



Politecnico di Torino - Turin (Italy)
Aerospace Engineering Department



Characterization of turbulent channel flows: From time-series to complex networks

Presenting Author: Giovanni Iacobello

Co-Authors: Stefania Scarsoglio, Luca Ridolfi

Introduction

Data: Numerical simulations / experimental measures

➔ Fully developed turbulent channel flow

Method: Application of complex networks to time-series

➔ Visibility Algorithm

Aim: Novel method to analyze turbulence data

➔ Extract non-trivial information: temporal structure

Complex Networks

Statistical Physics + Graph Theory



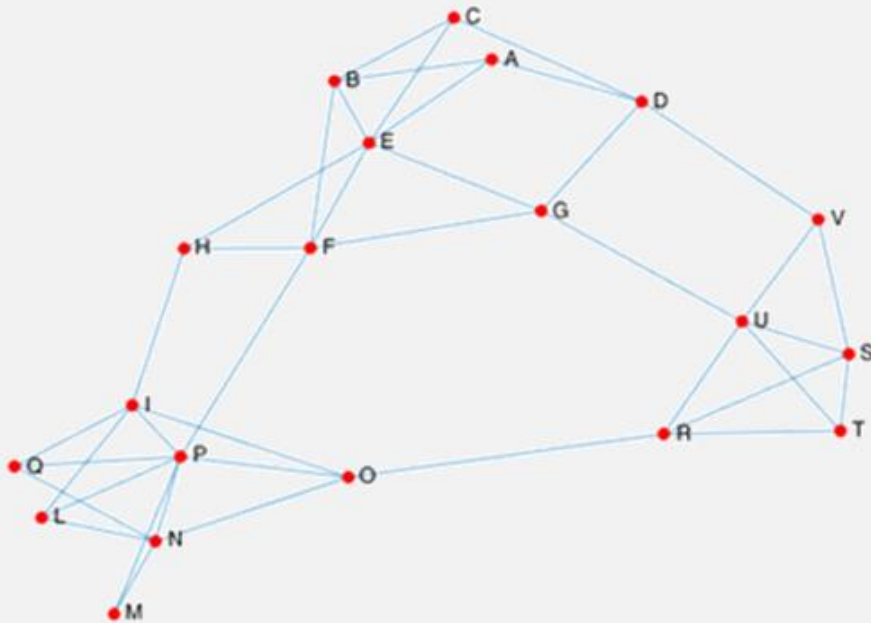
Nodes + Links

Complex Networks

Statistical Physics + Graph Theory



Nodes + Links



Non-trivial topological features

Complex Networks

Statistical Physics + Graph Theory



Nodes + Links

Successfully applied:

- ▶ Transportation
- ▶ Internet / WWW
- ▶ Biology/Medicine
- ▶ Economy
- ▶ Earth science
- ▶ Sociology
- ▶ Fluid Flows

Complex Networks

Statistical Physics + Graph Theory



Nodes + Links

Successfully applied:

- ▶ Transportation
- ▶ Internet / WWW
- ▶ Biology/Medicine
- ▶ Economy
- ▶ Earth science
- ▶ Sociology
- ▶ Fluid Flows



Turbulence

Very recently

Networks Building

Each series value \equiv a node

How are nodes linked together?

Networks Building

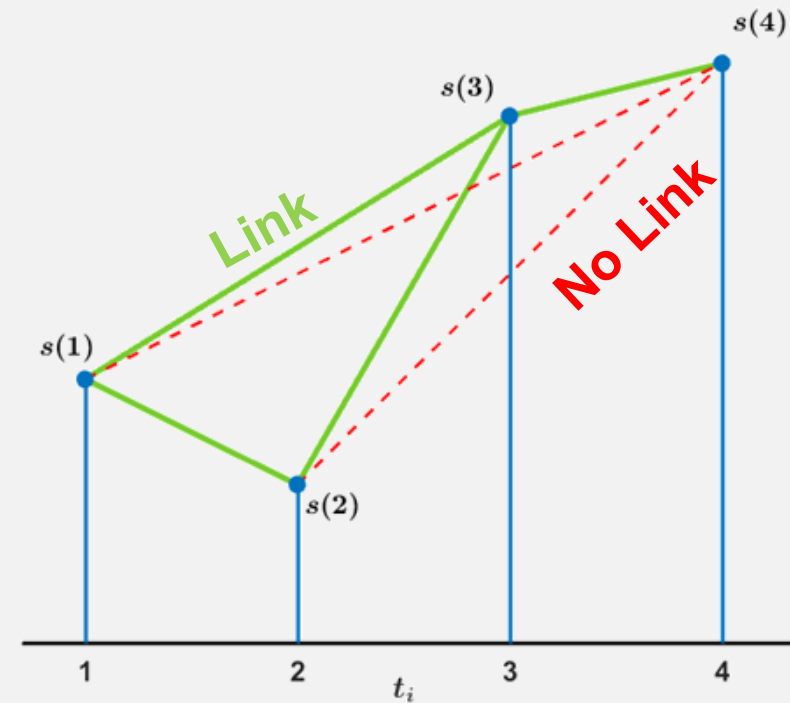
Each series value \equiv a node

How are nodes linked together?



