

Estimating Transient Water Storage from Hurricane Harvey using GPS observations of Vertical & Horizontal Land Motion

Chris Milliner, JPL – NPP Postdoc

Materna, K., UC Berkeley

Burgmann, R., UC Berkeley

Fu, Y., Bowling Green

Bekaert, D., JPL

Moore, A., JPL

Adhikari, S., JPL

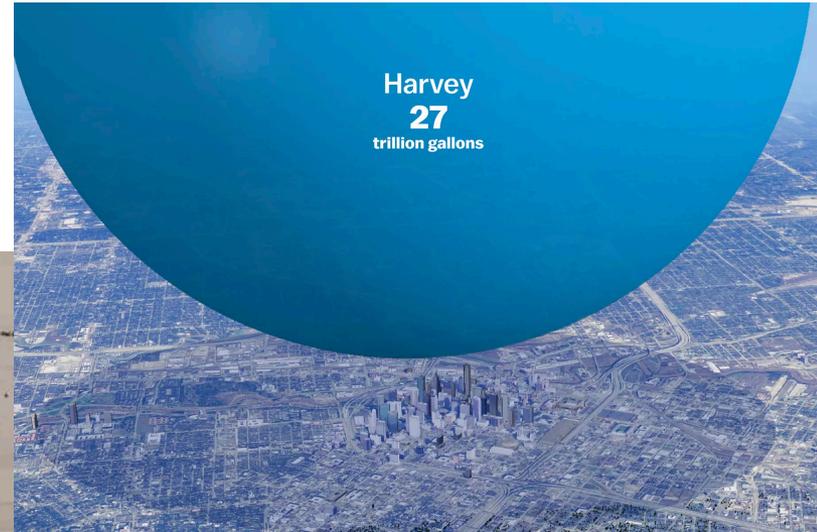
Argus, D., JPL

Introduction

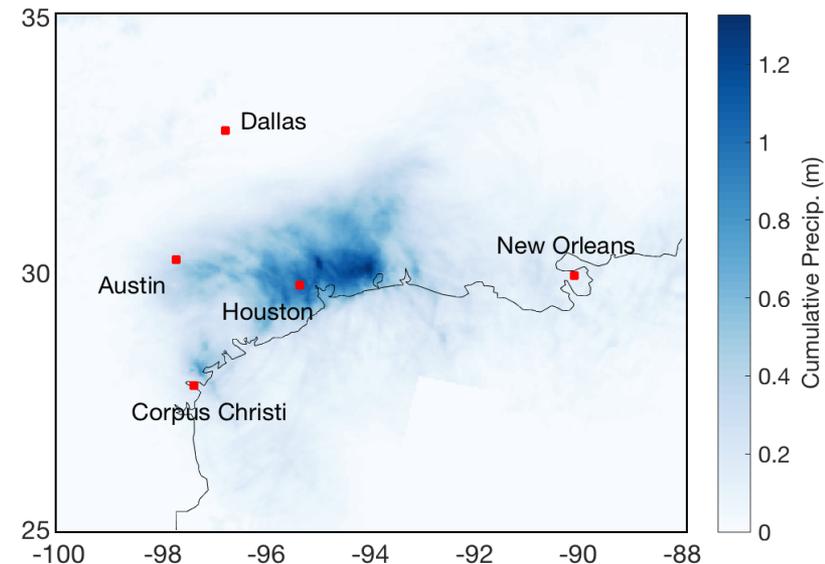
- **Question:** How does water accumulate and dissipate following a major hurricane? Can we measure this using GPS data?
- **Method:** Use cGPS data to measure Earth's deformation from water mass, this can be used to track the evolution of TWS
 - TWS = standing surface water, ground water + absorbed in soil.
- **Motivation:** Quantifying TWS important for:
 - **Understanding:** ability of drainage systems to respond and retain extreme influxes of water.
 - **Applications:** Stored water poses a secondary and **continued flood hazard**, once released into nearby streams. Observations of water storage could potentially improve **operational flood forecasting** used by flood managers.

Background

- Cat 4 event – hit US mainland August 26th, lasted 7 days
- Stalled in southern Texas, → retreated → Louisiana → Ms, Tn
- Wettest recorded US hurricane
 - Total rainfall: $\sim 102 \text{ km}^3$
 - $\sim 1.54 \text{ m}$ of cumulative rain recorded east of Houston.

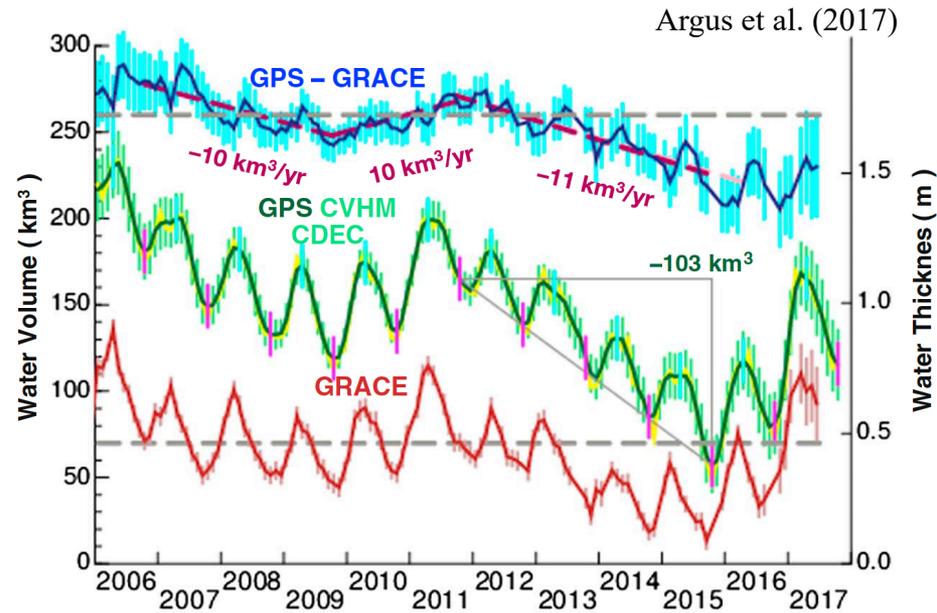


<https://www.vox.com/science-and-health/2017/8/28/16217626/harvey-houston-flood-water-visualized>

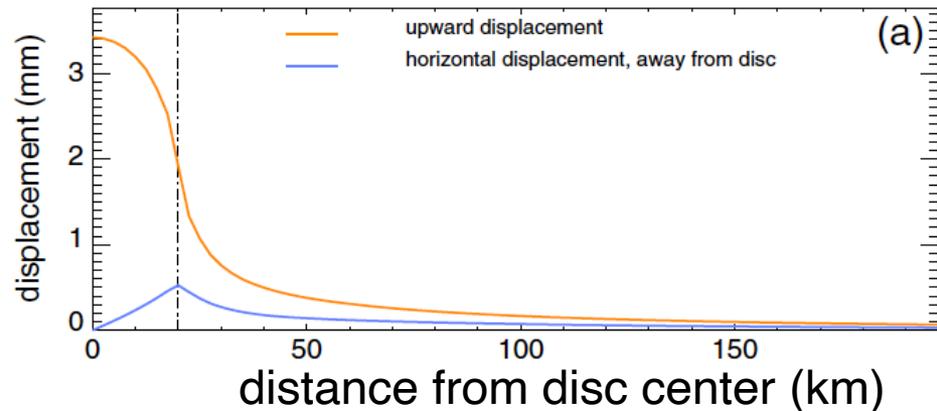


Outline

- Hydrologic loading primarily causes vertical surface motion
- **Challenge:** Noise level of vertical GPS is relatively high (~ 3 mm).
 - Usually we average over large areas (regional-continental)
 - Long timescales (months-seasons) to characterize loading.
- Is the stability of GPS positioning sufficient to resolve Harvey's transient loading signature?



