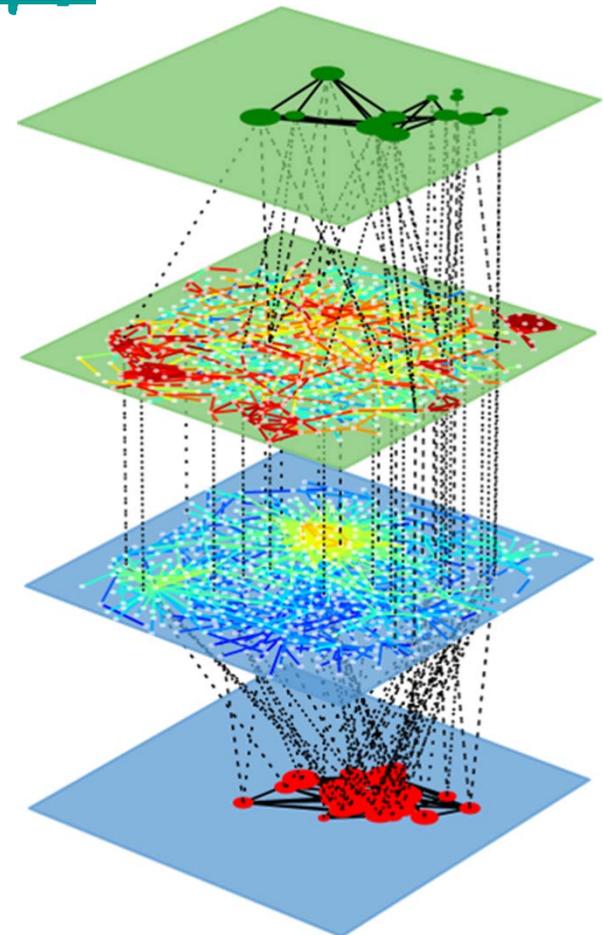
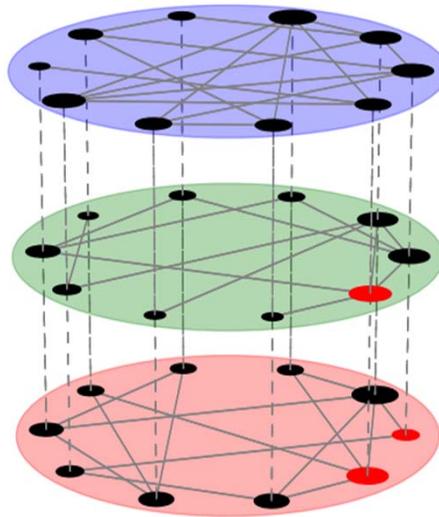
PlexMath project

- <http://www.plexmath.eu/>

The screenshot shows the PlexMath website homepage. At the top, there is a navigation menu with the following items: HOME, PEOPLE, PUBLICATIONS, RESOURCES » (expanded), and NEWS. The 'RESOURCES' dropdown menu is open, and the 'Software' option is circled in yellow. Below the navigation menu, there is a large banner for 'SCIENTIFIC DATA' featuring a MuxViz visualization of a multi-modal transportation system. The banner text reads: 'MuxViz visualization of multi-modal transportation system featured on Nature Scientific Data'. To the right of the banner is a vertical stack of three maps showing different transportation modes: Rail, Metro, and Bus. Below the banner is a 'Nature Scientific Data' section with a 'More »' link. The footer of the page includes a 'Home' section with a date of June 26, 2013, and an 'Abstract' section. A search bar and a 'NEWS' section are also visible.

