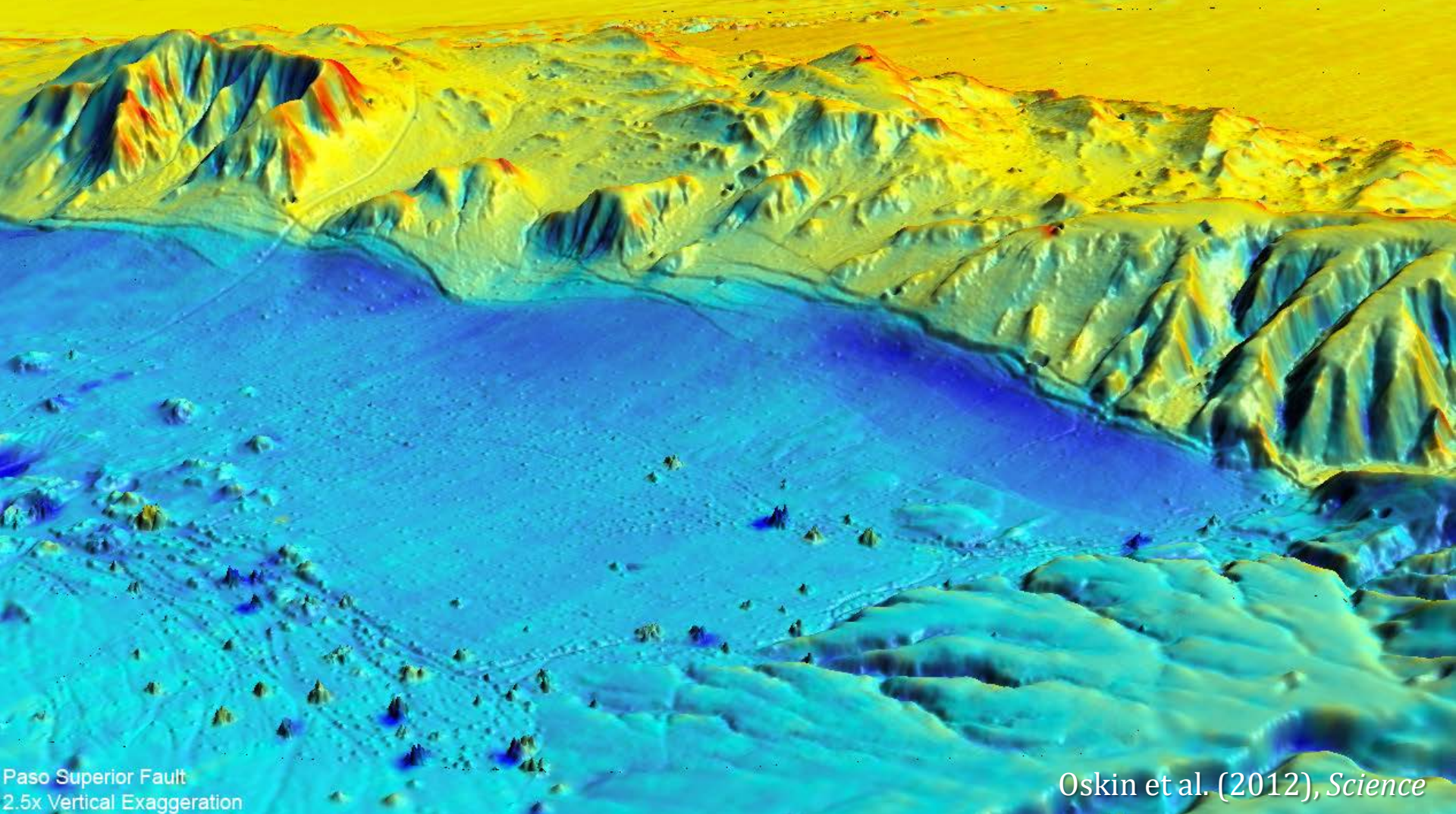


Aligning point clouds and topographic change detection



Edwin Nissen (Colorado School of Mines)

Thanks to: Ramon Arrowsmith, Srikanth Saripalli, Aravindhan Krishnan (ASU), Adrian Borsa (Scripps), Craig Glennie (Houston), Alejandro Hinojosa-Corona (CICESE), Tadashi Maruyama (AIST), Austin Elliott, Mike Oskin (UC Davis)

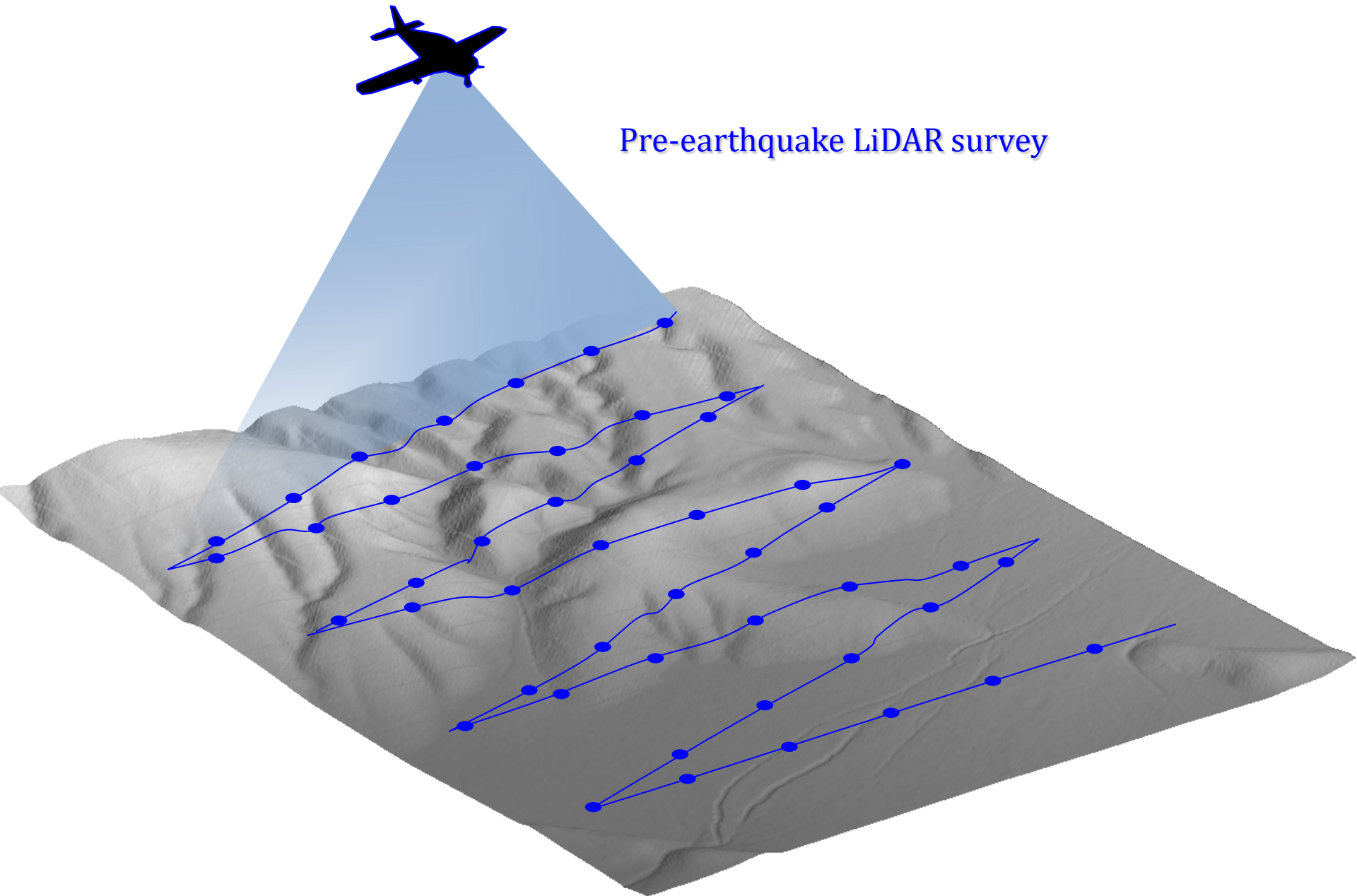
Aligning point clouds and topographic change detection

- Multi-temporal topography
- Earthquake examples:
 - aligning (registering) topography data with ICP
 - 2011 Fukushima earthquake (Japan)
- Other applications