Earth's elastic response to unloading water disk
20 km diameter and 1 m thickness



GPS Data

- Total of 219 GPS stations
 - Daily sampling
 - East, north and vertical
- 72 day period: Aug-mid-October
- Data processed using GIPSY-OASIS II, in PPP
 - Ref frame IGS08
 - VMF1 troposphere mapping function
(Boehm et al., 2006)
 - Remove ocean tide loading + solid Earth body tides
- Stations managed by assortment of agencies
 - Tx DOT
 - UT Austin
 - Univ. Houston
 - SmartNet
- Average spacing 20-30 km

Non-tidal atmosphere + ocean loading (IERS/GFZ)

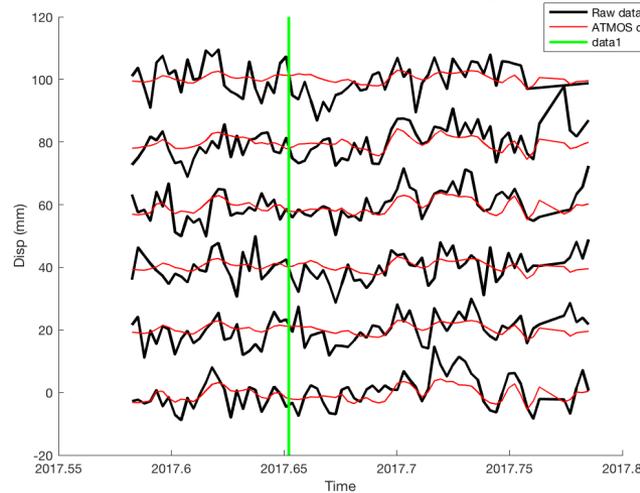
Black = data

Red = correction

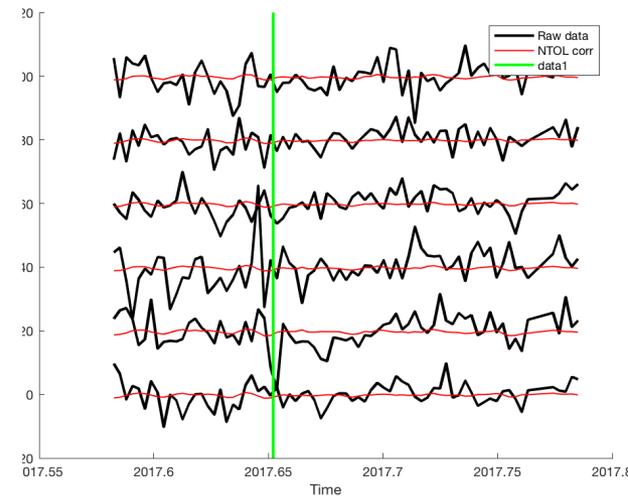
Green = landfall of Harvey

- Corrections for non-tidal atmospheric loading + ocean loading
- Effect of atmospheric pressure changes unloading-loading surface.
- ATMOS: RMS average reduction = up to ~20%
- NTOL RMS reduction up to 10% near-shore

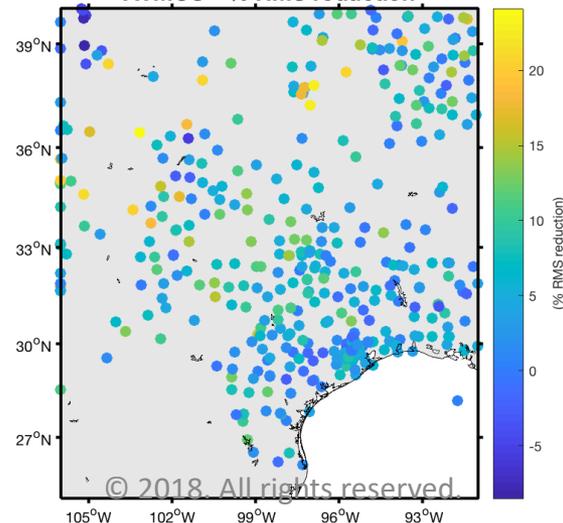
Atmospheric loading (red)



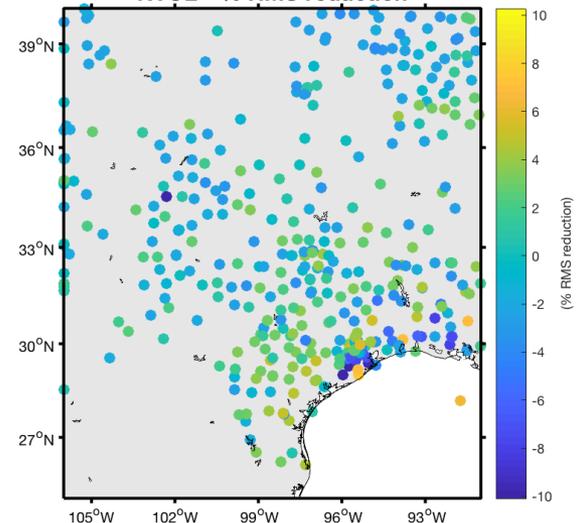
NTOL correction (red)



ATMOS - % RMS reduction

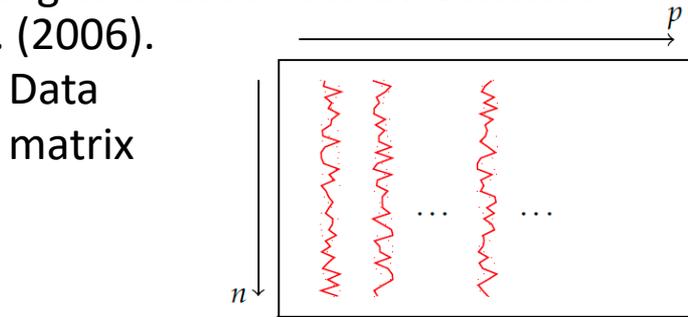


NTOL - % RMS reduction

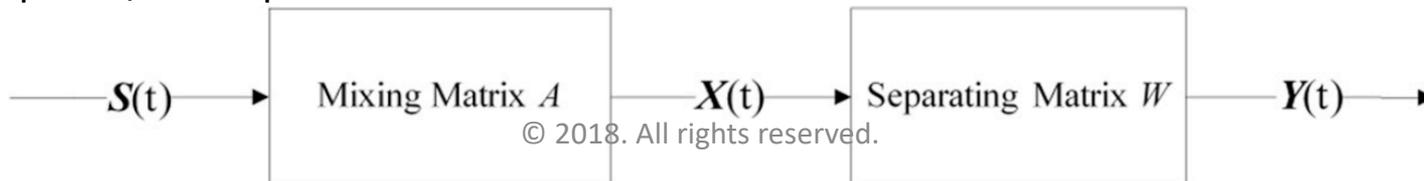
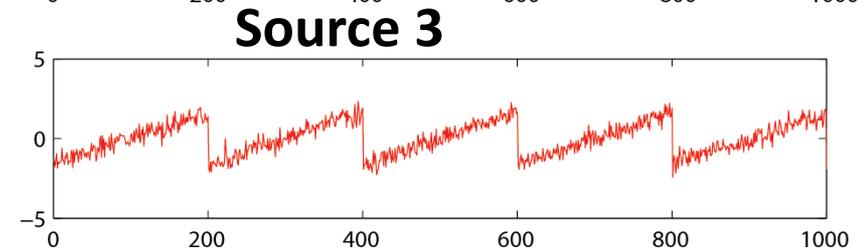
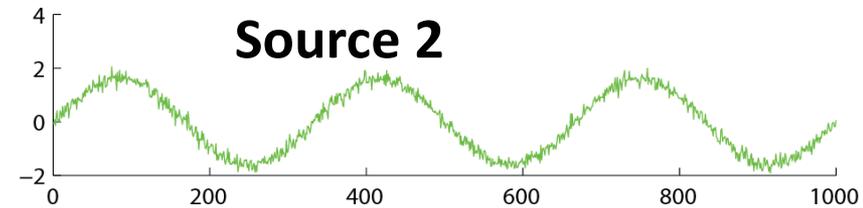
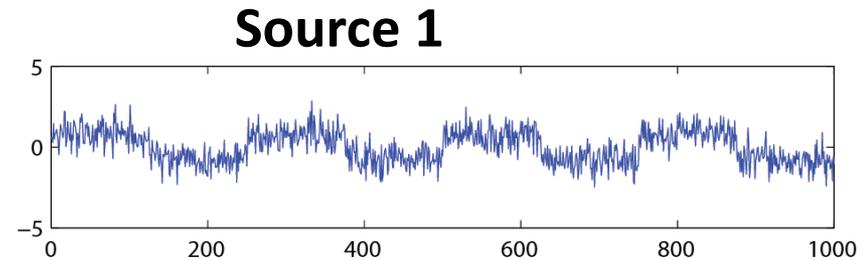
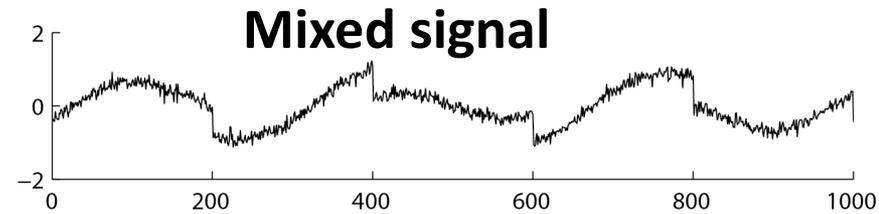


PCA – Principal Component Analysis

- GPS vertical most noisy \rightarrow PCA to extract hydro signal
- PCA – statistical approach to compresses data by identifying directions of max. variance - Dong et al. (2006).