Visibility Algorithm [2]

$$s(t_k) < s(t_j) + (s(t_i) - s(t_j)) \frac{t_j - t_k}{t_j - t_i}$$



Networks Building

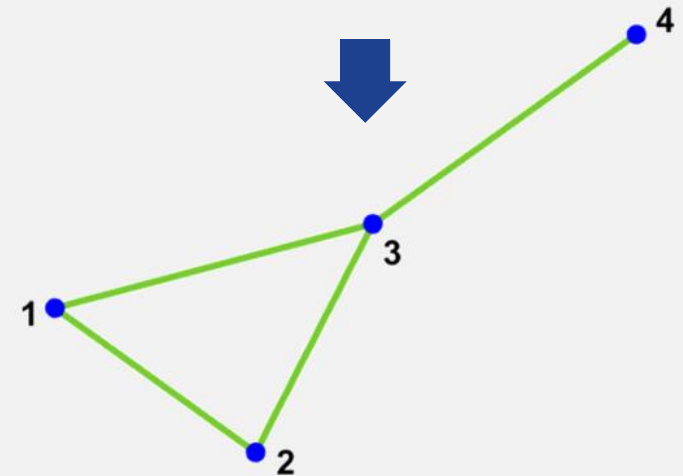
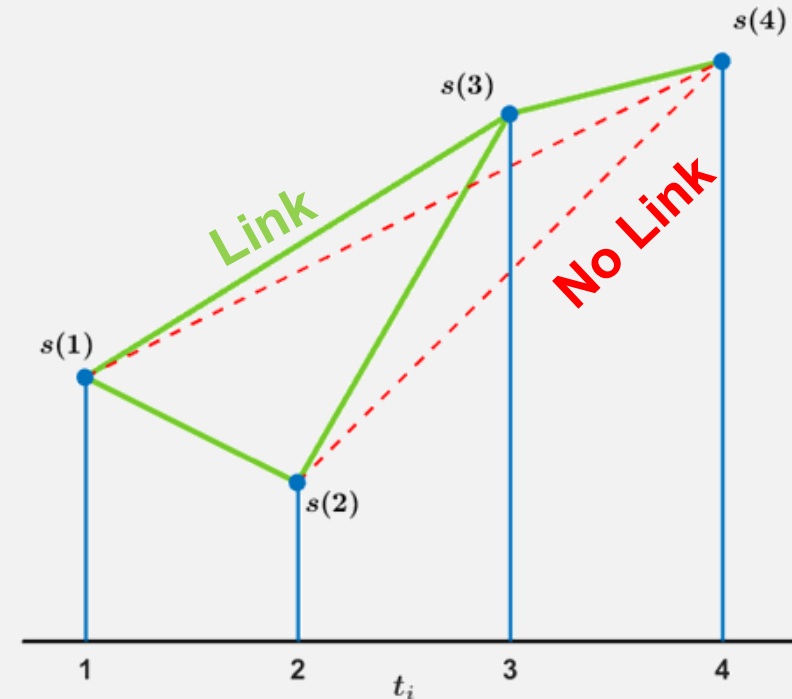
Each series value \equiv a node

How are nodes linked together?



Visibility Algorithm [2]

$$s(t_k) < s(t_j) + (s(t_i) - s(t_j)) \frac{t_j - t_k}{t_j - t_i}$$



Networks Building

Each series value \equiv a node

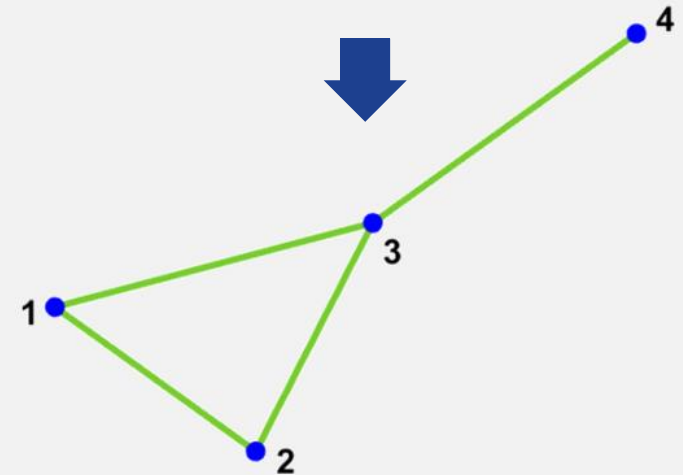
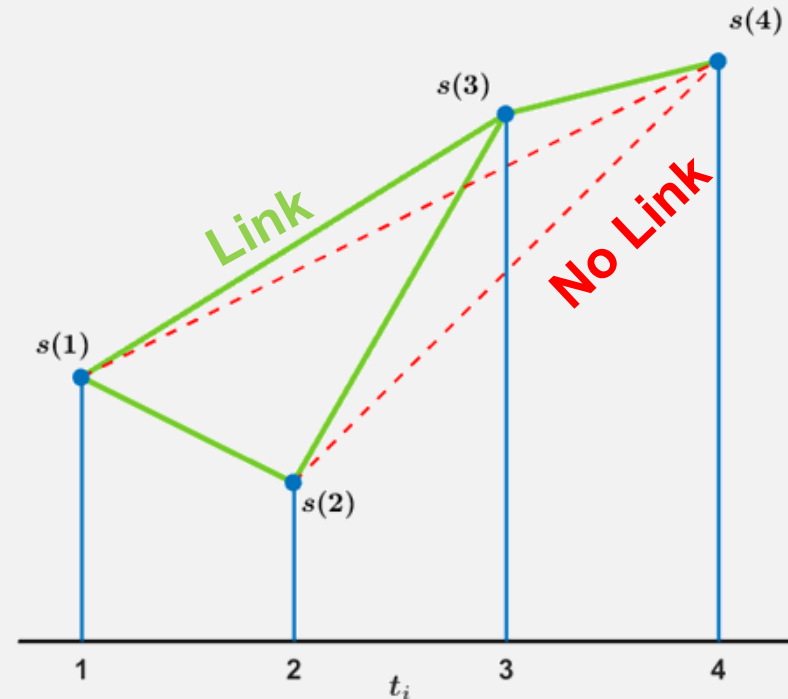
How are nodes linked together?