Pymnet library

- <http://www.mkivela.com/pymnet/>
 - Developed by Mikko Kivelä



Exercise

- Download the pymnet library and its documentation
- Follow its tutorial to learn how to build multilayer/multiplex networks
 - Place the "pymnet" folder to your working directory, or add it to your Python path
 - Commands are different from NetworkX, but access to all NetworkX commands are provided under "nx." prefix

Pymnet's data structure

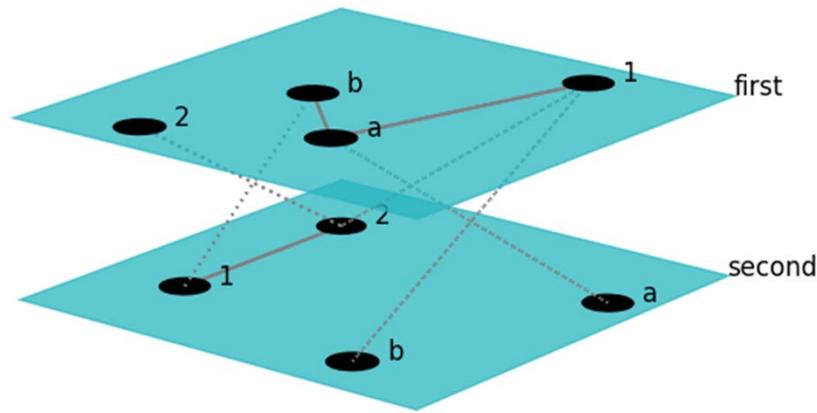
- **MultilayerNetwork(aspects = **)**
 - To represent general multilayer networks
- **MultiplexNetwork(couplings = **)**
 - To specifically represent multiplex networks
 - **nx.Graph(g)** converts these into a NetworkX Graph (monolayer only)

Pymnet: Basics

- `add_node()`, `add_layer()`
- `g[ID][ID] = 1` adds edge (ID: sequence of indices, including aspects)
- `list(g)` gives node list
- `list(g.iter_**())` gives node, layer, or node-layer list
- `list(g.edges)` gives edge list
- `list(g[ID])` gives neighbor list
- `g[ID].deg()` or `.str()` gives degree or strength

Drawing multilayer networks

- **Just `draw(g)` !!**



- If you draw in a separate window, you can click and drag to rotate it
- Many options are available to customize visualization results

Exercise

- Check out the references of pymnet
 - <http://www.mkivela.com/pymnet/reference.html>
- Create a few multilayer/multiplex networks and visualize them
- Explore customization of visualization as you like

Get the data!

- <http://www.plexmath.eu/>

The screenshot shows the Plexmath website interface. At the top, there is a navigation menu with the following items: HOME, PEOPLE, PUBLICATIONS, RESOURCES, and NEWS. The RESOURCES menu is expanded, and the 'Data' option is highlighted with a yellow circle. Below the navigation menu, there is a banner for 'SCIENTIFIC DATA' featuring a MuxViz visualization of a multi-modal transportation system. The banner includes the text: 'MuxViz visualization of multi-modal transportation system featured on Nature Scientific Data'. To the right of the banner is a 3D visualization of a transportation network with nodes and edges, labeled 'Rail' and 'Metro'. Below the banner, there is a section titled 'Nature Scientific Data' with a 'More »' link. At the bottom of the page, there is a 'Home' section with a date 'June 26, 2013' and a 'Home' link. A search bar is located on the right side of the page, and a 'NEWS' section is visible below it.

Exercise

- Download the “**Vickers & Chan 7th graders social network**” data from the PlexMath website
- Write a Python code to read the data and construct a multilayer network
 - As a **MultilayerNetwork** object
 - As a **MultiplexNetwork** object
- Visualize it

Some built-in measurements

- `degs(g)` gives degree distribution
- `density(g)` gives network density
- `multiplex_degs(g)` for each layer
- `multiplex_density(g)` for each layer
- Various clustering measurements
 - But not so many functions available yet...
(the author Mikko said he would welcome any contributions!)