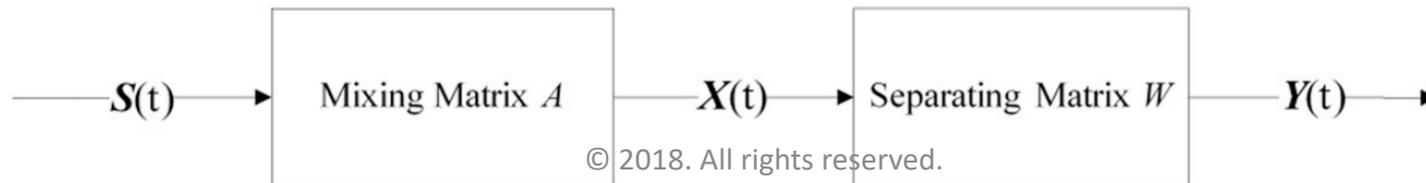
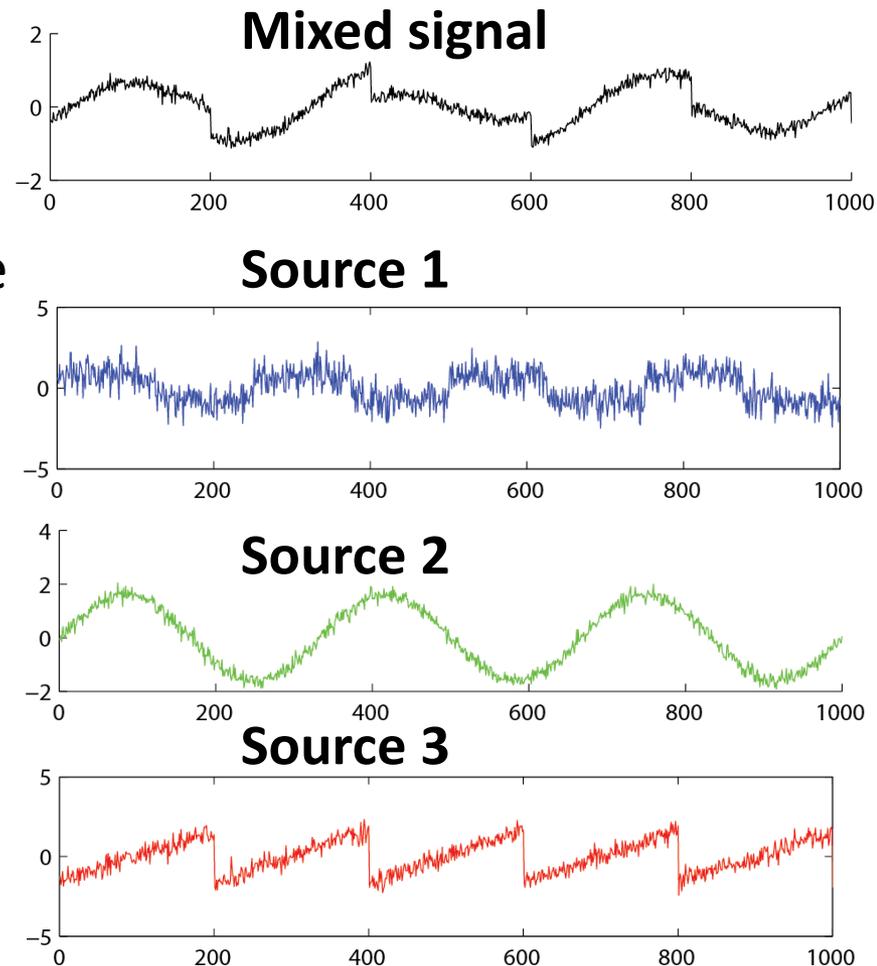


- Data = linear combination of principal components with spatial weight
- **CME** = artificial network wide motion due to inaccuracies in:
 - satellite orbits
 - ref. frame errors
 - Tropospheric/ atmospheric models.



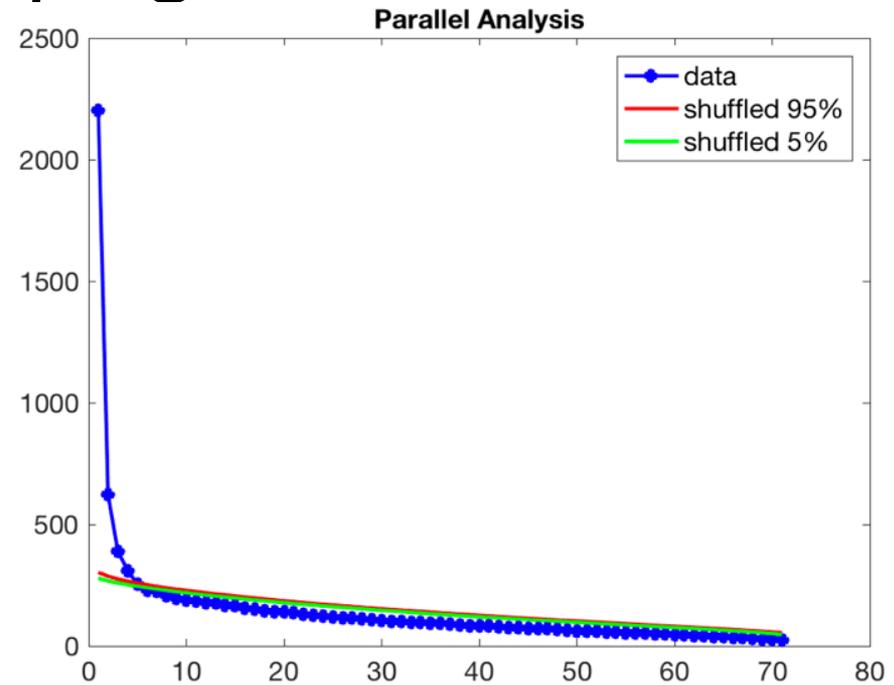
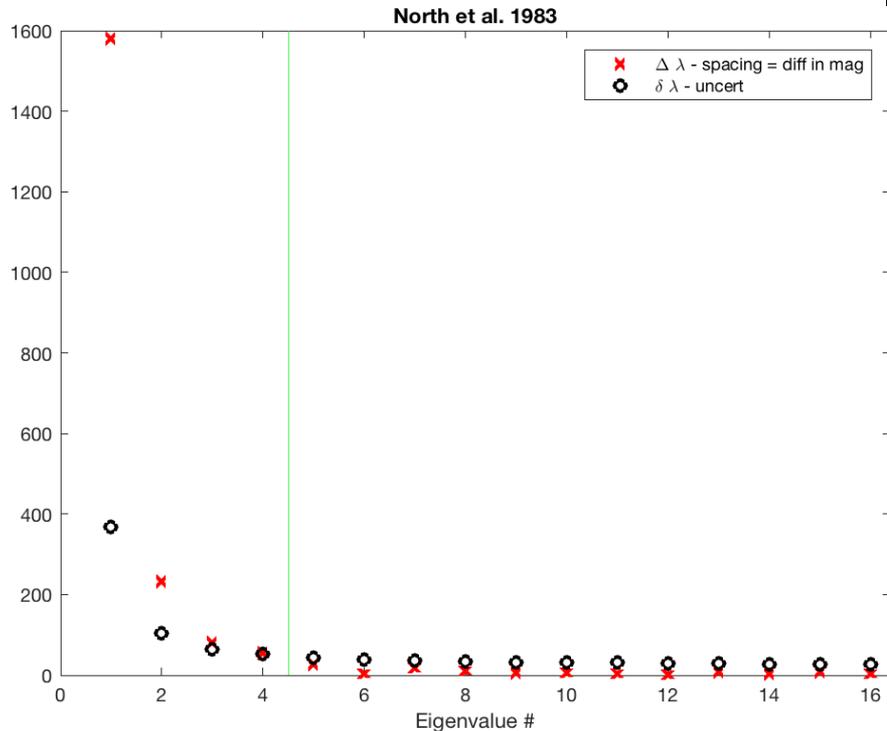
ICA - Independent Component Analysis

- ICA – identifies components that are statistically independent
- **Advantages:** ICA uses independence as a constraint to separate source, while PCA uses variance/correlation.
 - ICA suited for non-Gaussian distributions
- Use reconstruction ICA algorithm, Hyvärinen & Oja (2000).



© 2018. All rights reserved.