Visibility Algorithm [2]

$$s(t_k) < s(t_j) + (s(t_i) - s(t_j)) \frac{t_j - t_k}{t_j - t_i}$$

**Invariant under
affine transformations!**



Network Metrics

Global metrics: one value for each network

- ▶ Average Degree Centrality, k
- ▶ Transitivity, Tr
- ▶ Mean Link-Length, d_{1n}

Network Metrics

Global metrics: one value for each network

- ▶ **Average Degree Centrality, k**
 - Average visibility of nodes
- ▶ **Transitivity, Tr**
- ▶ **Mean Link-Length, d_{1n}**

Network Metrics

Global metrics: one value for each network

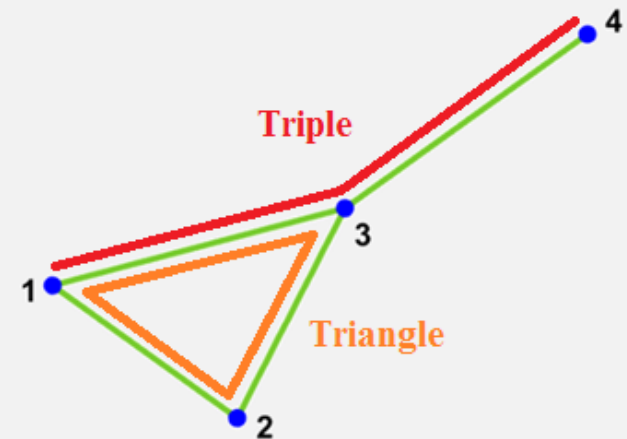
▶ **Average Degree Centrality, k**

- Average visibility of nodes

▶ **Transitivity, Tr** → $Tr = \frac{3N_{\Delta}}{N_V}$

- Inter-visibility among nodes

▶ **Mean Link-Length, d_{1n}**



Network Metrics

Global metrics: one value for each network

▶ **Average Degree Centrality, k**

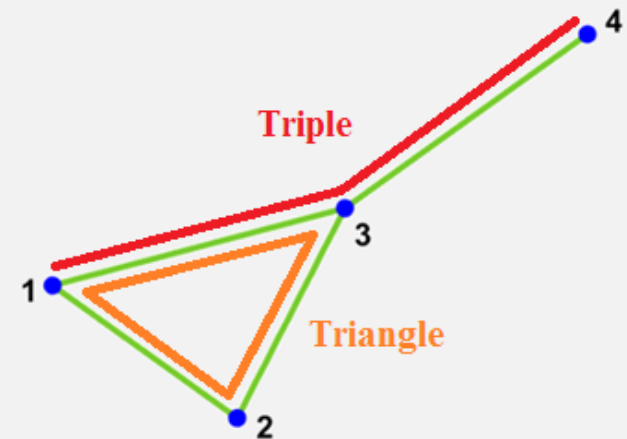
- Average visibility of nodes

▶ **Transitivity, Tr** → $Tr = \frac{3N_{\Delta}}{N_V}$

- Inter-visibility among nodes

▶ **Mean Link-Length, d_{1n}**

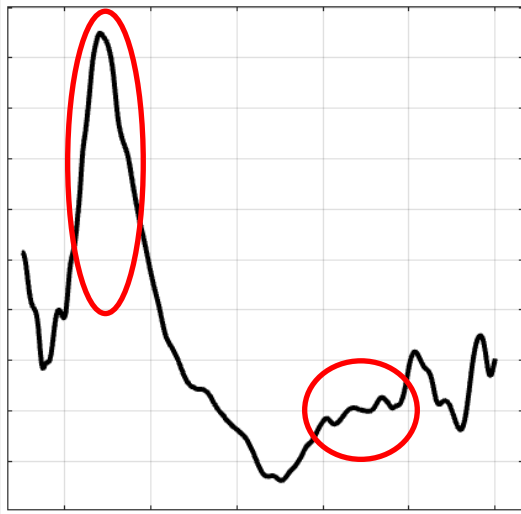
- Average temporal distance among nodes



Network Analysis

Focus on:

- ▶ *Peaks* (strong variations)
- ▶ *Irregularities* (small variations)



Network Analysis

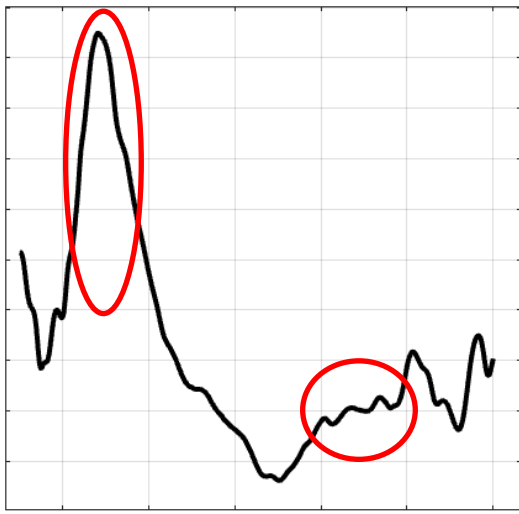
Focus on:

