



Detecting natural and anthropic effects on displacements and water level changes: a combined observation from rain gauges, piezometers and CGNSS

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The Po plain (Northern Italy) has largely subsided due to natural processes and human activities.



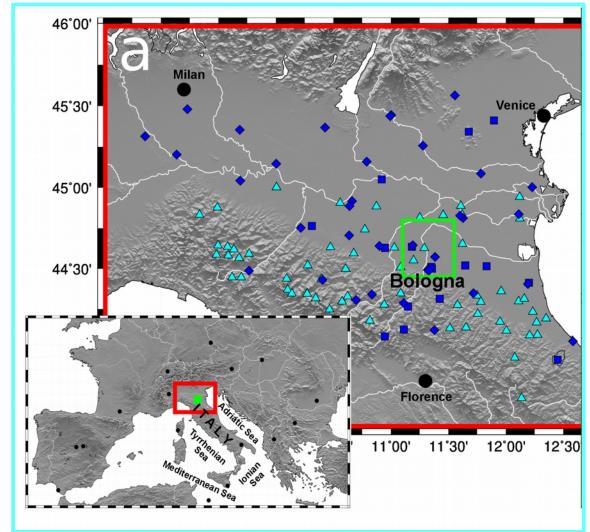
Such an area is characterized by an excellent monitoring activity

with a good **spatial** and **temporal** distribution of data



Continuous GNSS sites and rain gauges are distributed in the whole Po plain

In particular we focus on the **Bologna** metropolitan area.