How many components to decompose data? - Stopping rules



“North’s rule of thumb”: Measure of separability

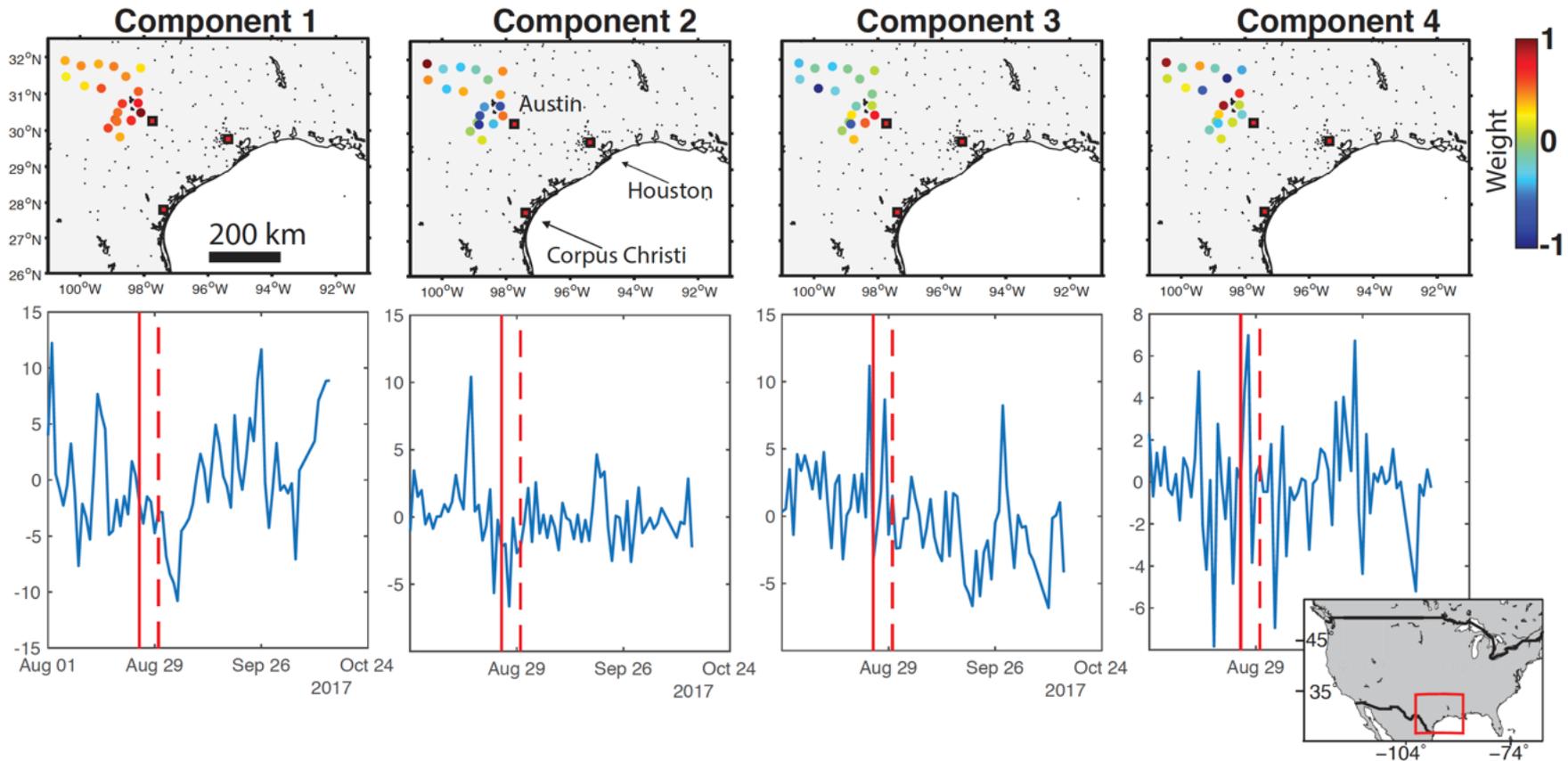
Idea: Assess which eigenvalues exceed that expected from a random process:

1. If uncert. exceeds separation, then component is deemed difficult to separate from its neighbor and from noise.

Horn’s Parallel Analysis

- Randomly scramble the data \rightarrow suite of random samples and eigenspectra with 95% CI.
- If eigenvalue $>$ 95% of eigenvalues from random data then component is retained.

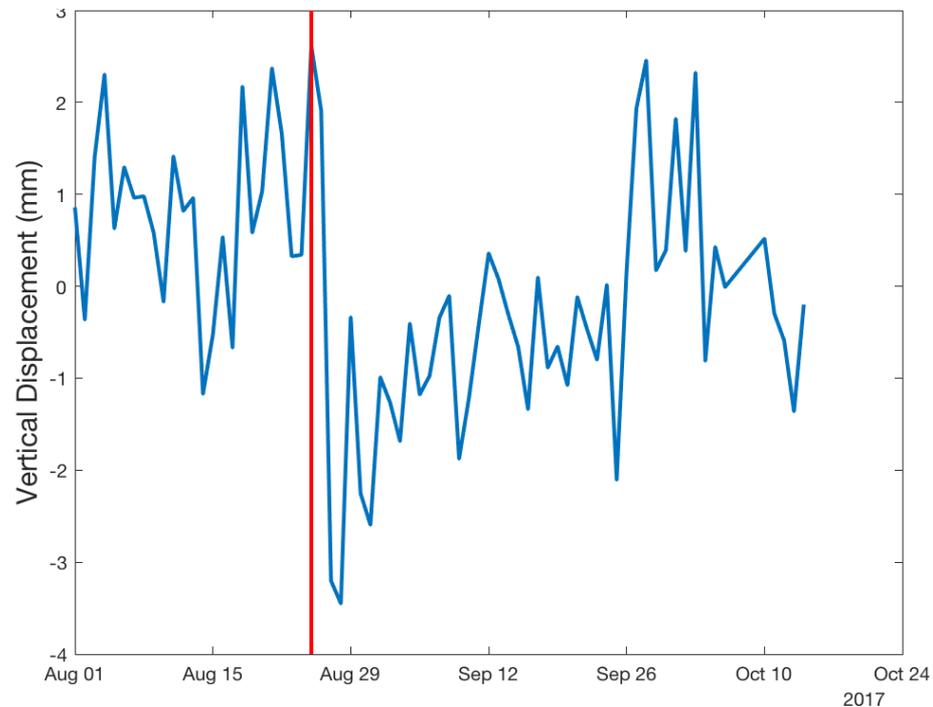
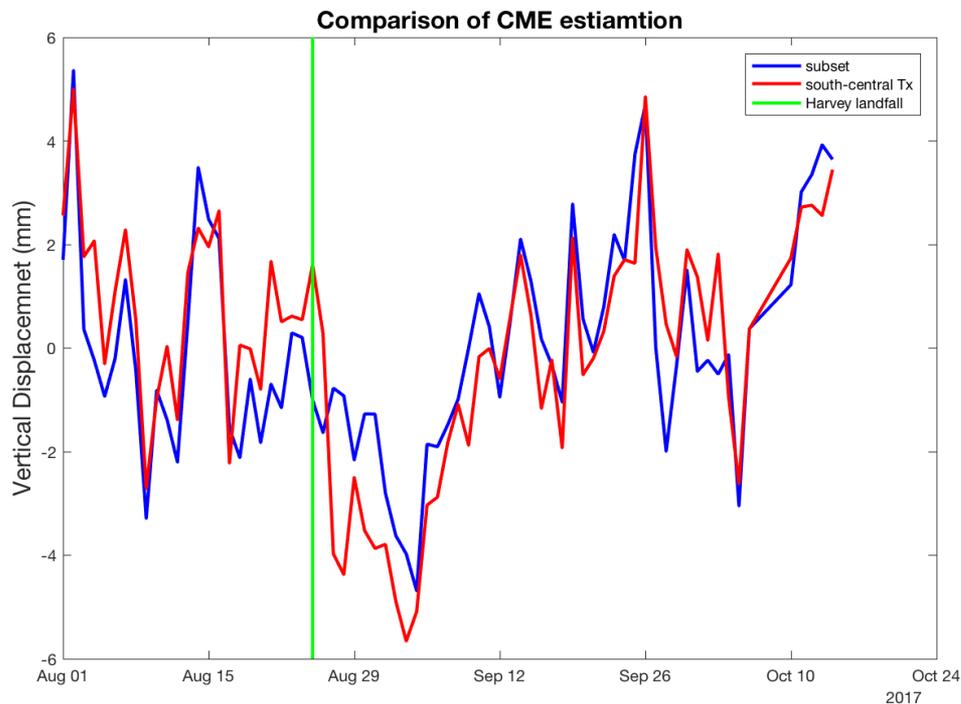
ICA filtering



- Components ordered in amount of motion explained.
- CME shows ~10 mm of subsidence, second landfall not detected.
- Hydrologic signal mixed onto first component

- Instead we estimate 'CME' from a subset of stations, distal from known precipitation
- Assume this CME is uniform across network

Comparing CME estimates

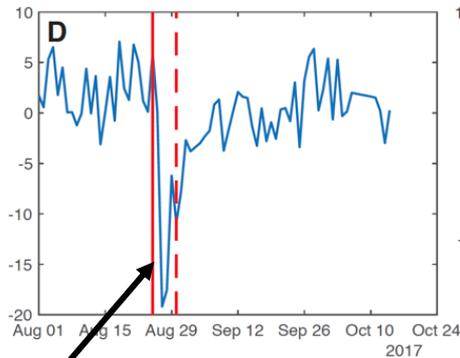
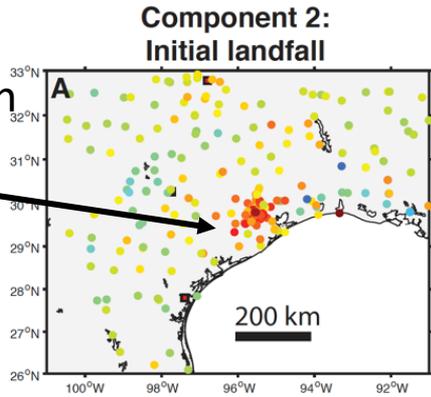


Red = CME from entire network
Blue = CME from subset of stations
Green = Landfall