- ▶ *Peaks* (strong variations)
- ▶ *Irregularities* (small variations)

▶ d_{1n}

▶ Tr

} k

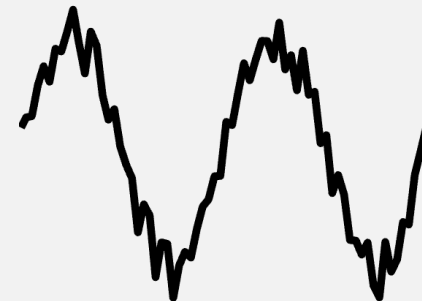
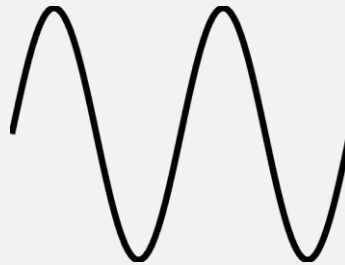
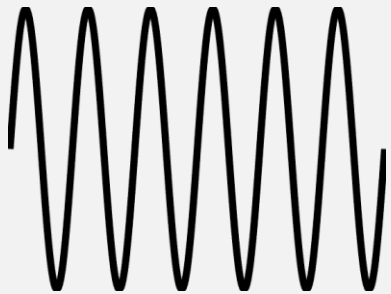


Network Analysis

Focus on:

- ▶ *Peaks* (strong variations) $\Rightarrow d_{1n}$
 - ▶ *Irregularities* (small variations) $\Rightarrow Tr$
- } k

Comparing the temporal structure of pairs of time-series



Network Analysis

Focus on:

- ▶ *Peaks* (strong variations) $\Rightarrow d_{1n}$
 - ▶ *Irregularities* (small variations) $\Rightarrow Tr$
- } k

Comparing the temporal structure of pairs of time-series



Network Analysis

Focus on:

- ▶ *Peaks* (strong variations) $\rightarrow d_{1n}$
 - ▶ *Irregularities* (small variations) $\rightarrow Tr$
- } k

Comparing the temporal structure of pairs of time-series

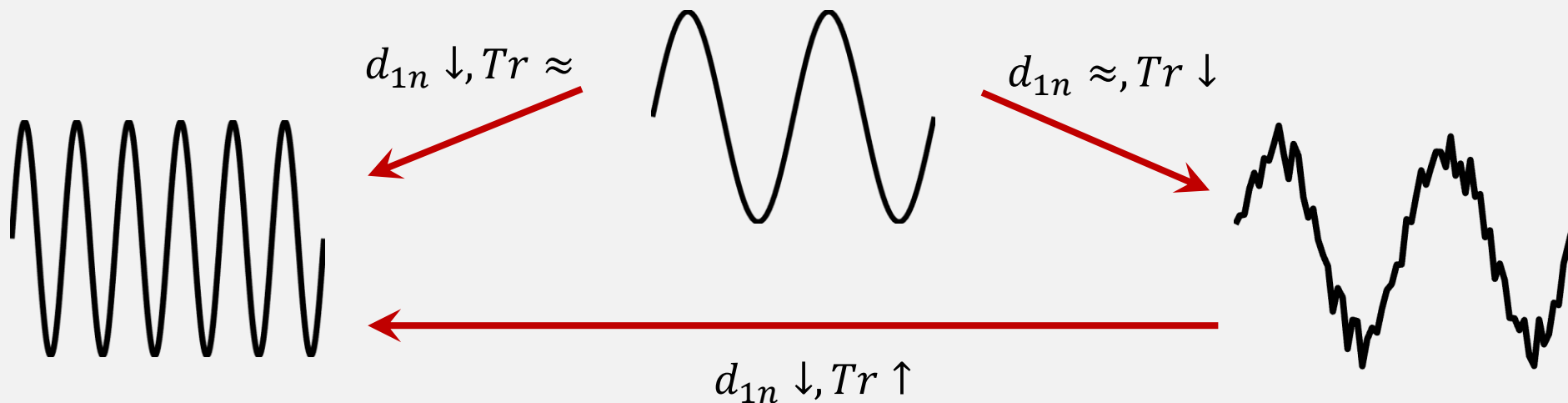


Network Analysis

Focus on:

- ▶ *Peaks* (strong variations) \Rightarrow d_{1n} $\left. \vphantom{d_{1n}} \right\} k$
- ▶ *Irregularities* (small variations) \Rightarrow Tr

Comparing the temporal structure of pairs of time-series



Data & Processing

John Hopkins Turbulence Database [1]

- ▶ **DNS:** $Re_\tau = 1000$
- ▶ **Grid Resolution:** $(2048 \times 512 \times 1536) \rightarrow (64 \times 70 \times 12)$
- ▶ **Time samples:** $N_T = 4000 \rightarrow 4000$ nodes

Data & Processing

John Hopkins Turbulence Database ^[1]

- ▶ **DNS:** $Re_\tau = 1000$
- ▶ **Grid Resolution:** $(2048 \times 512 \times 1536) \rightarrow (64 \times 70 \times 12)$
- ▶ **Time samples:** $N_T = 4000 \rightarrow 4000$ nodes