Exercise

- Plot the degree distribution for each layer in the Vickers & Chan 7th graders multiplex network
- Measure the density of each layer and determine which layer was most/least dense

Other network measurements

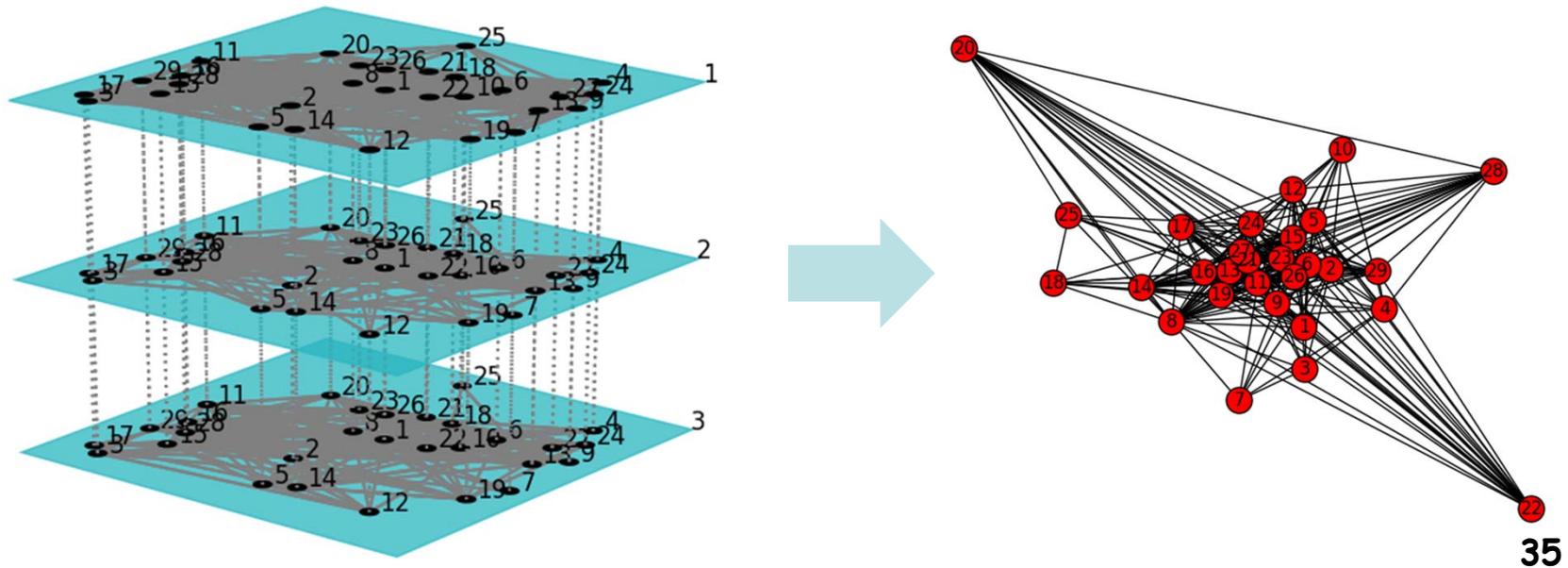
- You can represent a multilayer network as a plain monolayer network and apply various measurements
- Is it a multiplex network and does changing layers take no cost or distance?
 - Yes to both → Aggregate the network
 - Otherwise → Flatten the network

Note: When you create a monolayer network...

- Consider how the weights of intra-layer edges compare to each other between different layers
- Each layer may represent different connectivity; adjust their weights as needed
 - E.g. :
Best friend, friend, acquaintance
Flight, train, bus, bike, walk

Aggregating layers

- `aggregate(g, aspects)` creates a simple network aggregated over aspects
 - Works for both multilayer and multiplex, but most meaningful for multiplex