- Above: Difference between two CME estimates
- Marked subsidence coincident with Harvey landfall (red line)
- Followed by gradual uplift
- Suggests hydrologic signal is mixed with CME

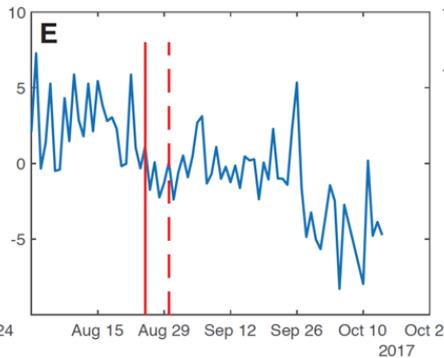
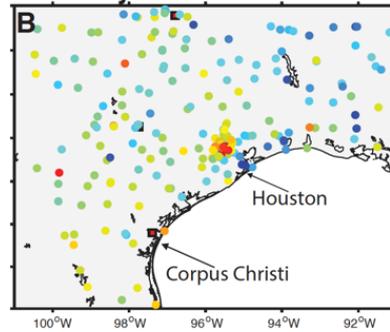
CME removed

Hydrologic signal:
Area of known precipitation



Hydrologic signal:
Coincident with initial landfall
Marked subsidence, gradual uplift

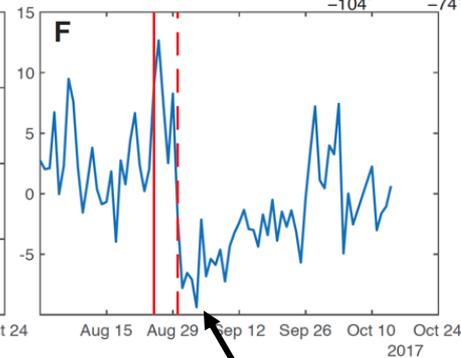
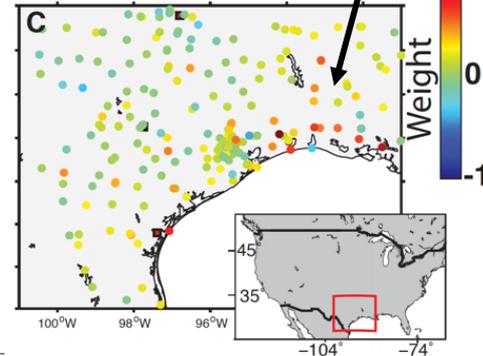
Component 3: Systematic error



3rd component, a linear trend
Groundwater extraction

Hydrologic signal:
Area of second landfall

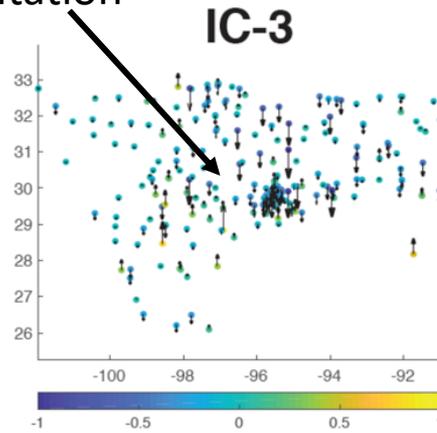
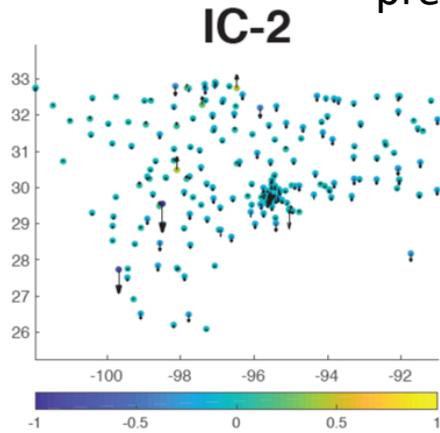
Component 4: Second landfall



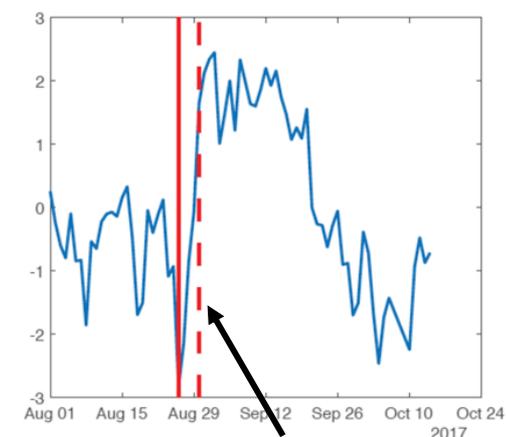
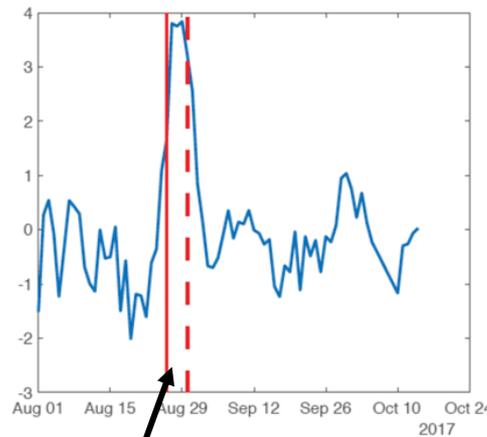
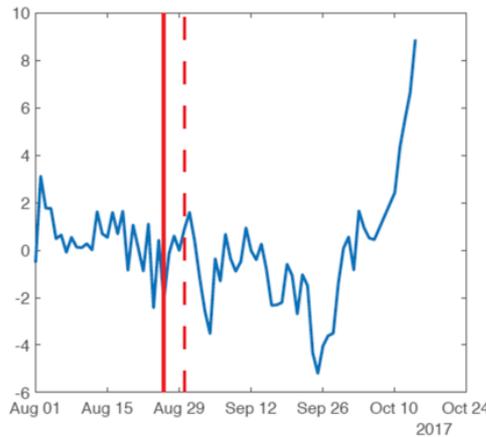
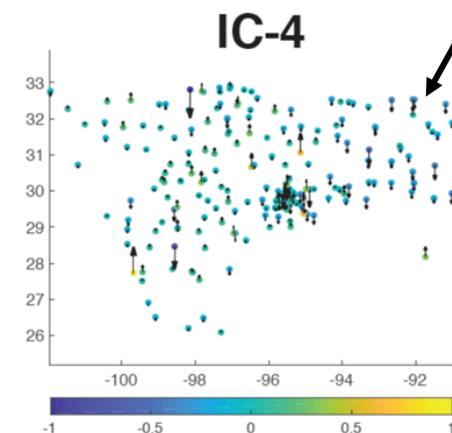
Hydrologic signal:
Coincident with second landfall
Marked subsidence, gradual uplift

East component (after CME removed)