Processing:

- ▶ Normalized velocity time-series: $u^* = (u - \mu) / \sigma$
- ▶ A network for each grid point \rightarrow 3 values of metrics
- ▶ Metrics depend only on y^+ \rightarrow $(x - z)$ averages

Data & Processing

John Hopkins Turbulence Database [1]

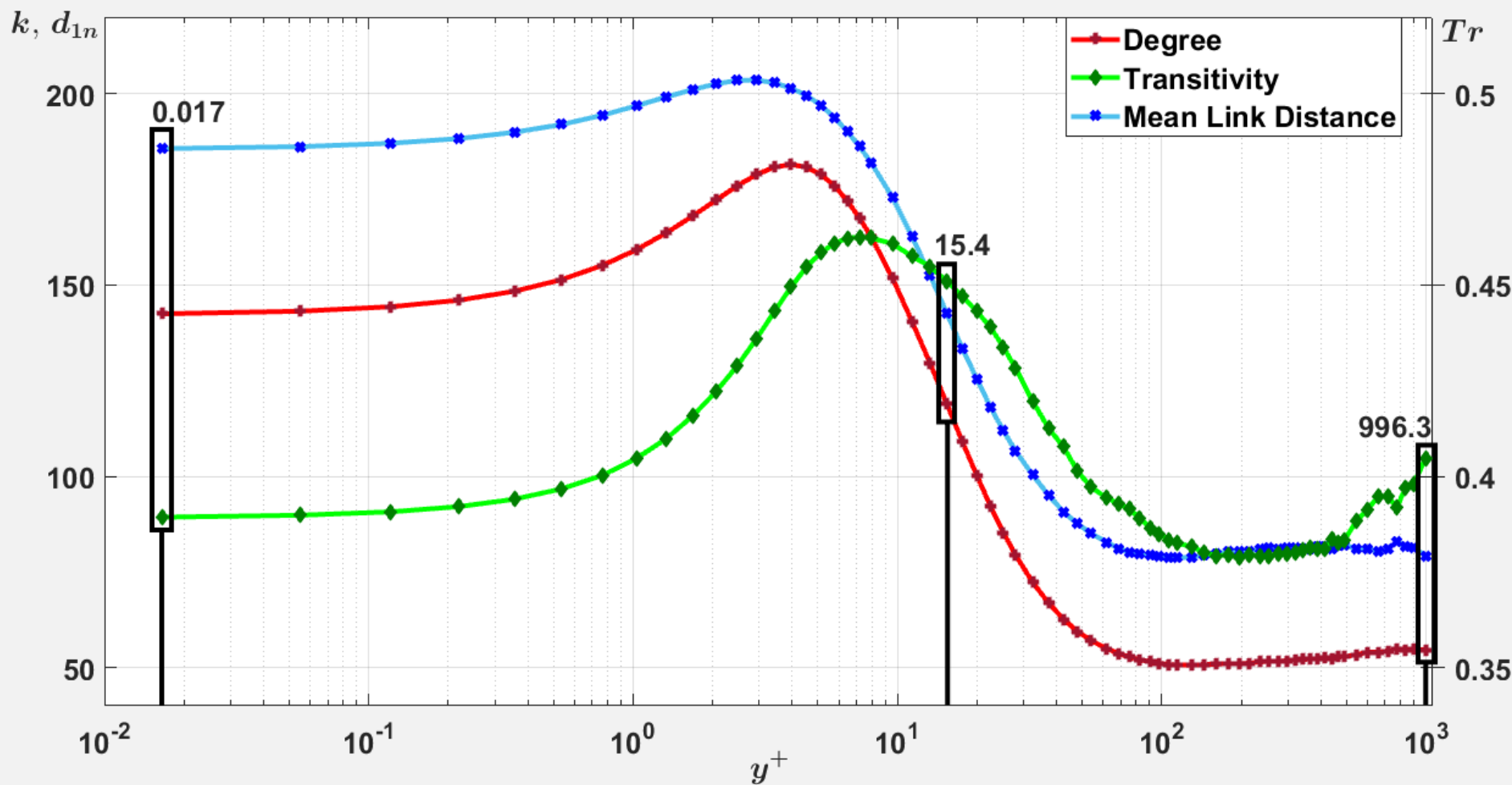
- ▶ **DNS:** $Re_\tau = 1000$
- ▶ **Grid Resolution:** $(2048 \times 512 \times 1536) \rightarrow (64 \times 70 \times 12)$
- ▶ **Time samples:** $N_T = 4000 \rightarrow 4000$ nodes

Processing:

- ▶ Normalized velocity time-series: $u^* = (u - \mu) / \sigma$
- ▶ A network for each grid point \rightarrow 3 values of metrics
- ▶ Metrics depend only on y^+ \rightarrow $(x - z)$ averages