Paso Superior Fault
2.5x Vertical Exaggeration

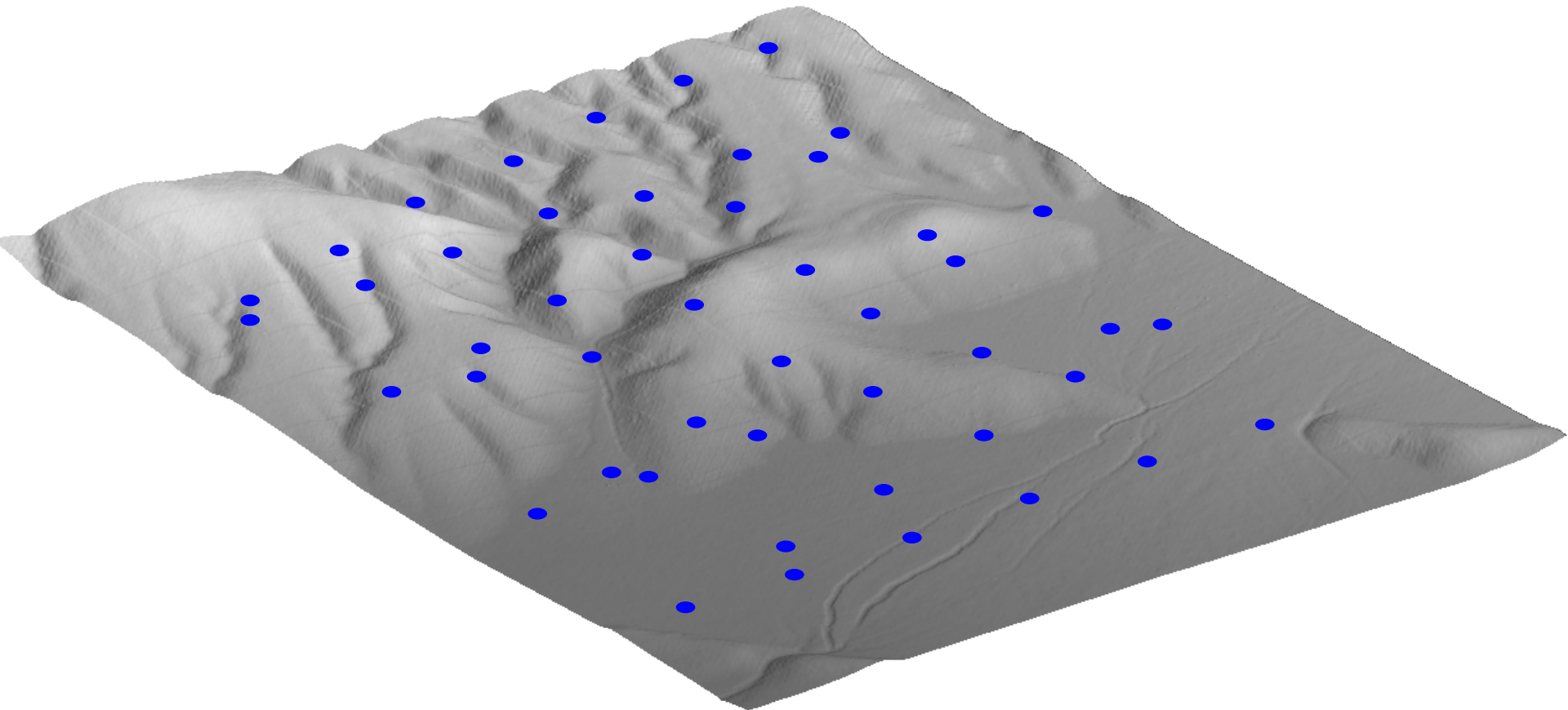


3-D earthquake deformation from repeat lidar

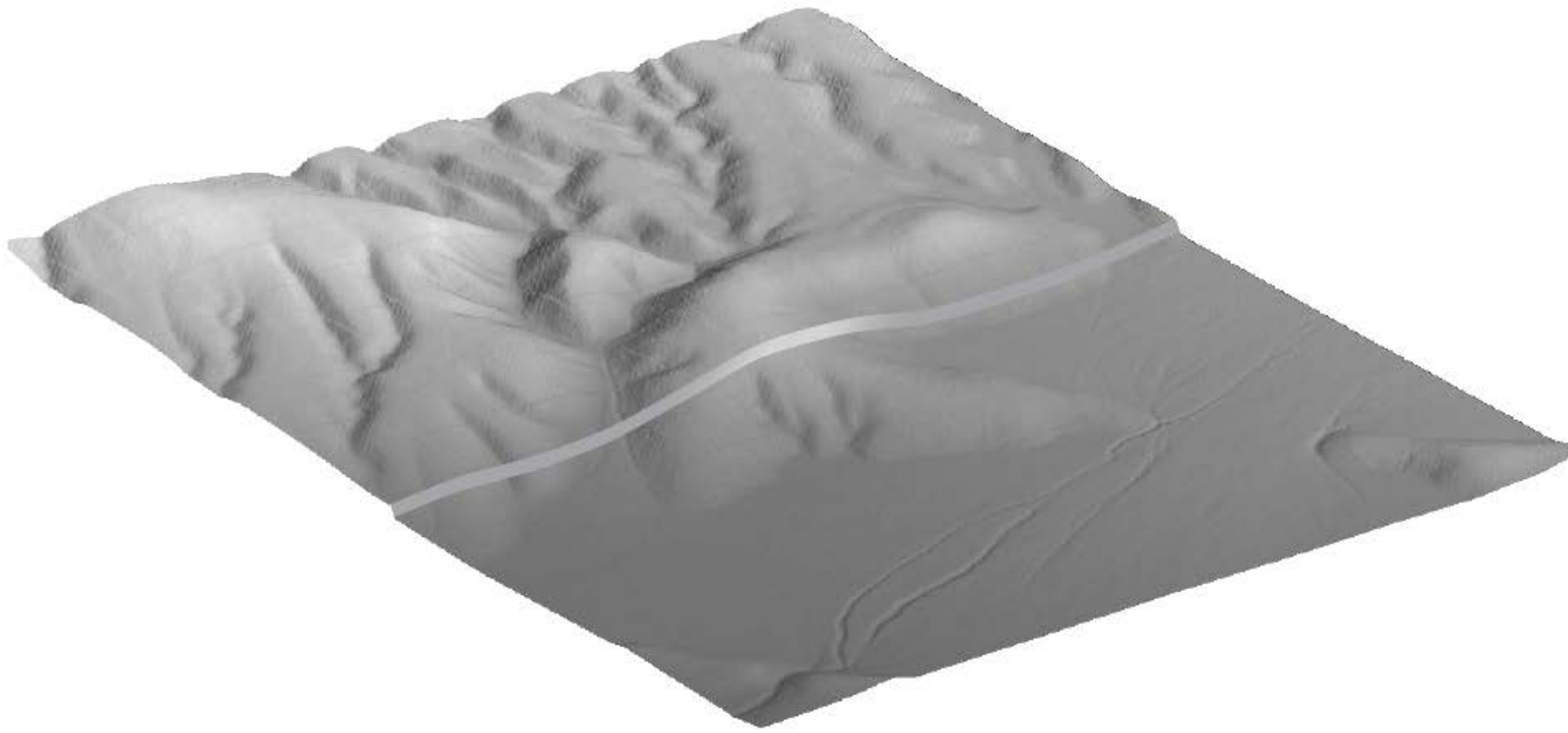


3-D earthquake deformation from repeat lidar

Pre-earthquake point cloud



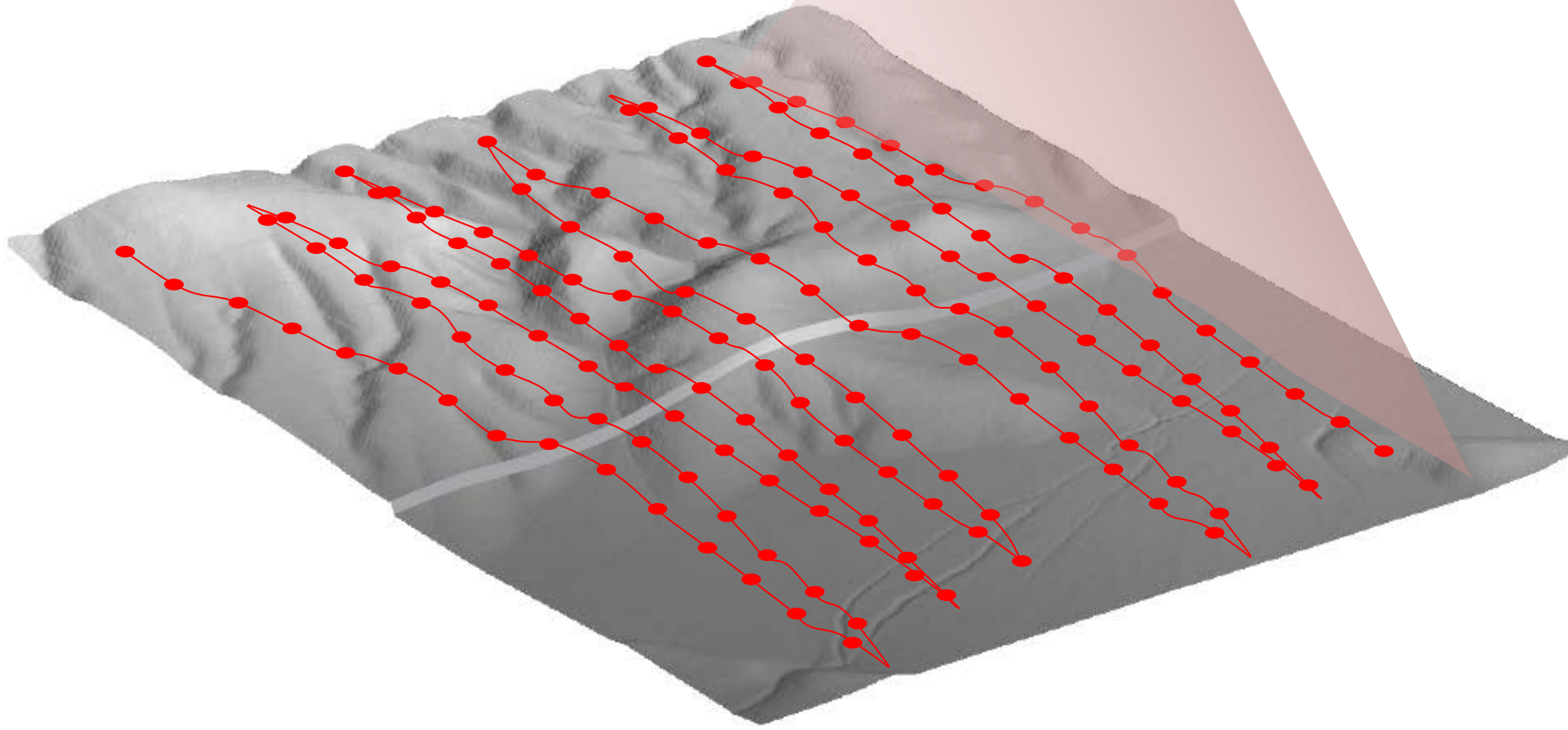
3-D earthquake deformation from repeat lidar