Hydrologic signal:
Area of known
precipitation



Hydrologic signal:
Area of second landfall

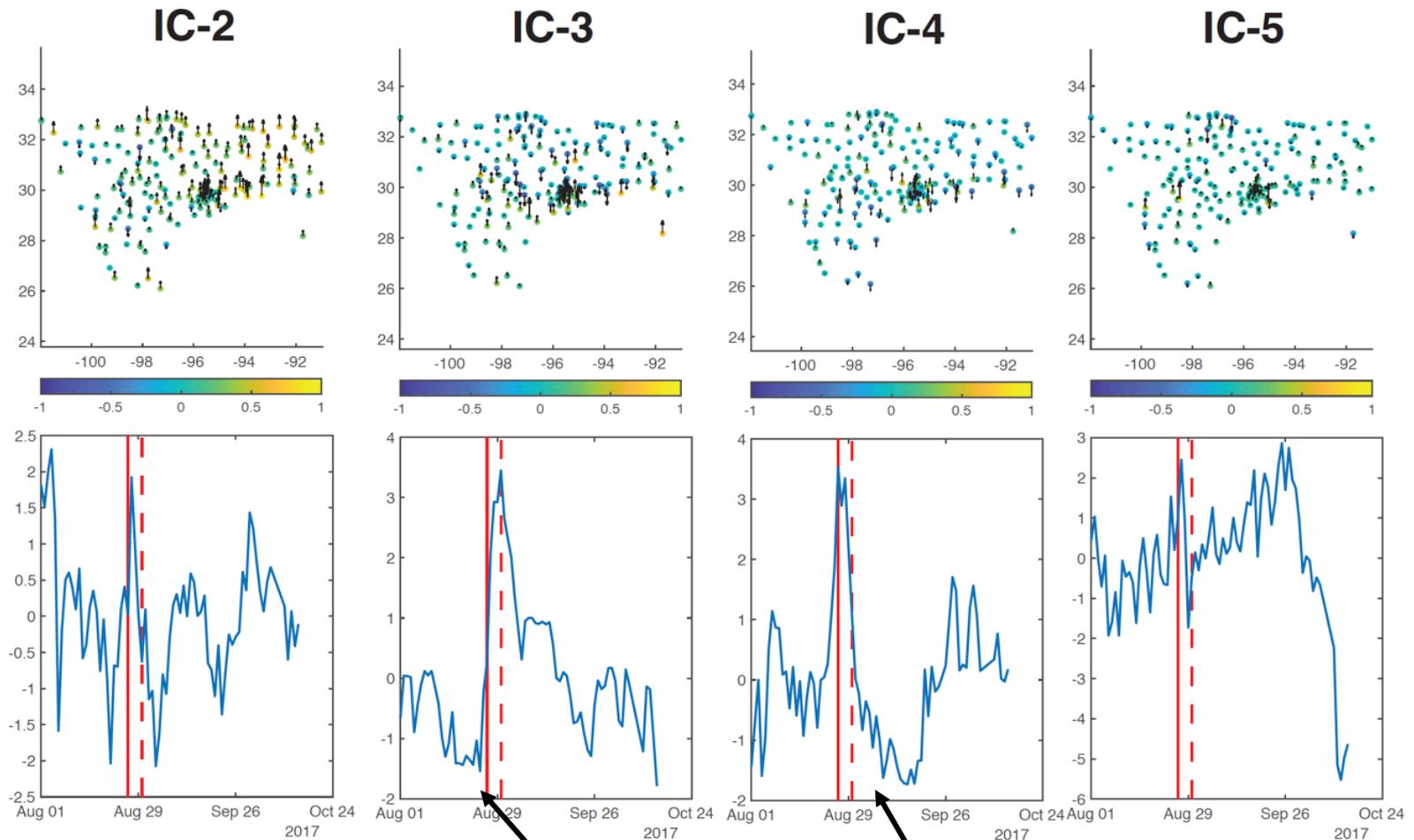


Hydrologic signal:
Coincident with
initial landfall

© 2018. All rights reserved.

Hydrologic signal:
Coincident with
second landfall

North component (after CME removed)

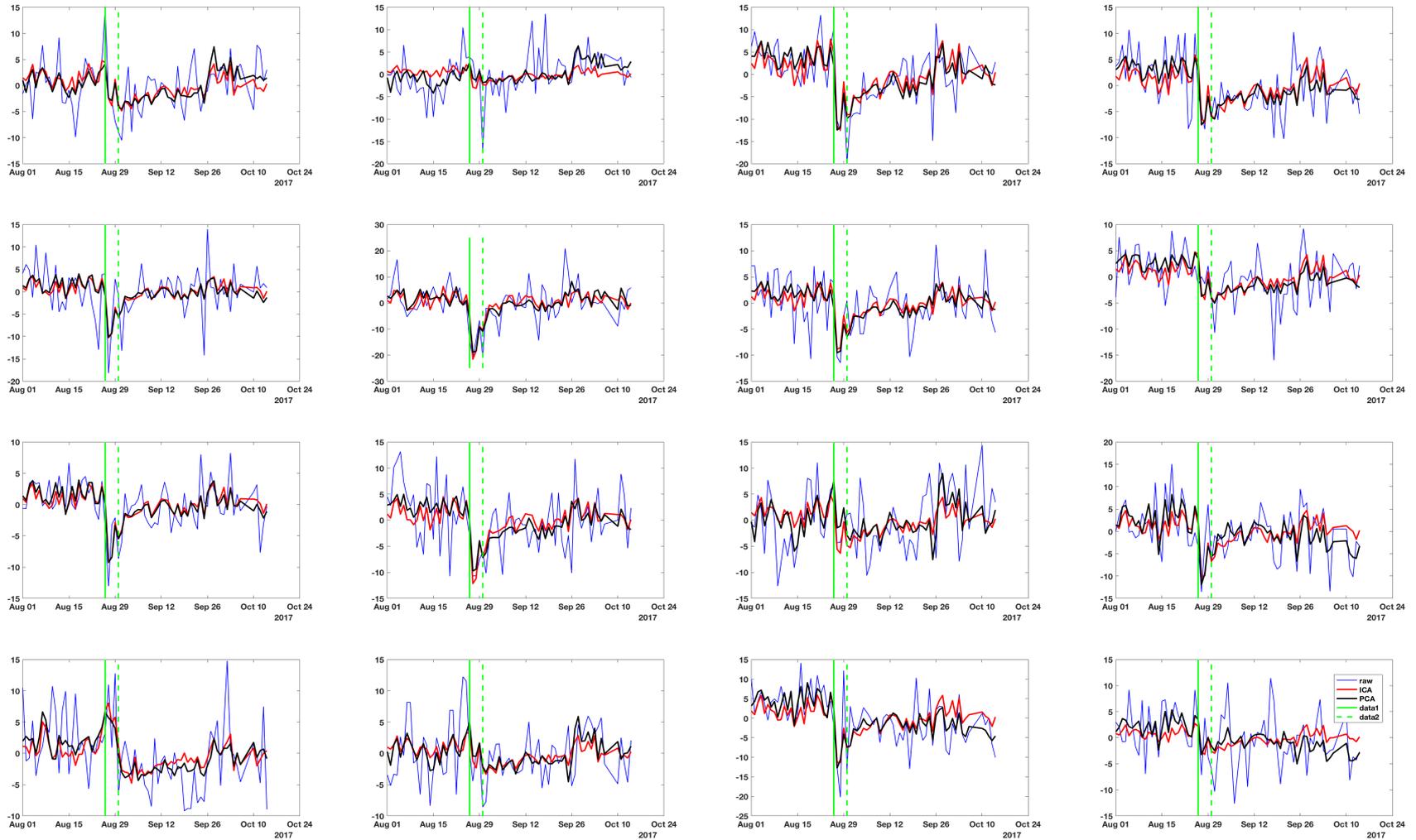


Other components not kept, show temporally incoherent motion or spatial patterns not consistent with Harvey's landfall pattern

Hydrologic signal:
Coincident with
initial landfall

Hydrologic signal:
Coincident with
second landfall

Filtered data

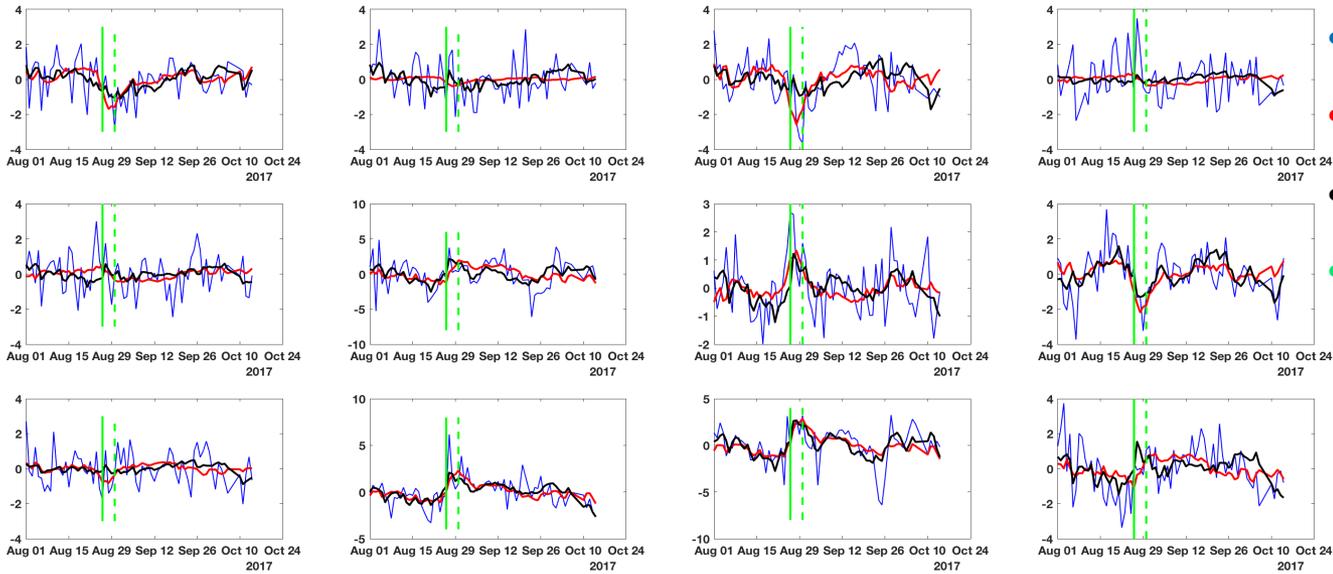


- PCA and ICA give similar results.
- Marked subsidence coincident with Harvey
- Return to pre-Harvey positions after ~5 weeks