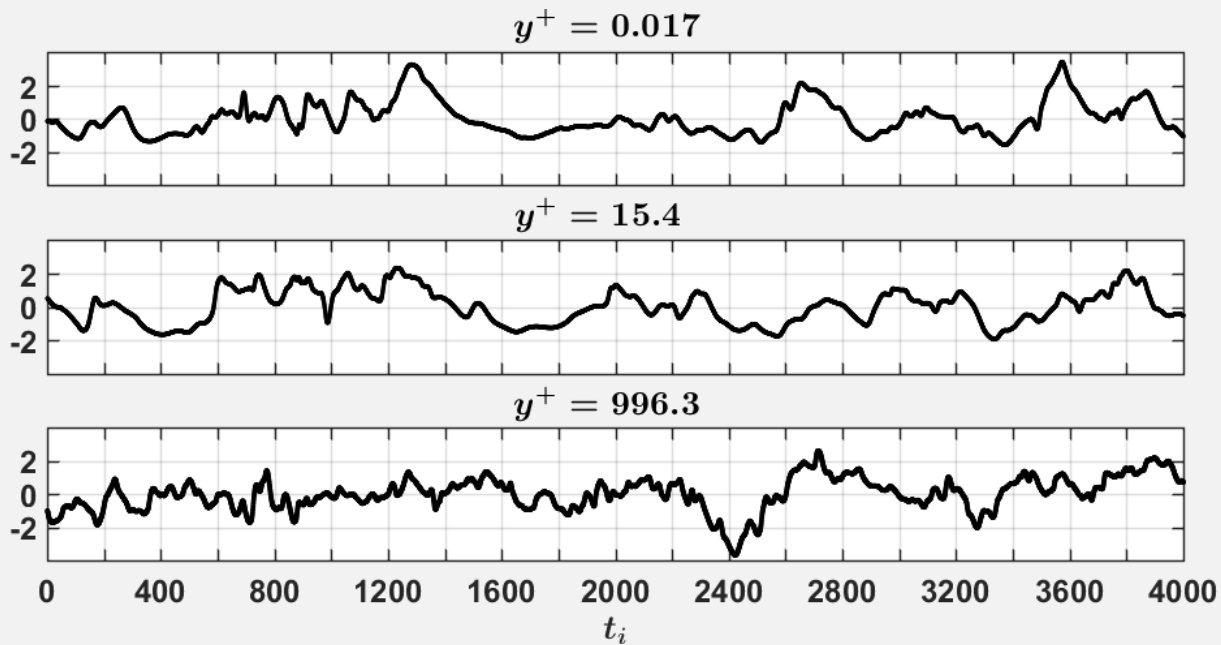
Results: u^*

Averaged Metrics



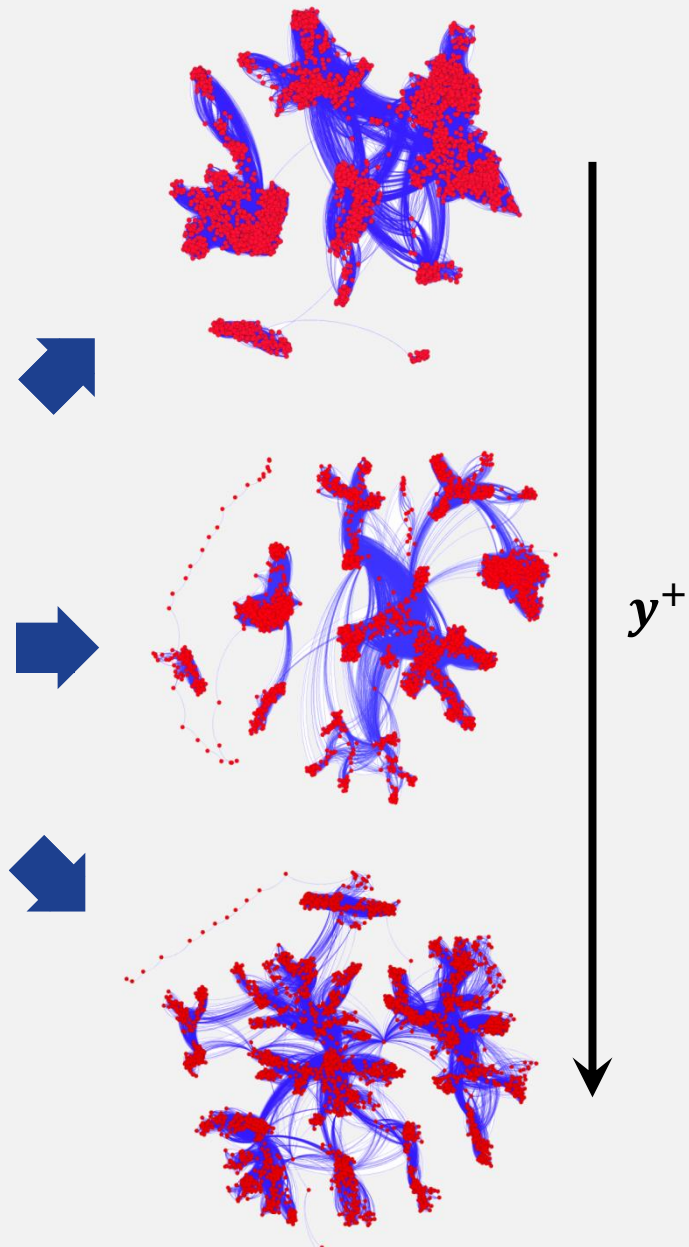
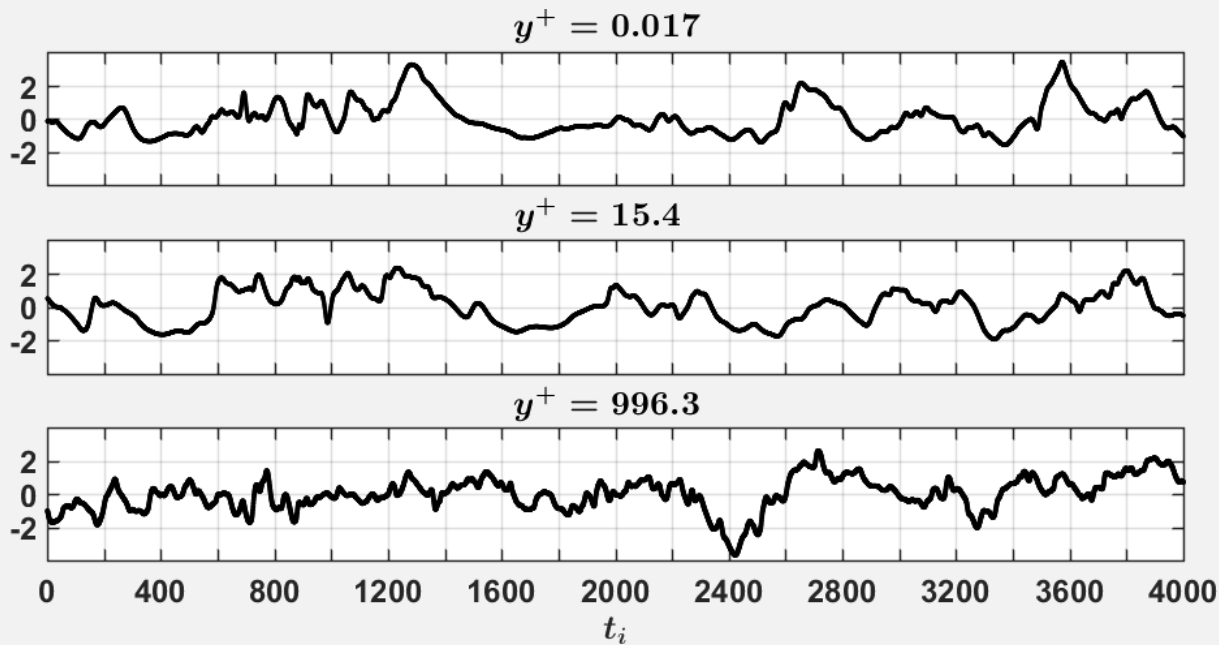
Results: u^*