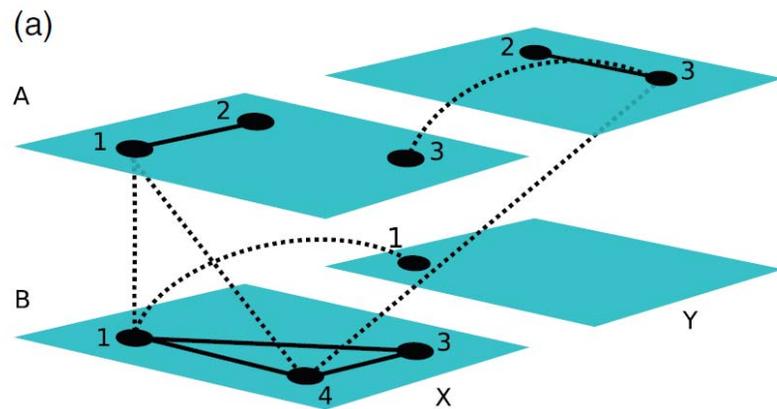


Exercise

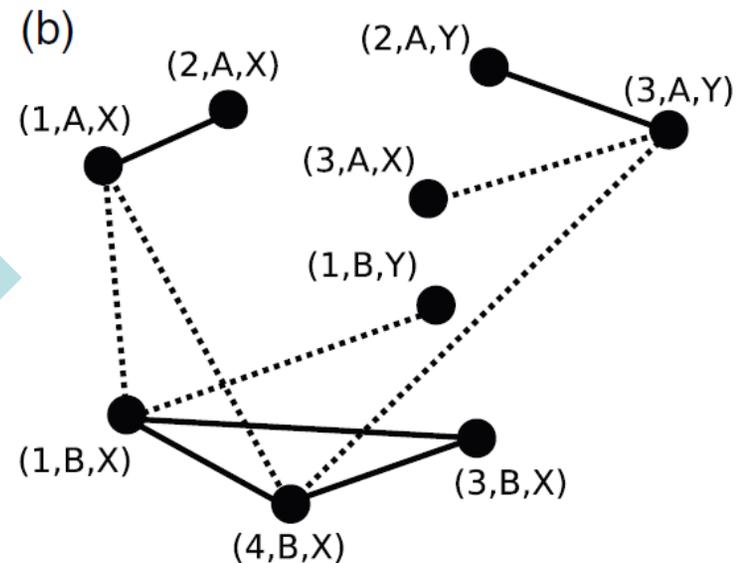
- Create an aggregated network of the Vickers & Chan 7th graders multiplex network
- Calculate several shortest paths, average shortest path length, and several centralities of the nodes
- Detect communities using the Louvain method

“Flattening” multilayer networks

- Ignore layers and represent all the connections in a plain network



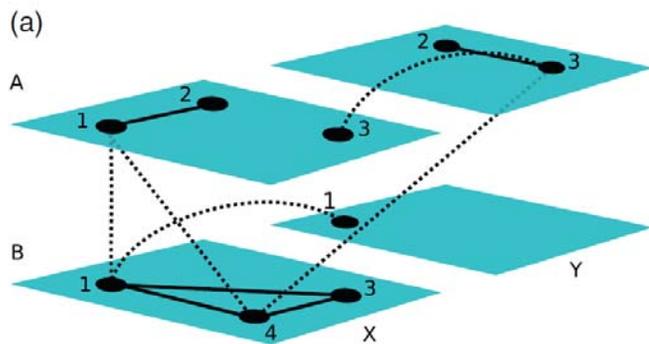
Source: Kivelä et al. (2014)



- (This is essentially the same as the construction of a space-time network for a temporal network)

Supra-adjacency matrix

- An adjacency matrix of a flattened version of a multilayer network



If inter-layer edges are “diagonal” couplings, the inter-layer matrices are diagonal too (hence the name)

		A		B					
		X	Y	X	Y				
A	X	Adj. matrix of (A,X)	Inter-layer matrix	Inter-layer matrix	Inter-layer matrix	X	A		
	Y	Inter-layer matrix	Adj. matrix of (A,Y)	Inter-layer matrix	Inter-layer matrix			Y	
	X	Inter-layer matrix	Inter-layer matrix	Adj. matrix of (B,X)	Inter-layer matrix			X	B
	Y	Inter-layer matrix	Inter-layer matrix	Inter-layer matrix	Adj. matrix of (B,Y)			Y	

Supra-adjacency matrix in pymnet

- `supra_adjacency_matrix(g)` creates the supra-adjacency matrix of `g` and a list of node IDs (arranged in the order used in the matrix)
- `nx.from_numpy_matrix(A)`, with `A` being the supra-adjacency matrix, creates a flattened monolayer network

Exercise

- Create a flattened network of the Vickers & Chan 7th graders multiplex network
- Calculate several shortest paths, average shortest path length, and several centralities of the nodes
- Detect communities using the Louvain method