- **Blue = raw**
- **Red = ICA**
- **Black = PCA**
- **Green = Harvey landfall**

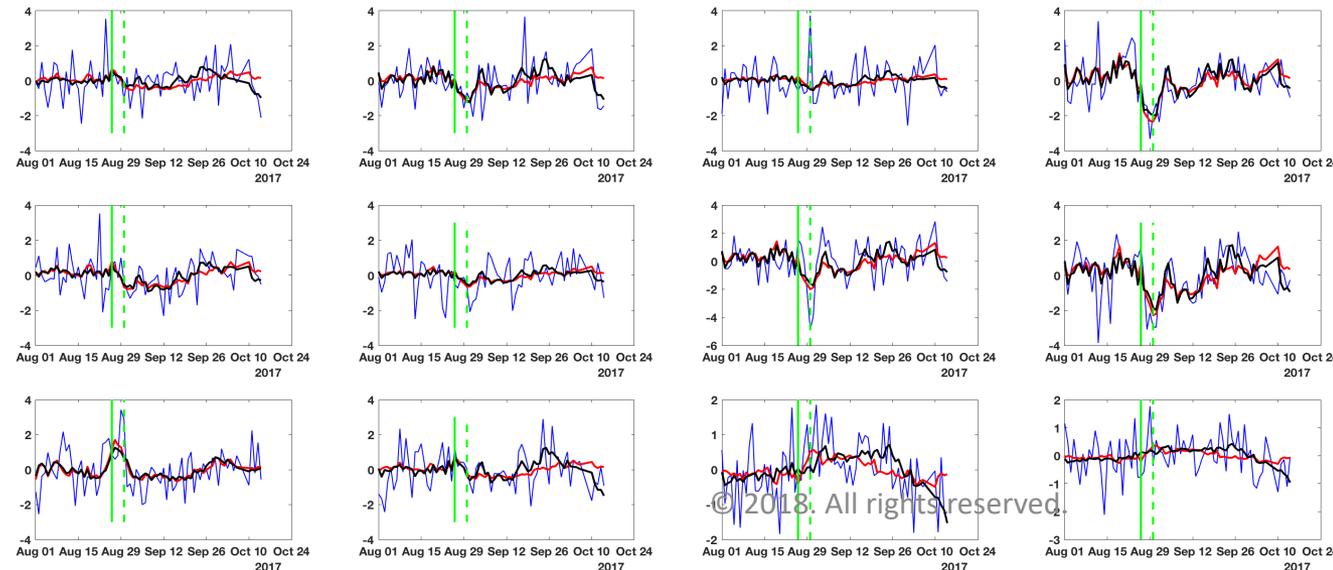
Data horizontals

North



- Blue = raw
- Red = ICA
- Black = PCA
- Green = Harvey landfall

East



- Up to 4 mm change in position
- Coincident with Harvey
- Return to pre-Harvey positions after ~5 weeks

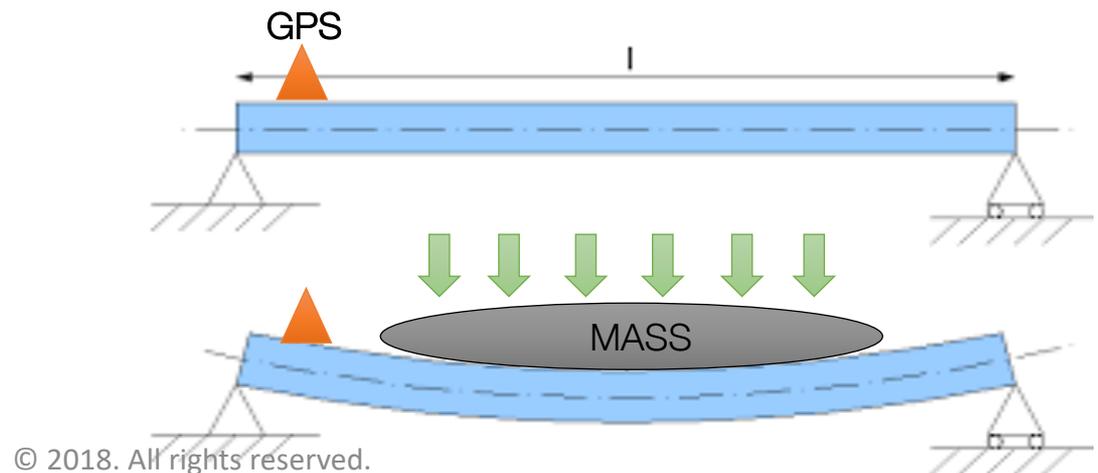
Motions in different areas

- 1. Largest subsidence around Houston (up to 21 mm).
- 2. Fastest uplift around Houston.
- 3. Smaller subsidence to east (up to 8 mm), along Texas-Louisiana border
- 4. Return of positions to pre-Harvey values after ~5 weeks. Possible small subsidence remains - remaining groundwater...?

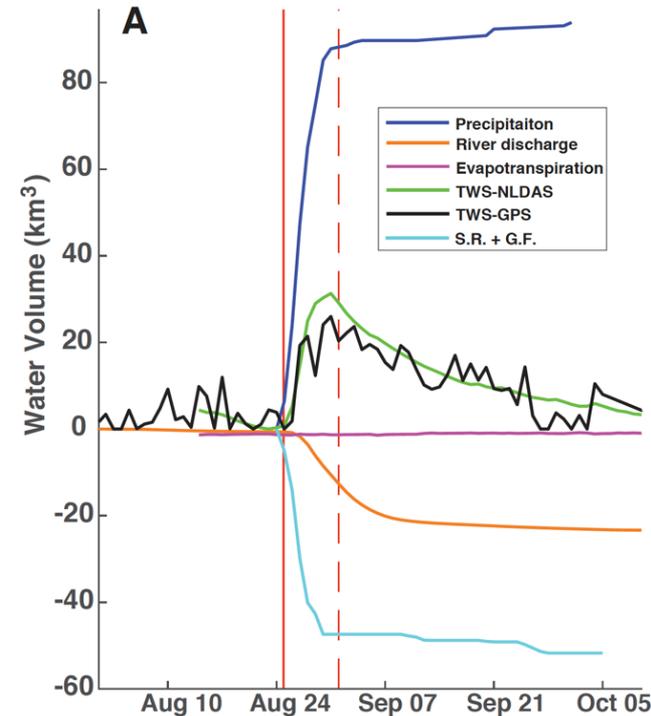
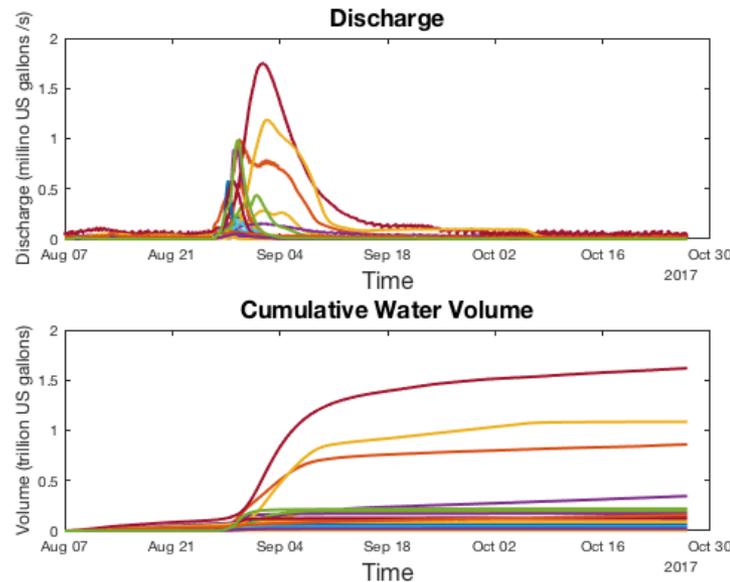
Invert GPS (E,N,V) → water thickness

- Invert subsidence for water mass (Farrell, 1972).
- Assume a 1D layered, spherical elastic structure - PREM velocity model (Dziewonski and Anderson, 1981).

$$\begin{bmatrix} WG_v \\ WG_u \\ WG_u \\ \lambda S \\ \beta U \end{bmatrix} [m_t] = \begin{bmatrix} Wd_t^v \\ Wd_t^e \\ Wd_t^n \\ 0 \\ \beta U m_{t-1} \end{bmatrix}$$



Discussion – Components of the hydrologic system



Surface runoff
+ groundwater
flow

$$S = P - \Delta TWS - ET - R$$

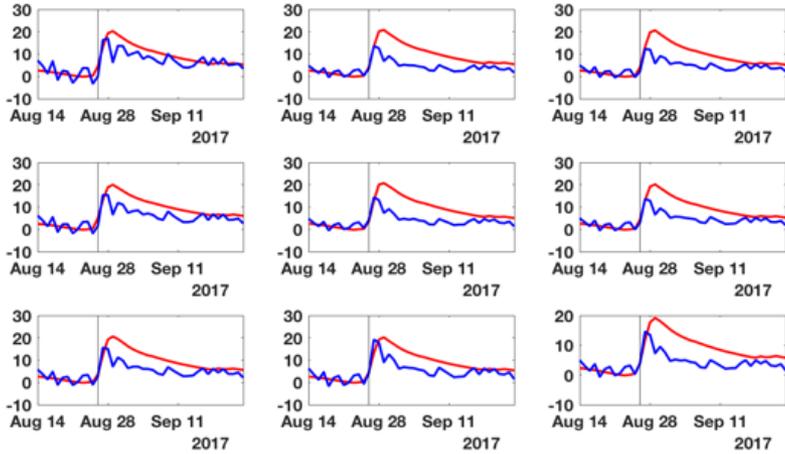
Precipitation
Water storage
Evapotranspiration