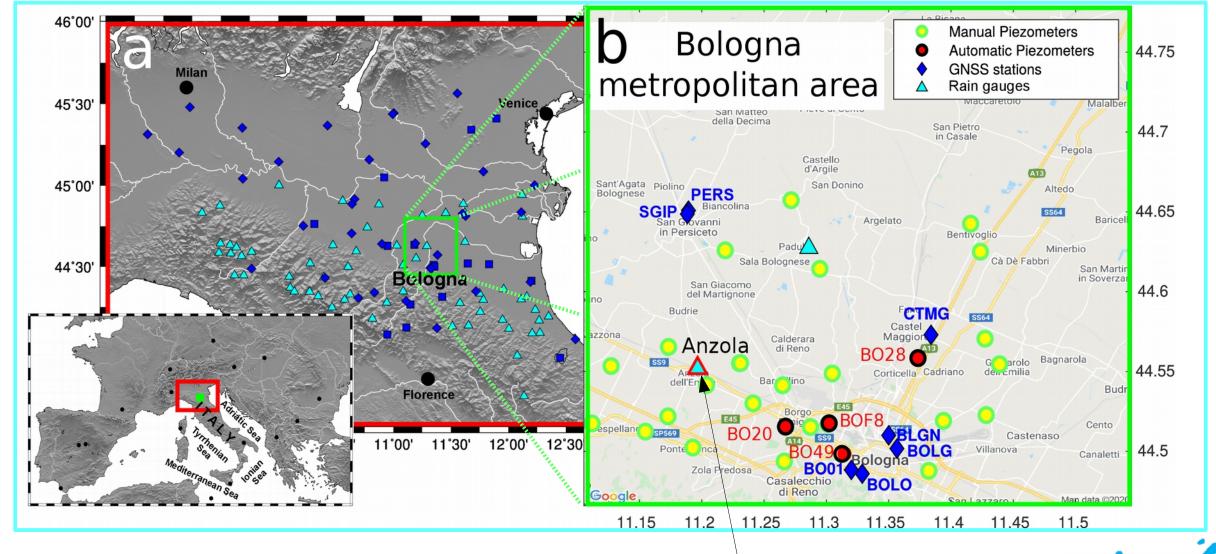






Rain gauges

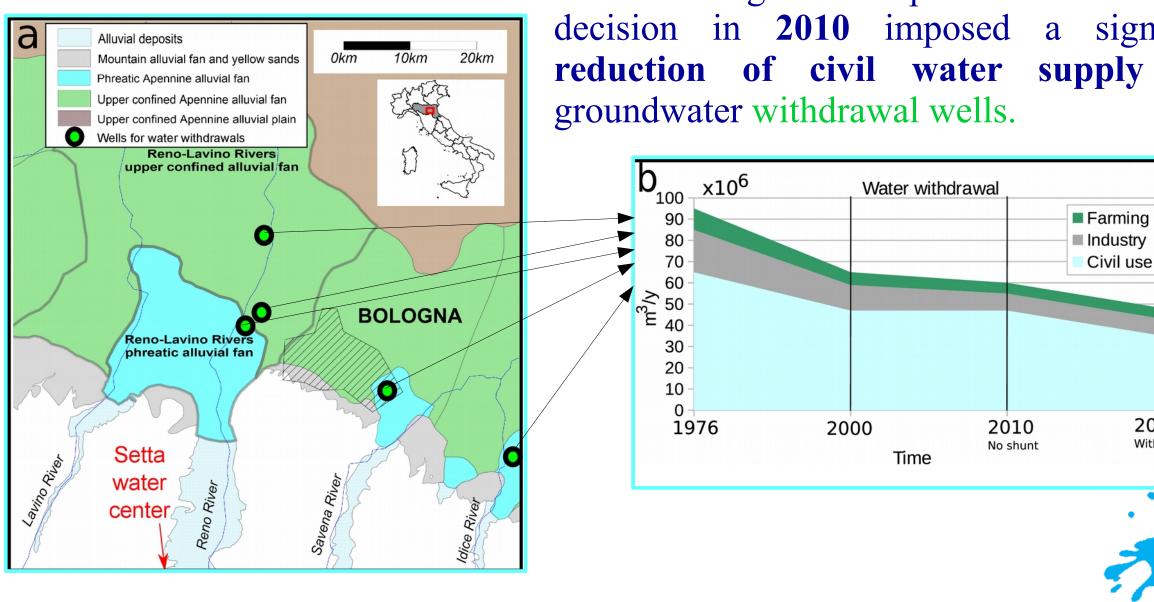




Several manual and automatic piezometers are distributed near Bologna

The closest rain gauge (Anzola)



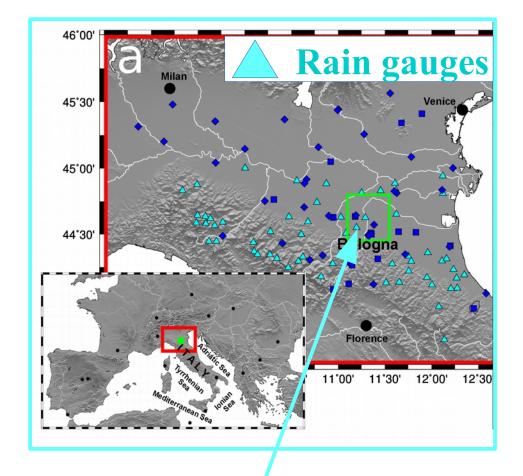


In the Bologna metropolitan area a politic decision in 2010 imposed a significant reduction of civil water supply

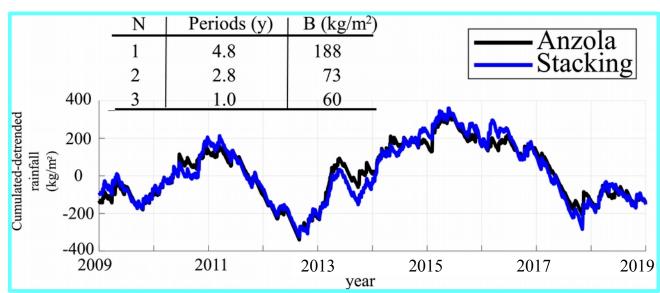
2010

With shunt

Analyzing the rainfall data



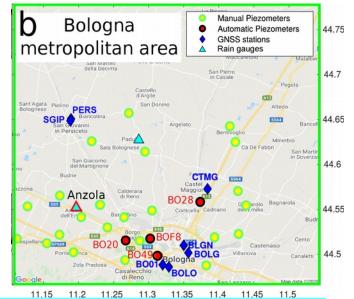
- Cumulated rainfall
- Removal of linear trend
- Stacking of residual time series
- Find 3 meaningful seasonal signals