3-D earthquake deformation from repeat lidar

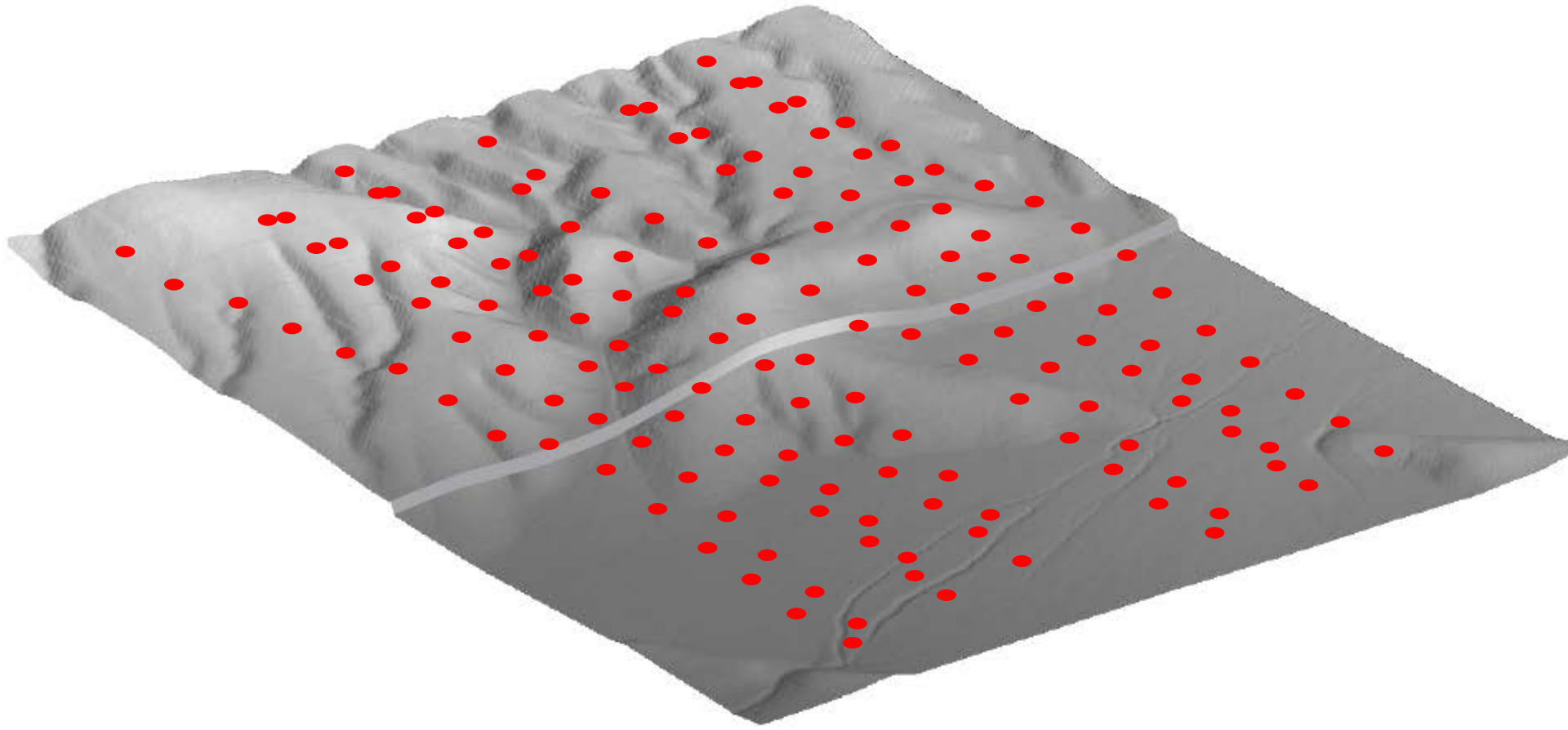


Post-earthquake LiDAR survey



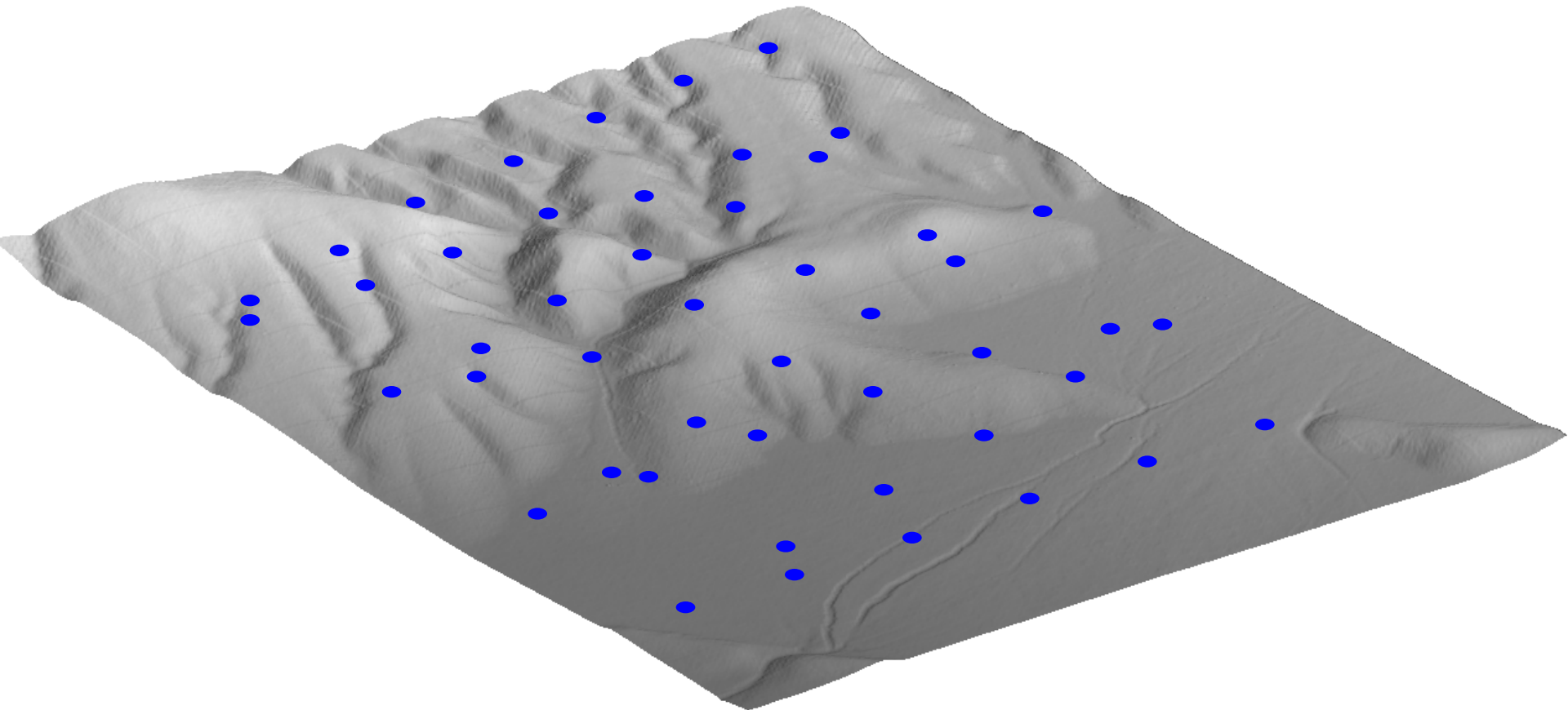
3-D earthquake deformation from repeat lidar

Post-earthquake point cloud



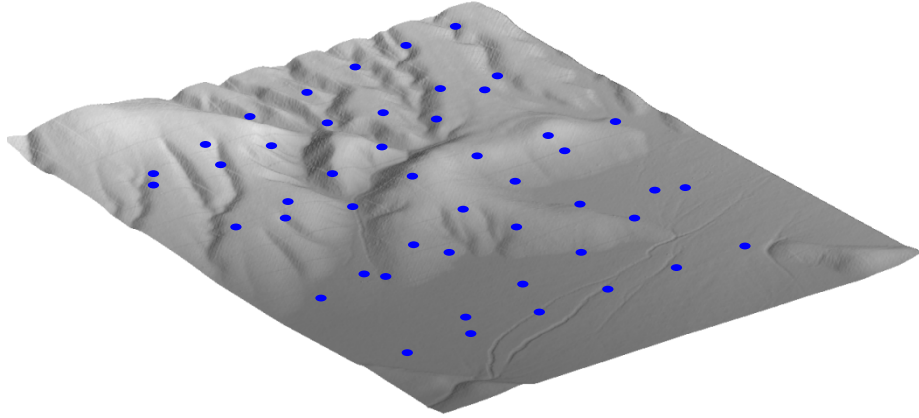
3-D earthquake deformation from repeat lidar

Pre-earthquake point cloud

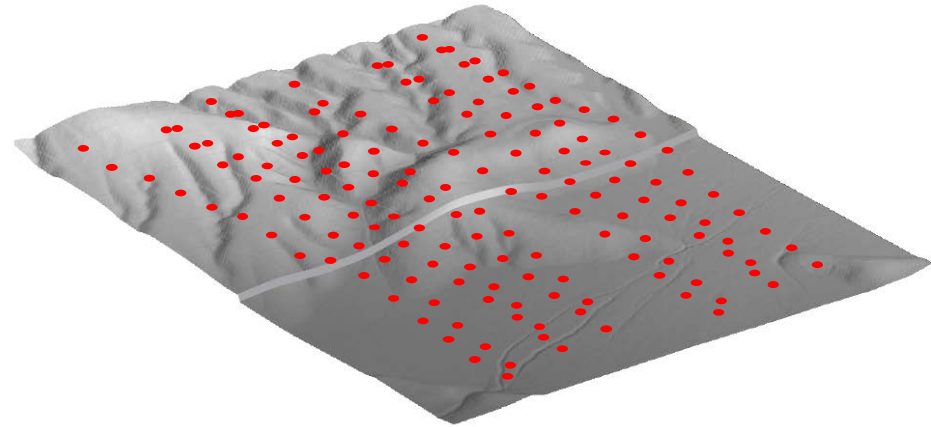


3-D earthquake deformation from repeat lidar

Pre-earthquake point cloud



Post-earthquake point cloud

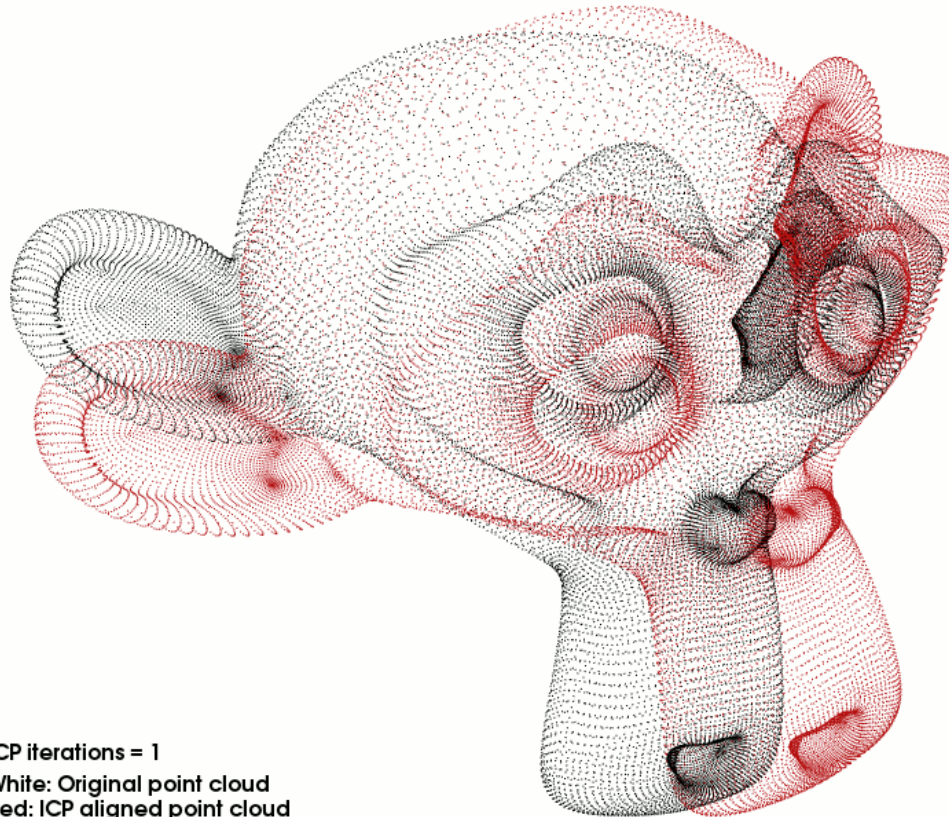


The Challenges of LiDAR differencing

- Data are irregularly spaced (we can rasterize them, but lose information doing so)
- There can be large mismatches in point density (legacy datasets vs modern surveys)
- ... and mismatches in data quality and metrics (third party vs research-grade)
- Treatment of vegetation returns in forested areas

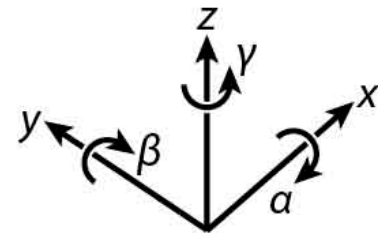
3-D earthquake deformation from repeat lidar

- The **iterative closest point** algorithm (ICP) is a method for registering (aligning) irregular point clouds, well known in computer vision and medical imaging
- ICP minimizes closest point pair distances using iterative **rigid-body transformations**, each one comprising a **translation** $[t_x t_y t_z]$ and a **rotation** $[\alpha \beta \gamma]$