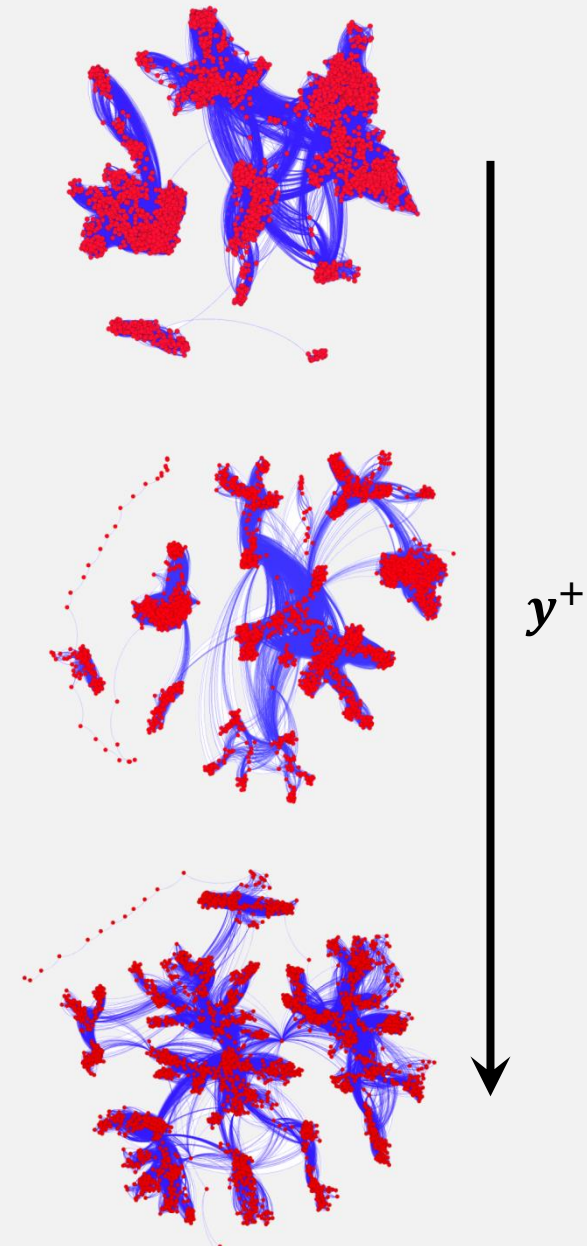
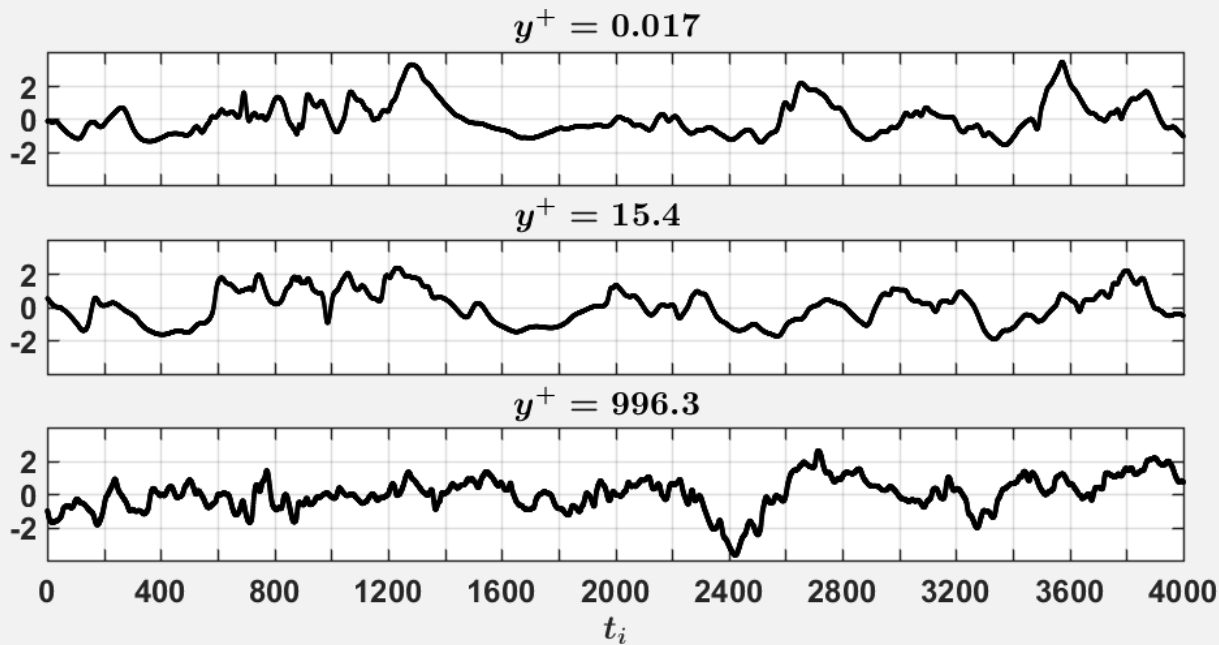
Normalized time-series – $u^*(t_i)$