Inter-layer measurement

- “Interdependence”

Ratio of the number of shortest paths that use multiple layers to the total number of shortest paths

High interdependence

→ Communication/transportation in a multilayer network uses multiple layers more often (i.e., those layers depend on each other)

Exercise

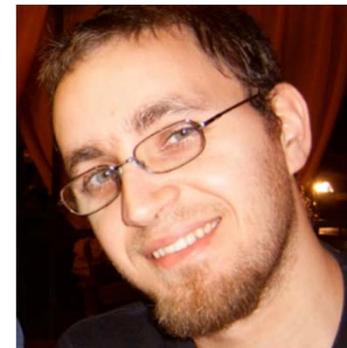
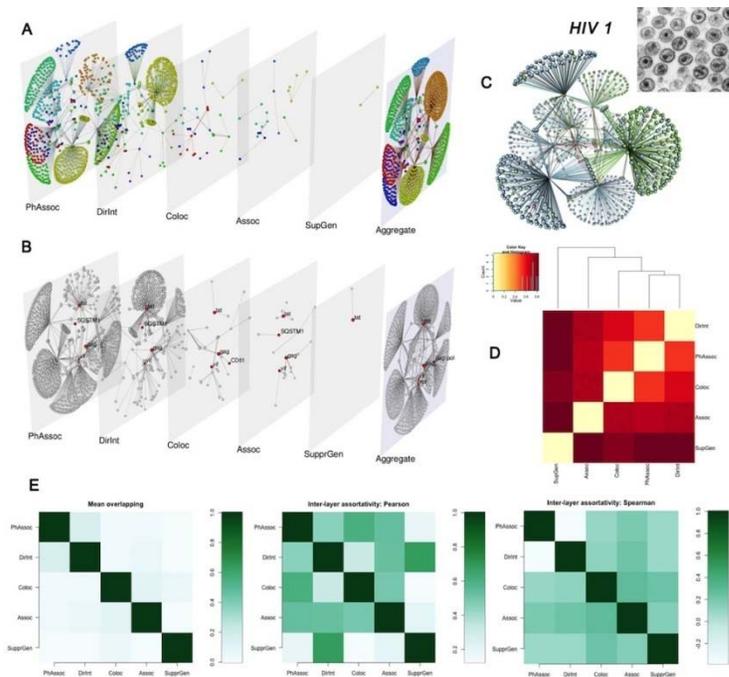
- Design and implement a code to measure the level of interdependence in the Vickers & Chan 7th graders multiplex network
- Apply the code to other multilayer network data

FYI: If you use R...

- **muxViz**

- <http://muxviz.net>

- http://www.youtube.com/watch?v=gcpYSdi_xI



Developed by
Manlio De Domenico

Dynamical Processes on Multilayer Networks

Dynamics on multilayer networks

- Several dynamical models have been studied recently
- Mostly focused on **two-layer multiplex networks**

Cascade of failures

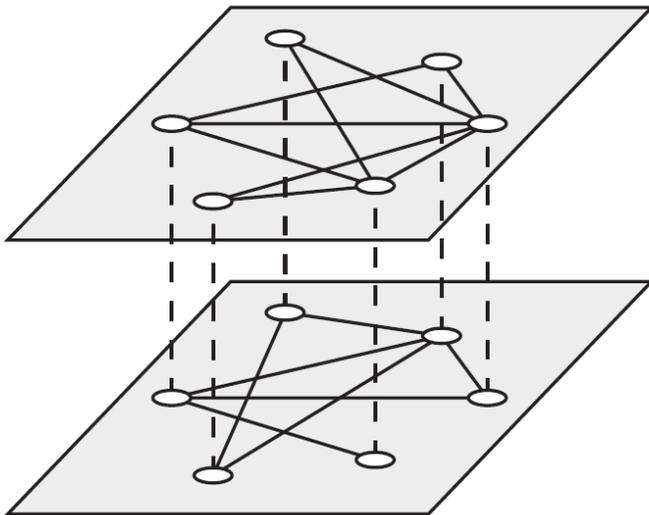
- Buldyrev, S. V., Parshani, R., Paul, G., Stanley, H. E., & Havlin, S. (2010). Catastrophic cascade of failures in interdependent networks. *Nature*, 464(7291), 1025-1028.