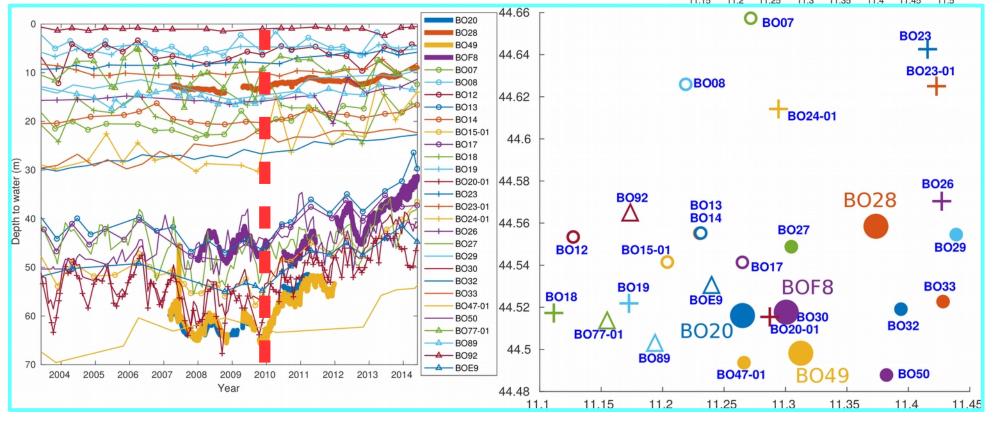


Anzola rain gauge

Depth to water time series

• Resampling of raw piezometric data (step of 3 month)

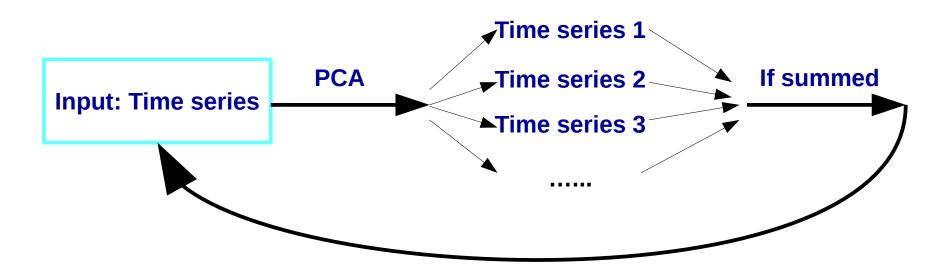






Depth to water time series

- The resampled time series of the piezometers were analyzed with a **Principal Component Analysis** (PCA)
- Spatiotemporal separation of the data into a set of linearly uncorrelated principal components
- Each component is as a signal due to a source generating the observed water level variations.





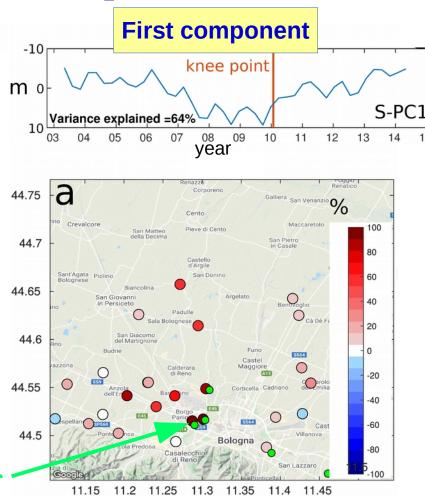
Depth to water time series

• According to the PCA, the first principal component (PC1) has the largest possible variance, so it explains much of the observation, and the subsequent components are ordered with decreasing variance values.