Results: u^*

Normalized time-series – $u^*(t_i)$

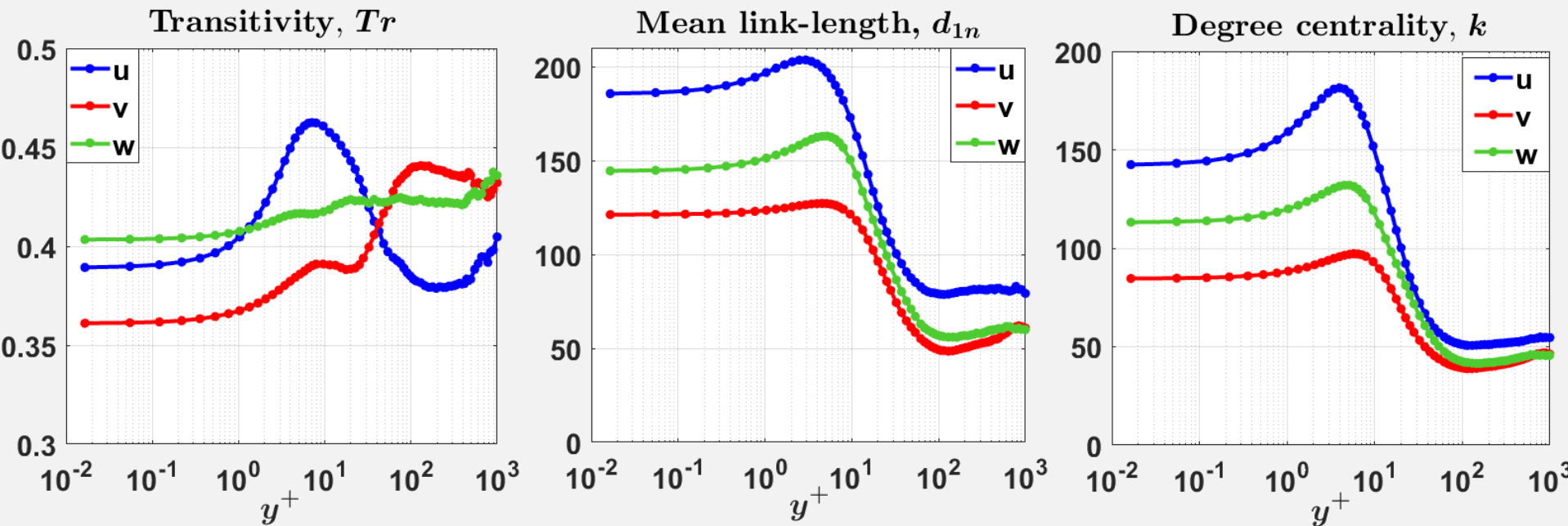


Results: u^* Normalized time-series – $u^*(t_i)$ 

Networks inherit the temporal structure of the series

Results

Velocity components: (u^*, v^*, w^*)



Variations of peaks and irregularities can be qualitatively inferred looking at the metrics behaviors along y^+

Conclusions

Meaningful Parameters:

- ▶ Transitivity \rightarrow irregularities
- ▶ Mean link-length \rightarrow peaks

Main Results:

- ▶ Qualitatively infer the temporal structure of the series
- ▶ Complex networks inherit the main flow dynamics

Future works:

- ▶ Deeper analysis / Different Reynolds
- ▶ Application to other turbulent flows

Thank You for Your Attention
Questions?