- Proposed a model of two-layer interdependent network (e.g., power grid and the Internet)
- Fragmentation by a cascade of failures occurs quite differently on this model than monolayer networks

Diffusion

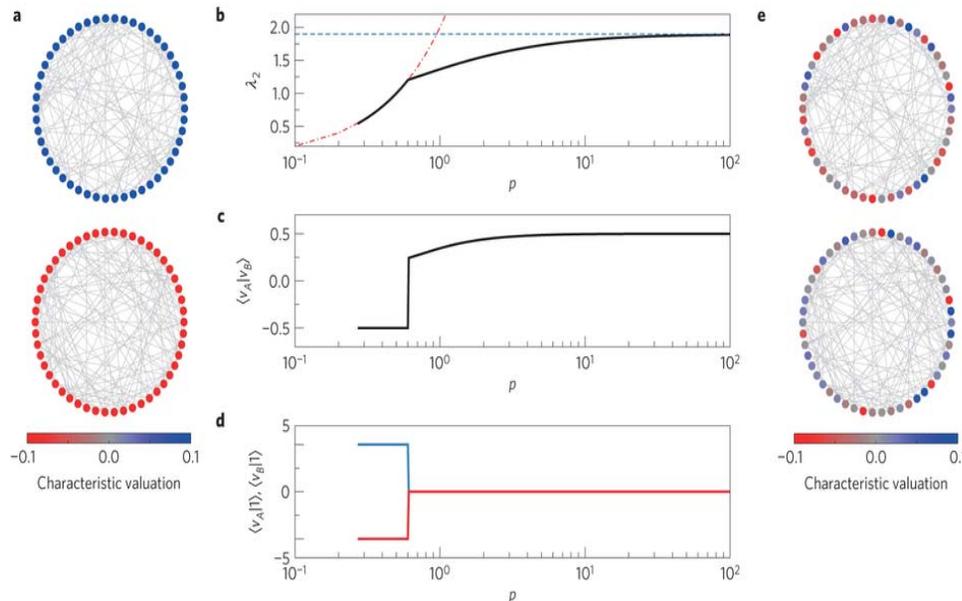
- Gomez, S., Diaz-Guilera, A., Gomez-Gardeñes, J., Perez-Vicente, C. J., Moreno, Y., & Arenas, A. (2013). Diffusion dynamics on multiplex networks. *Physical Review Letters*, 110(2), 028701.



- Studied diffusion processes on two-layer multiplex networks
- Analyzed the spectrum of a “supra-Laplacian” matrix of the network

Structural transition

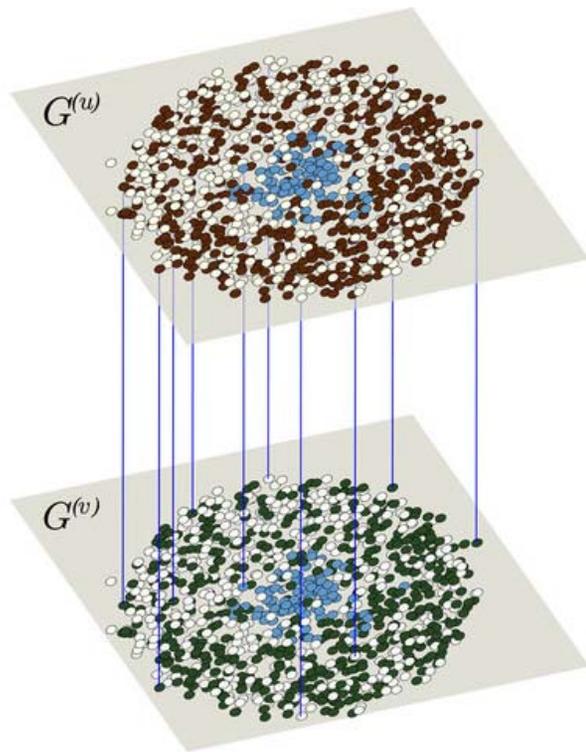
- Radicchi, F., & Arenas, A. (2013). Abrupt transition in the structural formation of interconnected networks. *Nature Physics*, 9(11), 717-720.