ICP iterations = 1
White: Original point cloud
Red: ICP aligned point cloud

$$\Phi = \begin{pmatrix} 1 & -\gamma & \beta & t_x \\ \gamma & 1 & -\alpha & t_y \\ -\beta & \alpha & 1 & t_z \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

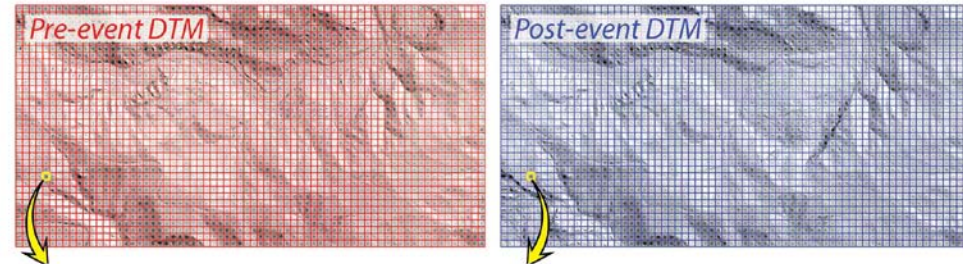


3-D earthquake deformation from repeat lidar

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- ICP minimizes closest point pair distances using iterative **rigid-body transformations**, each one comprising a **translation** $[t_x t_y t_z]$ and a **rotation** $[\alpha \beta \gamma]$

- (1) the two LiDAR datasets are first split into square “cells”

1 Split both datasets into square cells



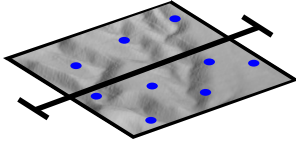
- (2) ICP is run on each equivalent pair of cells.

The **translation** $[t_x t_y t_z]$ corresponds to the cell displacement

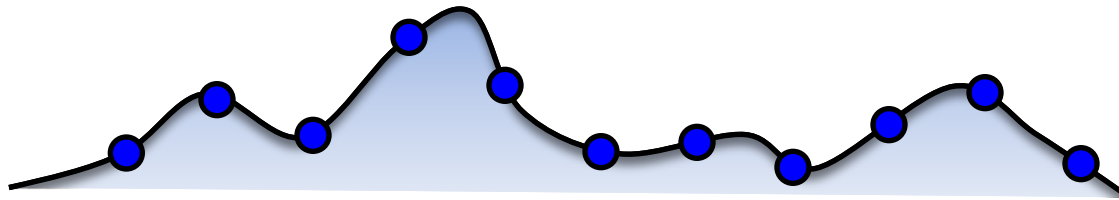
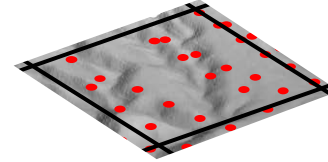
- (3) this is repeated for the next pair of cells

Iterative Closest Point algorithm (ICP)

Pre-earthquake cell

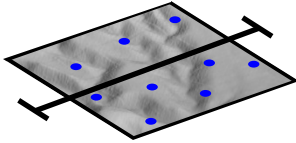


Post-earthquake cell

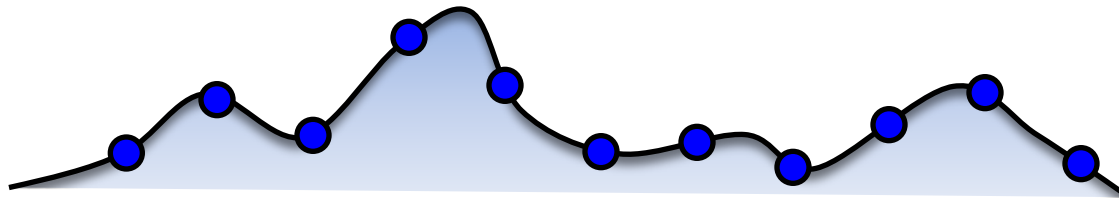
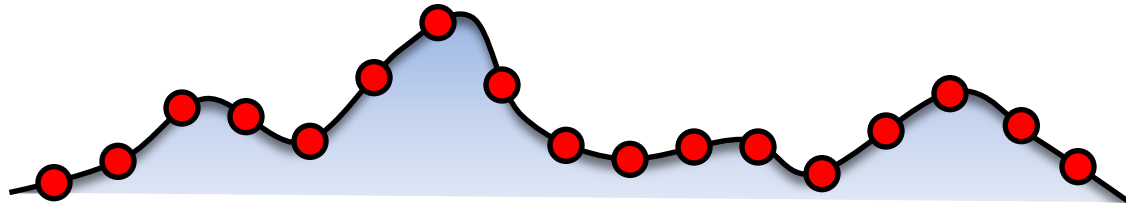
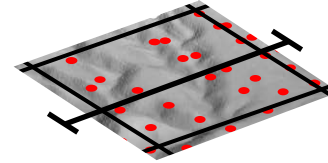


Iterative Closest Point algorithm (ICP)

Pre-earthquake cell

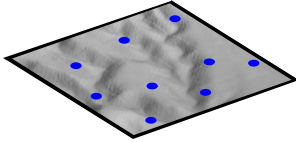


Post-earthquake cell

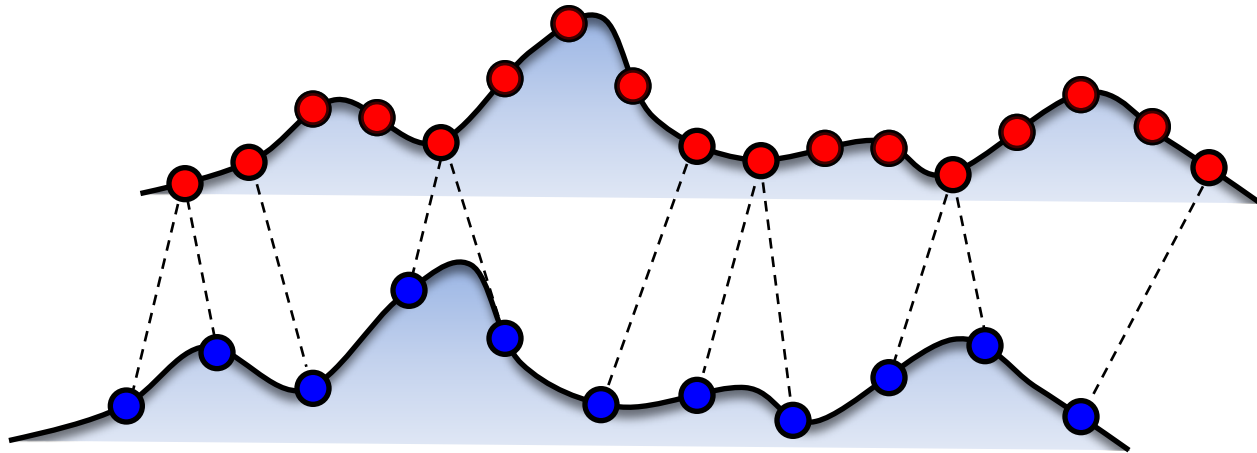
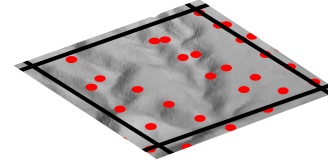


Iterative Closest Point algorithm (ICP)