River
discharge

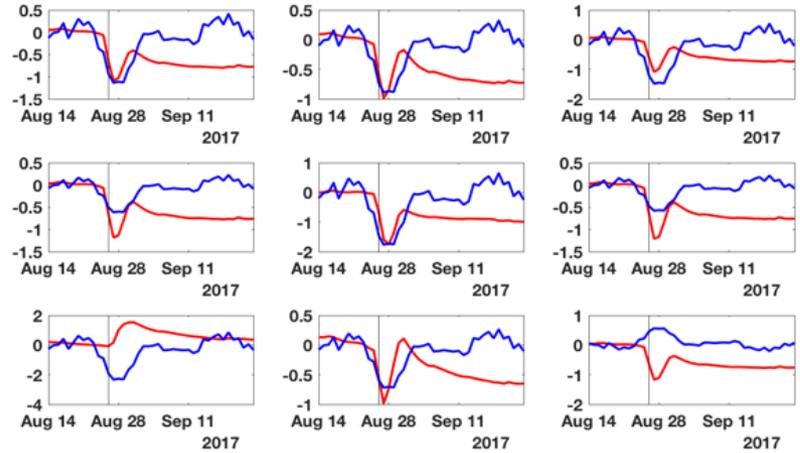
- River discharge from 31 USGS gauges – accounts for 25 km³ water loss, (minimum) ~27% of total,
- Evapotranspiration – accounts for ~18% of water loss, estimated from Fisher et al. (2008) using:
 - FLUXNET eddy covariance towers --> water + energy fluxes
 - MODIS instrument for radiation and vegetation indices
- Surface runoff and groundwater flow not well constrained.
- Closing water budget we estimate (maximum) ~50 km³ of water lost via S , ~54% of total water.

Validation

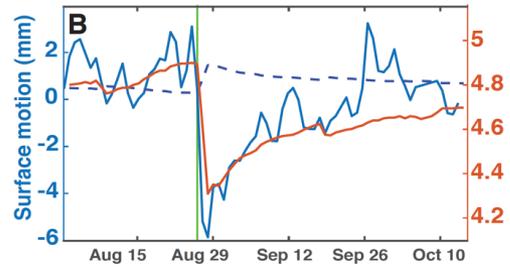
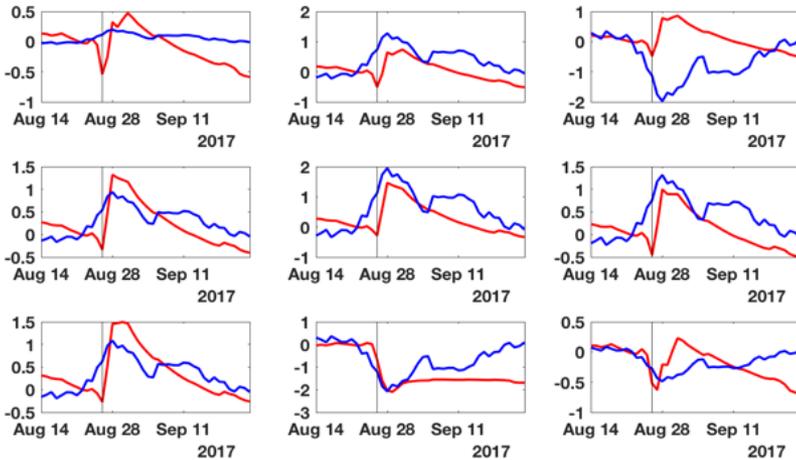
Vertical



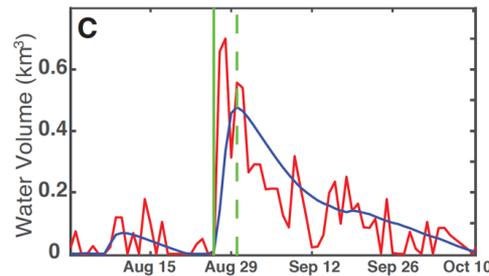
East



North



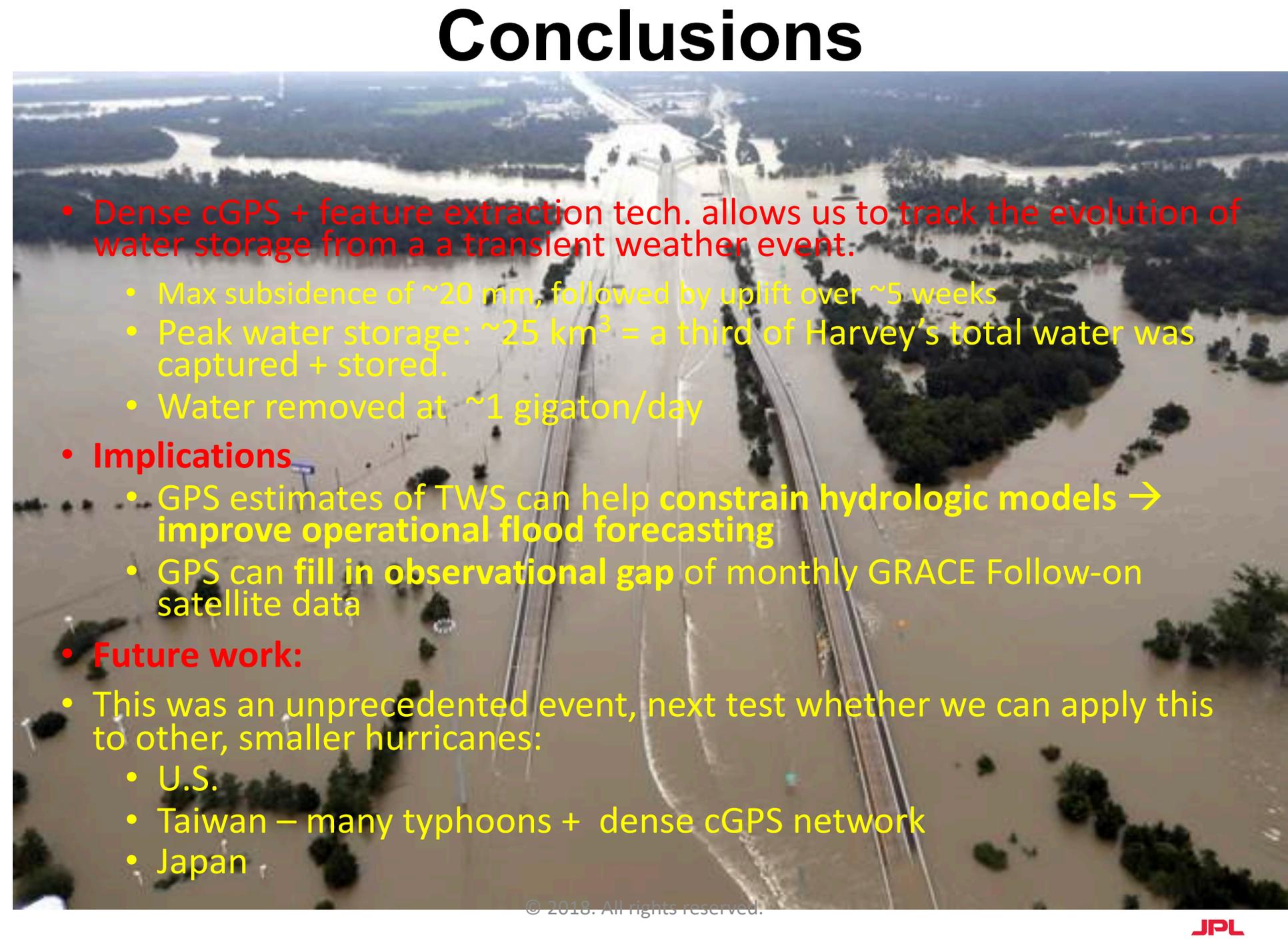
Well head water levels are inconsistent with poroelastic motions



USGS water volume measured at Barker and Addicks Reservoirs vs our TWS estimate

- NLDAS – National Land Data Assimilation model – hydrologic model driven by:
 - observed precipitation.
 - Shortwave radiation.
- Simulates TWS hourly and at $1/8^\circ$ degree

Conclusions

- 
- **Dense cGPS + feature extraction tech. allows us to track the evolution of water storage from a transient weather event.**
 - Max subsidence of ~ 20 mm, followed by uplift over ~ 5 weeks
 - Peak water storage: $\sim 25 \text{ km}^3$ = a third of Harvey's total water was captured + stored.
 - Water removed at ~ 1 gigaton/day
 - **Implications**
 - GPS estimates of TWS can help **constrain hydrologic models** \rightarrow **improve operational flood forecasting**
 - GPS can **fill in observational gap** of monthly GRACE Follow-on satellite data
 - **Future work:**
 - This was an unprecedented event, next test whether we can apply this to other, smaller hurricanes:
 - U.S.
 - Taiwan – many typhoons + dense cGPS network
 - Japan