- Discovered that the algebraic connectivity and Fiedler's vector undergo a sharp transition as inter-layer connectivity is increased on two-layer multiplex networks

Pattern formation

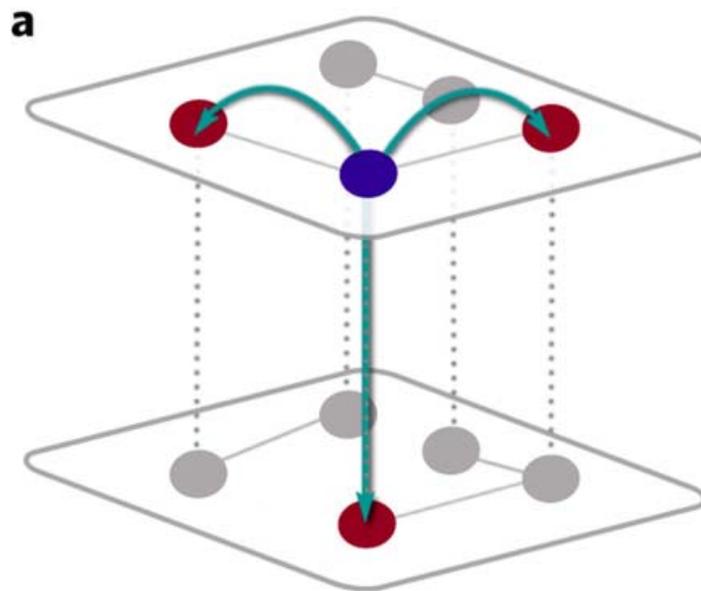
- Kouvaris, N. E., Hata, S., & Díaz-Guilera, A. (2015). Pattern formation in multiplex networks. *Scientific reports*, 5, 10840.



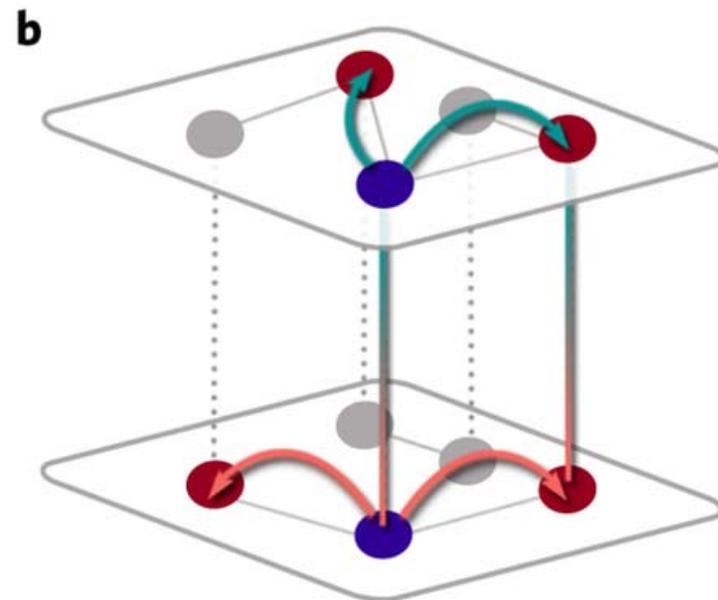
- Studied reaction-diffusion dynamics on two-layer multiplex networks
- Showed that the difference in topologies between two layers can cause pattern formation (a.k.a. Turing instability) *even if diffusion constants are the same*

Spreading

- De Domenico, M., Granell, C., Porter, M. A., & Arenas, A. (2016). The physics of spreading processes in multilayer networks. *Nature Physics* 12, 901-906.



Single dynamics



Coupled dynamics

Exercise

- Implement a “cascade of failure” model on a random multilayer network with k layers
- Conduct numerical simulations with the probabilities of intra- and inter-layer connections systematically varied
- Identify the parameter values with which large-scale failure occurs