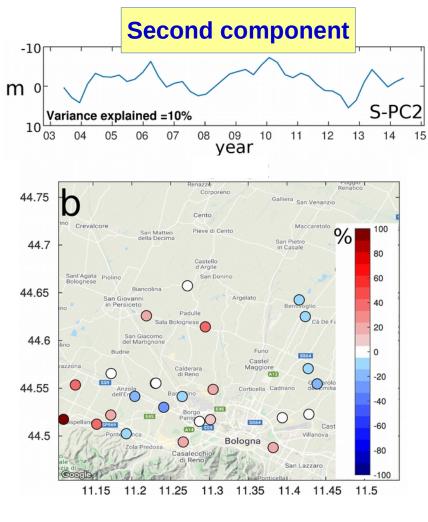
PCA on depth to water time series



Water wells

Reduction of fluid withdrawal (2010)

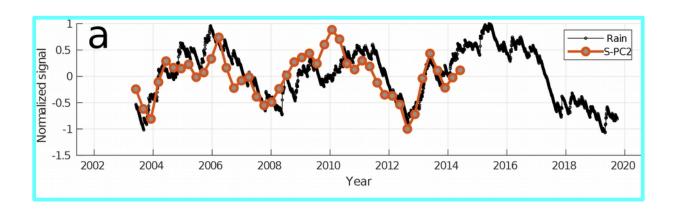
11.4 11.45



Rainfall



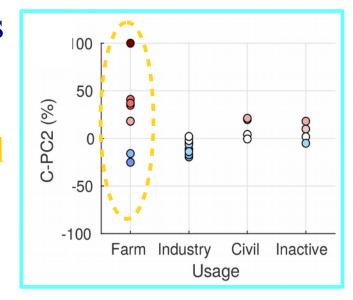
PCA on depth to water time series

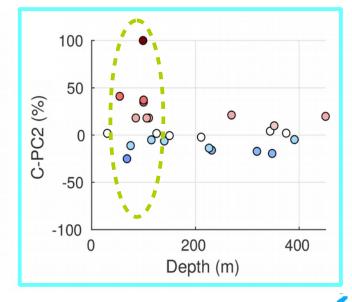


The second PC (PC2) correlates with rainfall

The magnitude of **PC2** is greater in piezometers:

- located in agricultural or farming zones
- with a small depth







Piezometers

PCA on GNSS time series

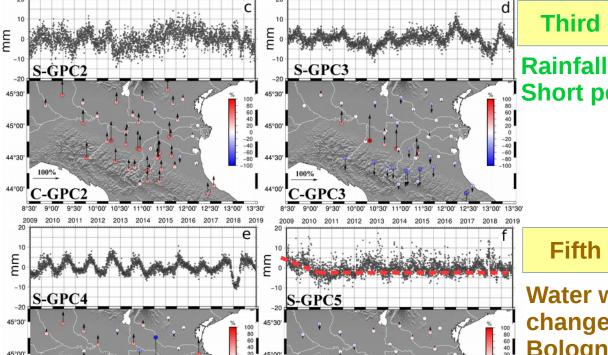
First component S-GPC1 Kinematic pattern

Second component

Rainfall **Long periods**

Fourth component

Periodic, local, antrhopic water withdrawal



Third component

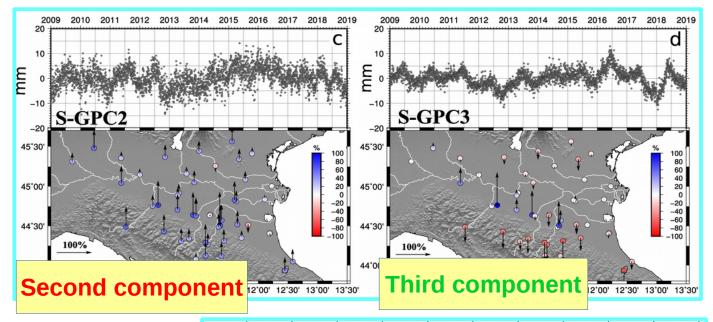
Short periods

Fifth component

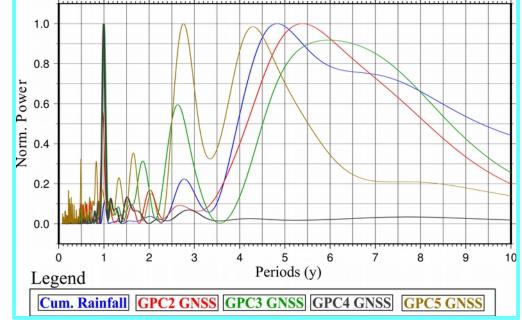
Water withdrawal change in **Bologna (2010)**