Pre-earthquake cell

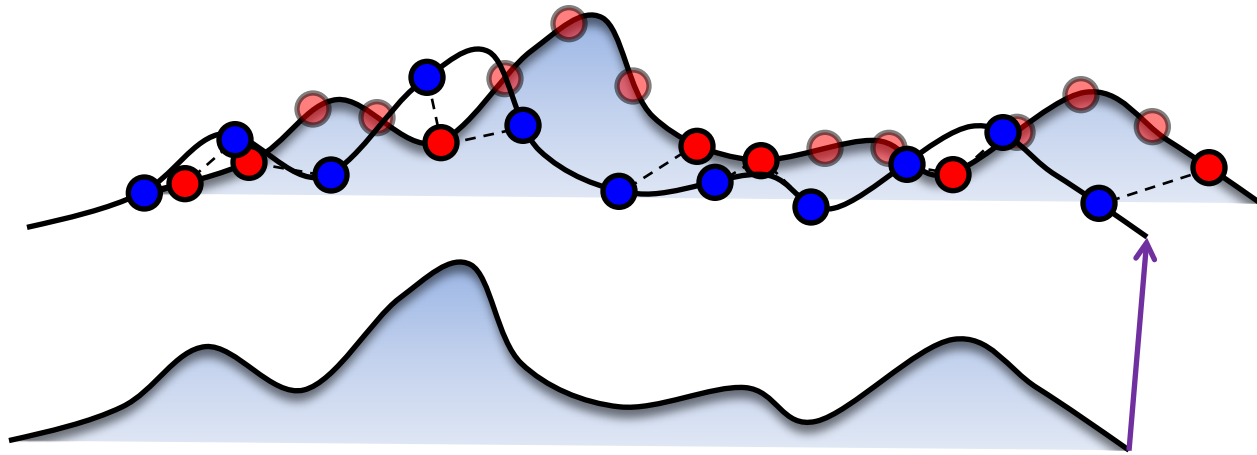



Post-earthquake cell



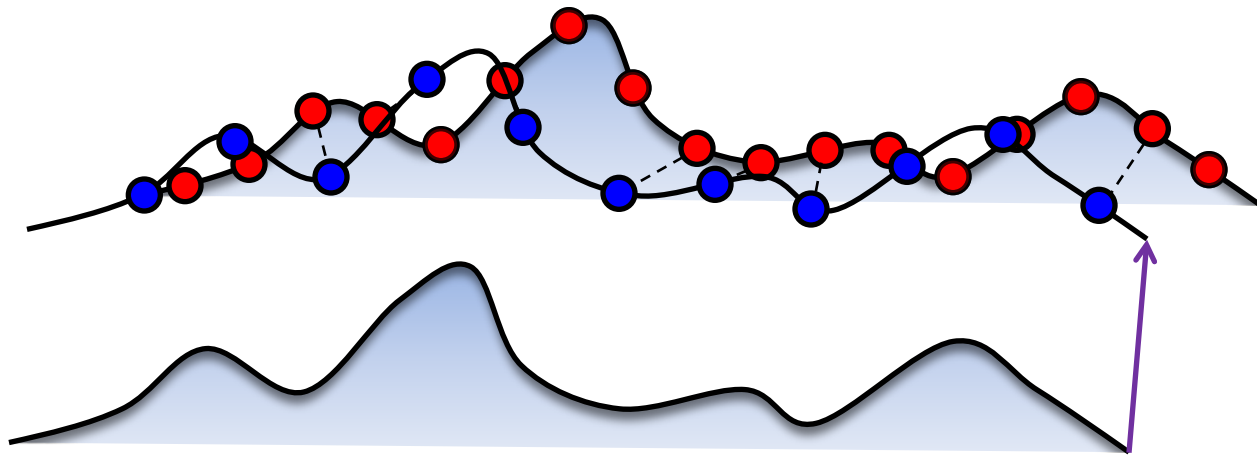
Find closest points

Iterative Closest Point algorithm (ICP)



Iterate  Find closest points
Transform point cloud $\phi = \begin{pmatrix} 1 & -\gamma & \beta & t_x \\ \gamma & 1 & -\alpha & t_y \\ -\beta & \alpha & 1 & t_z \\ 0 & 0 & 0 & 1 \end{pmatrix}$

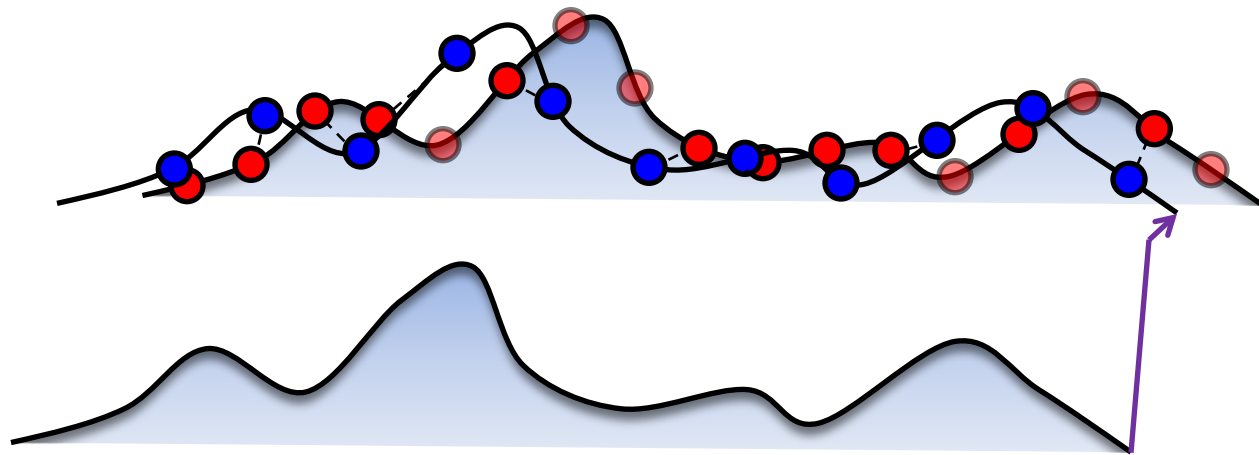
Iterative Closest Point algorithm (ICP)



Find closest points

Transform point cloud

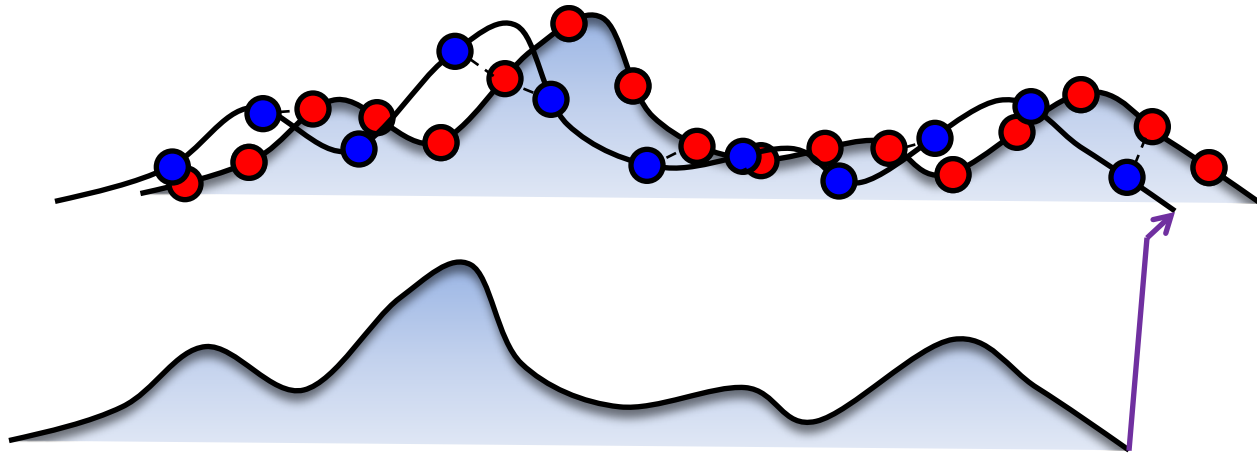
Iterative Closest Point algorithm (ICP)



Find closest points

Transform point cloud

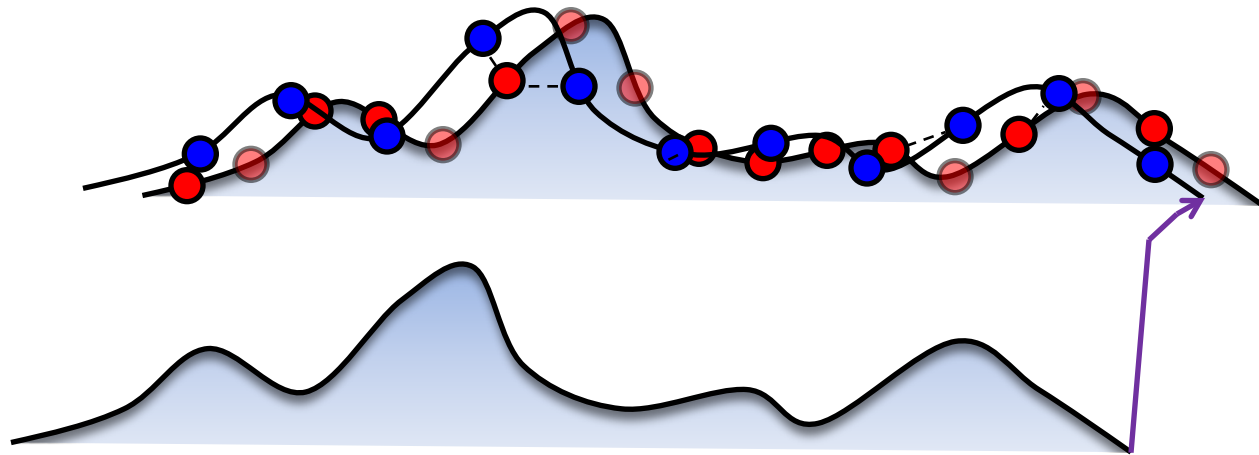
Iterative Closest Point algorithm (ICP)



