



**PhD in Information Technology and Electrical Engineering**

**Università degli Studi di Napoli Federico II**

**PhD Student: Anna Di Meglio**

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**XXXII Cycle**

**Training and Research Activities Report – Second Year**

**“Modelling and control of  
complex dynamical evolving networks”**

**Tutor: Prof. Franco Garofalo**

**Co-Tutor: Pietro De Lellis**



### 1. GENERAL INFORMATION

Graduated in Management Engineering – Università di Napoli, Federico II  
 XXXII Cycle – ITEE – Università di Napoli, Federico II  
 M.I.U.R. grant  
 Tutor: Prof. Franco Garofalo

### 2. CREDIT SUMMARY

	Credits year 2							Summary
	Estimated	1	2	3	4	5	6	
<b>MODULES</b>	10	0	0	1	3	5	4,5	13
<b>SEMINARS</b>	5	0,8	0	0,9	1,1	1,2	1	5
<b>RESEARCH</b>	45	12	0	8	8	7	7	42
	60	13	0	9,9	12	13	13	60

 Maternity.

### 3. STUDY AND TRAINING ACTIVITIES

- i Courses attended:
  - “Applied Matrix Theory”; Prof. Jens Lorenz -External- [lorenz@math.unm.edu](mailto:lorenz@math.unm.edu) ;
  - “Convex Optimization”; Prof. David Coop -External- [dcopp@unm.edu](mailto:dcopp@unm.edu) ;
  - “Paradigmatic model in social science”; Prof. Fabio Dercole -External- [fabio.dercole@polimi.it](mailto:fabio.dercole@polimi.it)
- ii Other courses:
  - “Nonlinear control and Chaos”; Prof. Jens Lorenz -External- [lorenz@math.unm.edu](mailto:lorenz@math.unm.edu);
- iii Seminars:
  - “Optimal input placement in lattice graphs.” Isaac Samuel Klicstein;
  - “Prediction of optimal drug schedules for controlling autophagy.” Afroza Shirin;
  - “Exact controllability of complex networks” Ying-Cheng Lai;
  - “Study of cluster synchronization by means of spectral characteristics” Fabio Della Rossa;
  - “Optimal Attack Strategies for Maximizing Failures of Transmission Lines in Power Grids” Pankaz Das;
  - “MPC to efficiently control power grid” David Coop;
  - “Networks symmetries and synchronization”, Francesco Sorrentino;
  - “Optimal control in drug delivery” Francesco Sorrentino;
  - “Generating symmetric graphs.” Isaac Samuel Klickstein;
  - “Sensitivity to a new velocity/pressure-gradient model to Reynolds number” Svetlana Poroseva;
  - “Cluster synchronization analysis on multilayer networks” Fabio Della Rossa “Aging and Autophagy” Mark Mc Corminck and Micheal A. Mandell;
  - “Prediction of optimal drug schedules for controlling autophagy.” Afroza Shirin. “Ultra wide spectrum photovoltaic-thermoelectric solar cell” Tito Busani;

- “Optimal regulation of blood glucose level in type I diabetes using insulin and glucagon.” Afroza Shirin;
- “Targeted synchronization in externally driven mechanical oscillators” Karen Blaha;
- “Suenos del Coyote: the Emergence of Genizaros in the Nuevomexicano Literary imagination” Enrique lamadrid.

### 3. RESEARCH ACTIVITY

#### “Modelling and control of complex dynamical networks”

This year my research has been focused on how to face with complex dynamical networks [1] in a more realistic scenario in which the interactions among the nodes are time-varying and the information of the underlying model is incomplete.

A complex network is a set of dynamical systems coupled through a graph with non-trivial topological features. Whichever the systems and their interactions are complex, their first attempt model is often borrowed from linear system theory and follows the linear dynamics

$$\dot{x}(t) = A(t)x(t) + B(t)u(t) \quad (*)$$

where  $A(t)$  is the adjacency matrix of the generic graph  $\mathcal{G}$ ,  $B(t)$  is the drivers' matrix and  $u(t)$  is the control input. The types of complex networks belonging to this class is enough large to motivate the study in terms of equation (\*) and moreover, this formulation is helpful also in nonlinear formulations as it can be used after the linearization process, where needed.

The classical hypothesis in studying complex networks through (\*) are:

1. The network topology is static, the number of nodes and the edges' weights are fixed over the time horizon control, *i.e.* the adjacency matrix is time invariant  $A(t) = A$ .
2. The knowledge of the network is complete.

Both the above simplified hypotheses are at the base of the widely studied minimum control energy problem [3-5 and references therein]. The control goal is that to find an input  $u(t)$  for the complex network (\*) that solves the following optimization problem

$$\begin{aligned} \min \int_0^T u(t)^T u(t) dt \\ \text{s. t.} \\ \dot{x} = Ax + Bu \\ x(0) = x_0 \\ x(T) = x_T \end{aligned} \quad (1)$$

where the objective function represents the control energy needed to steer the network from a specified initial condition ( $x_0$ ) to a desired final condition ( $x_f$ ). The systems for which neither hypothesis 1. or hypothesis 2. hold are ubiquitous and the list of the applications is very long. For example, we see them in biological (protein-protein interaction, gene regulatory, metabolic and neuronal networks) [13-14]; technological (internet network; power grid system; transportation and distribution networks; sensor network) [12] and social systems (financial markets; social networks; scientific collaboration networks; influence spreading; leadership; political comping; opinion dynamics) [8-11].