The second and third components: the effects of rainfall

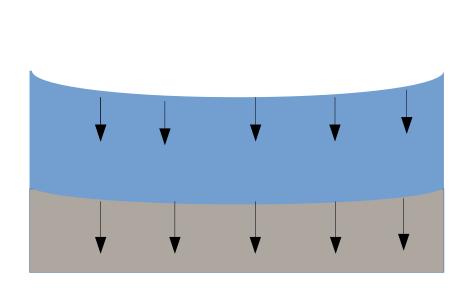


There is an agreement between the spectra of residual cumulative rainfall, S-GPC2 and S-GPC3, in terms of peak periods

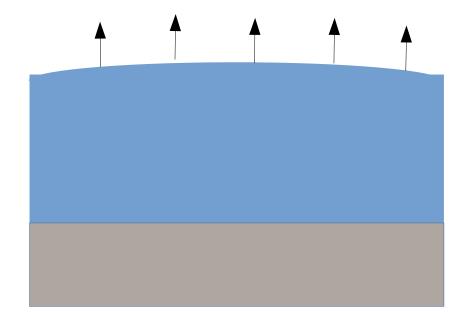




Elastic or poro-elastic response to the rainfall?



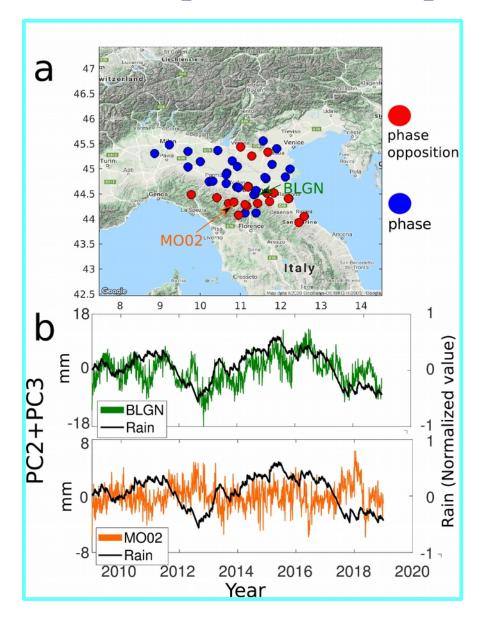
Dominant Elastic loading
Subsidence



Dominant Poro-elastic expansion

Uplift

Elastic or poro-elastic response to the rainfall?



In each station we determine if the GPC2 + GPC3 response is

in phase (Poro-Elastic response)

or

in opposition of phase (Elastic response)

with rainfall time series



Discussion

The limitations due to the **different sampling rate** (1 day for GNSS and about 3 months for piezometric data) and **discontinuous observation time span** in the available datasets as well as the **spatial heterogeneity** of networks, represent a problem/challenge.

We showed that the **PCA results** represents a possible solution.



Discussion

While the analysis of piezometric data shows a clear increase in the water level following the withdrawal decrease, the **anthropic induced surface displacements** are significantly smaller (about few mm) than the ones locally induced by rainfall (about 10 mm).

Accordingly, without a multivariate analysis such effect on vertical displacements would have remained hidden in the raw time series.

It is worth to notice that after 2010 the subsidence rate in the Bologna city area is decreased of about 4 mm/y, this represents an important contribution to reducing the long-term subsidence effects.

Conclusions

To correctly model the rainfall contribution to the vertical movement of the surface, it is necessary to consider different prevailing effects: **elastic** in the case of mountains and **poro-elastic** in the case of the sedimentary plain.

The combined observation of **GNSS** and **piezometers**, **rainfall**, **geological setting** and the use of **PCA**, are very important in order to get a complete understanding of the anthropic and natural signals in cases of aquifer over-exploitation.

The study of different observables can provide precious hints to assess the **best** practices for the governance of groundwater resources in a climate change scenario.

Thanks for your attention!

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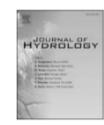
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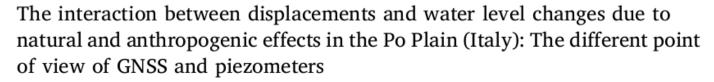
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Research papers



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