Find closest points

Transform point cloud

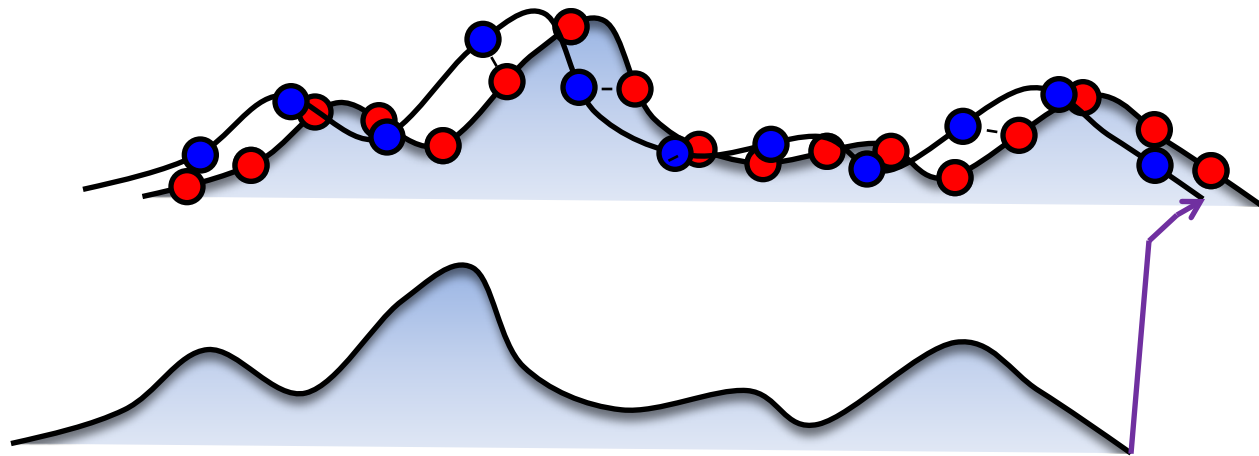
Iterative Closest Point algorithm (ICP)



Find closest points

Transform point cloud

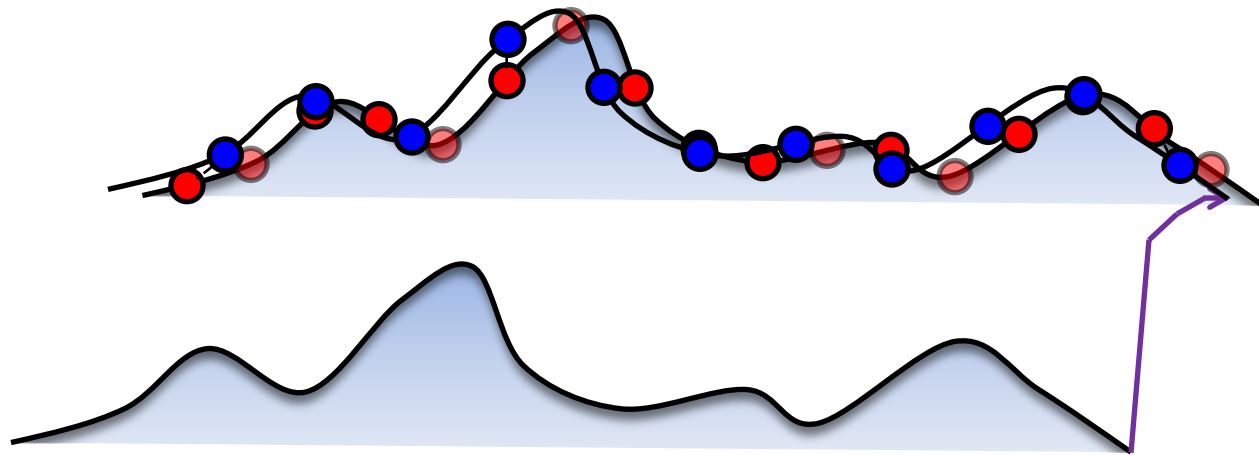
Iterative Closest Point algorithm (ICP)



Find closest points

Transform point cloud

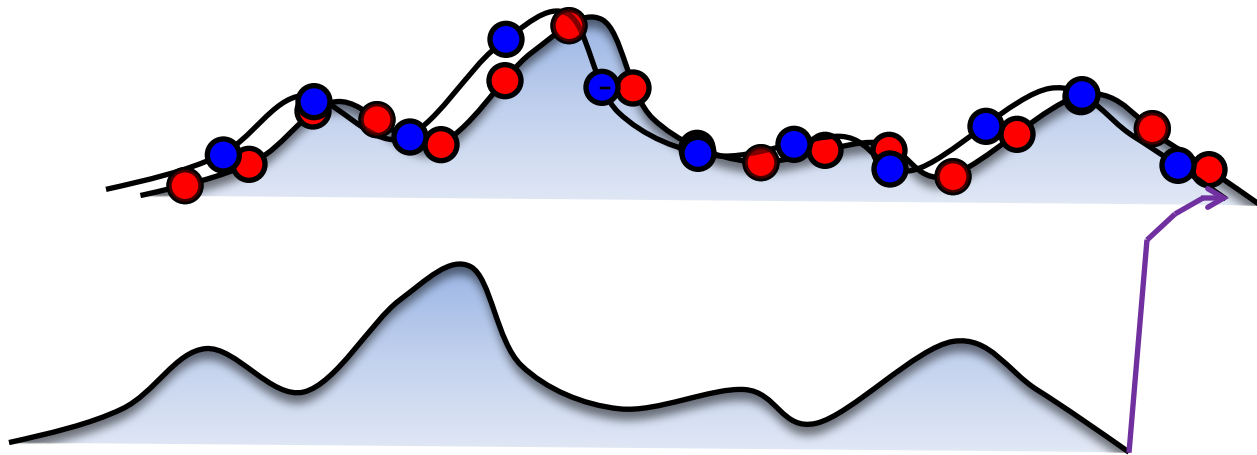
Iterative Closest Point algorithm (ICP)



Find closest points

Transform point cloud

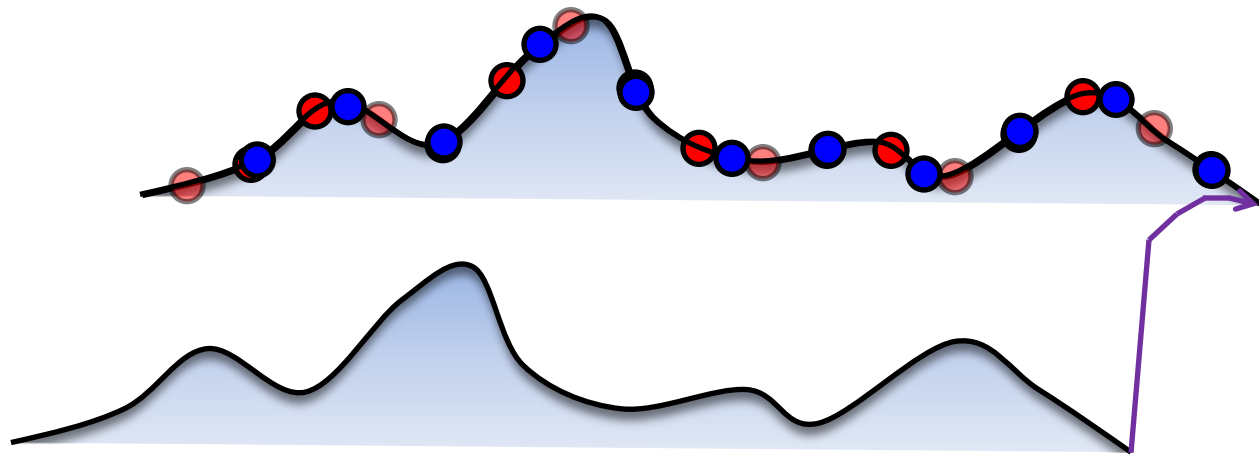
Iterative Closest Point algorithm (ICP)