Nevertheless, sometimes adding more information about times of interactions can make predictions and so control strategies more accurate. At the same way, consider a scenario with incomplete information makes the selected control strategy more “robust”. Therefore, inspired by [2] we have formulated problem (1) for temporal networks [7], *i.e.* the adjacency matrix is time varying  $A = A(t)$ , and it is affected by stochasticity, *i.e.* the adjacency matrix over the control horizon is the realization

of a stochastic process. In this more realistic scenario we aim to investigate if the fundamental energy control advantages listed in [2] are still effective and which are the possible blinded advantages of the “temporality” of the network. In the second year of my PhD I began the study of the literature on the minimum control energy of complex networks and I began the numerical investigations supporting some prelaminary theoretical conjectures.

- [1] Boccaletti S., Latora V., Moreno Y., Chavez M., Hwang D.-U. “Complex networks: structure and dynamics” *Physics Reports* 424, 4-5 (175-308) 2006.
- [2] Li, A., Cornelius, S. P., Liu, Y. Y., Wang, L., & Barabási, A. L. (2017). The fundamental advantages of temporal networks. *Science*, 358(6366), 1042-1046.
- [3] B. Barzel J.-J. Slotine Y.-Y. Liu G. Yan, G. Tsekis and A.-L. Barabási. Spectrum of controlling and observing complex networks. *Nature Physics*, 11(9):779, 2015.
- [4] I. Klickstein, A. Shirin, and F. Sorrentino. Energy scaling of targeted optimal control of complex networks. *Nature communications*, 8:15145, 2017.
- [5] J. Sutter and A. E. Motter. Controllability transition and nonlocality in network control. *Physical review letters*, 110(20):208701, 2013.
- [6] A. Li, S. P. Cornelius, Y.-Y. Liu, L. Wang, and A.-L. Barabási. Control energy scaling in temporal networks. arXiv preprint arXiv:1712.06434, 2017.
- [7] Holme, P., & Saramäki, J. (2012). Temporal networks. *Physics reports*, 519(3), 97-125.
- [8] Iribarren, J. L., & Moro, E. (2009). Impact of human activity patterns on the dynamics of information diffusion. *Physical review letters*, 103(3), 038702.
- [9] De Lellis, P., Di Meglio, A., & Iudice, F. L. (2018). Overconfident agents and evolving financial networks. *Nonlinear Dynamics*, 92(1), 33-40.
- [10] DeLellis, P., DiMeglio, A., Garofalo, F., & Iudice, F. L. (2017). Steering opinion dynamics via containment control. *Computational social networks*, 4(1), 12.
- [11] DeLellis, P., DiMeglio, A., Garofalo, F., & Iudice, F. L. (2017). The evolving cobweb of relations among partially rational investors. *PLoS one*, 12(2), e0171891.
- [12] Cattuto C, Van den Broeck W, Barrat A, Colizza V, Pinton J-F, Vespignani A (2010) Dynamics of Person-to-Person Interactions from Distributed RFID Sensor Networks.
- [13] G. Chechik, E. Oh, O. Rando, J. Weissman, A. Regev, D. Koller, Activity motifs reveal principles of timing in transcriptional control of the yeast metabolic network, *Nature Biotechnol.* 26 (2008) 1251–1259.
- [14] F. de Vico Fallani, V. Latora, L. Astolfi, F. Cincotti, D. Mattia, M.G. Marciani, S. Salinari, A. Colosimo, F. Babiloni, Persistent patterns of interconnection in time-varying cortical networks estimated from high-resolution eeg recordings in humans during a simple motor act, *J. Phys. A* 41 (2008) 224014.

## 4. PRODUCTS

- i. Conference Paper:
  - DeLellis P., DiMeglio A., Garofalo F., Lo Iudice, F. (2018, October). “**Partial containment control over signed graphs.**” Submitted for *European Control Conference (ACC), 2019 IEEE*.
  - DiMeglio A., Dercole F., Della Rossa F. (2018, October). “**Direct reciprocity and model-predictive rationality: A setup for network reciprocity over social ties.**” Submitted for *European Control Conference (ACC), 2019 IEEE*.
- ii. In preparation:
  - “The fundamental challenge of temporal networks”
  - “Containment control of large signed networks.”
  - “Fast leader-following consensus”.

### 5. ACTIVITY ABROAD

- Albuquerque, NM (USA) University of New Mexico, Department of Mechanical Engineering. Research activity under the supervision of Professor Francesco Sorrentino ([fsorrent@unm.edu](mailto:fsorrent@unm.edu)).

### 6. TUTORSHIP

- Co-supervision of thesis master students for the course of Identificazione dei modelli e controllo ottimo.
- TOT. HOURS: 10.