Find closest points

Transform point cloud

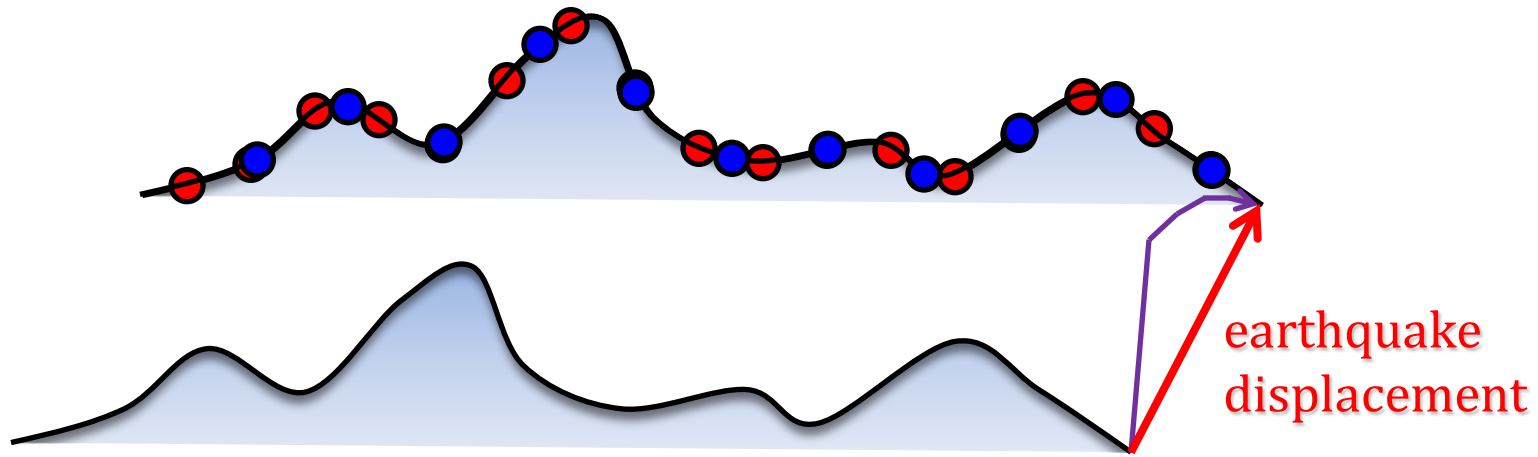
Iterative Closest Point algorithm (ICP)



Find closest points

Transform point cloud

Iterative Closest Point algorithm (ICP)



Find closest points

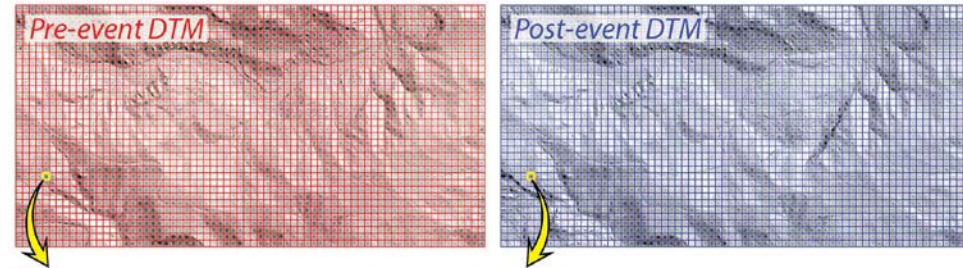
Transform point cloud

3-D earthquake deformation from repeat LiDAR point clouds

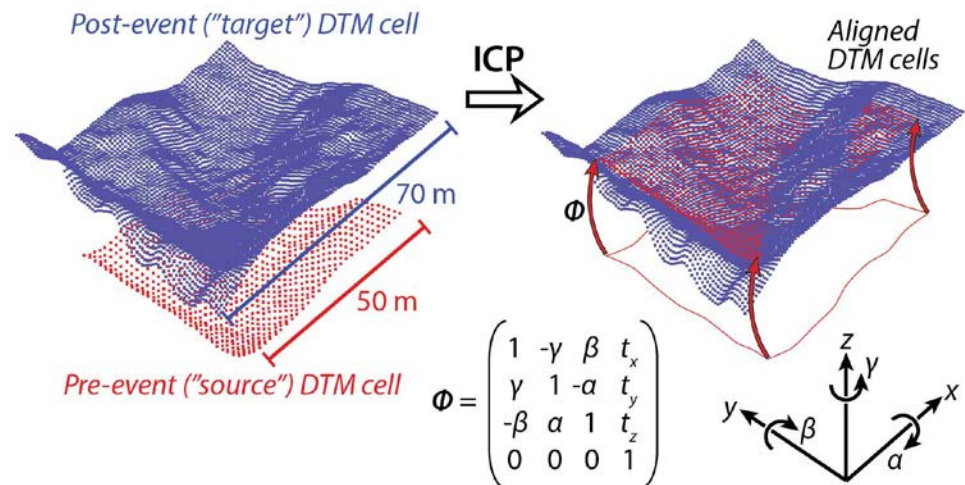
Caveats

- ICP will not work if there are large changes to the shape of the cell, e.g. through landsliding
- ICP will generate spurious results in areas that are very planar

1 Split both datasets into square cells

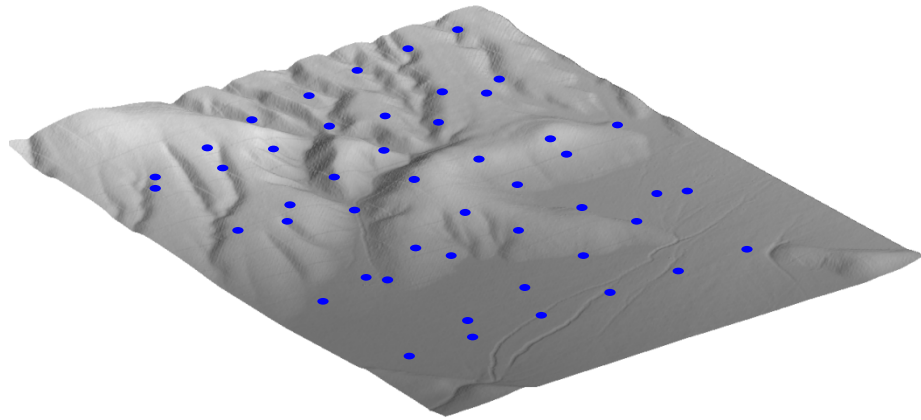


2 Take two equivalent cells and align with ICP

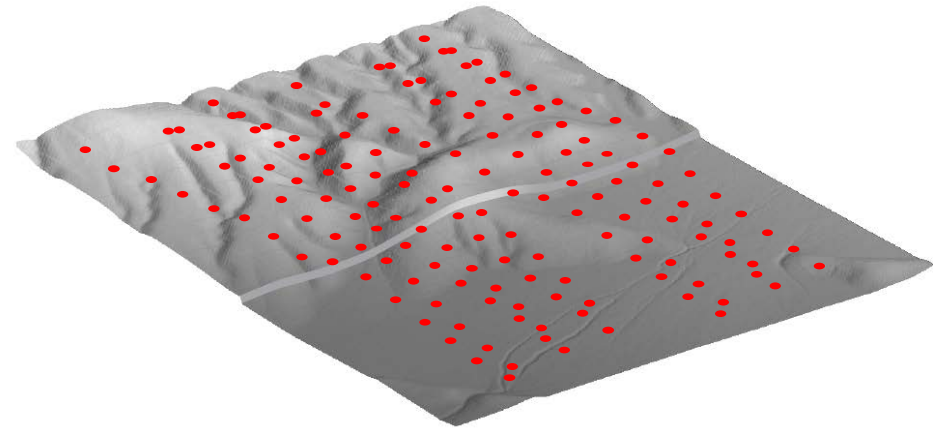


3 Move on to next pair of cells and repeat step 2

Pre-earthquake point cloud



Post-earthquake point cloud



The **Iterative Closest Point** algorithm: a method for registering (aligning) two sets of points

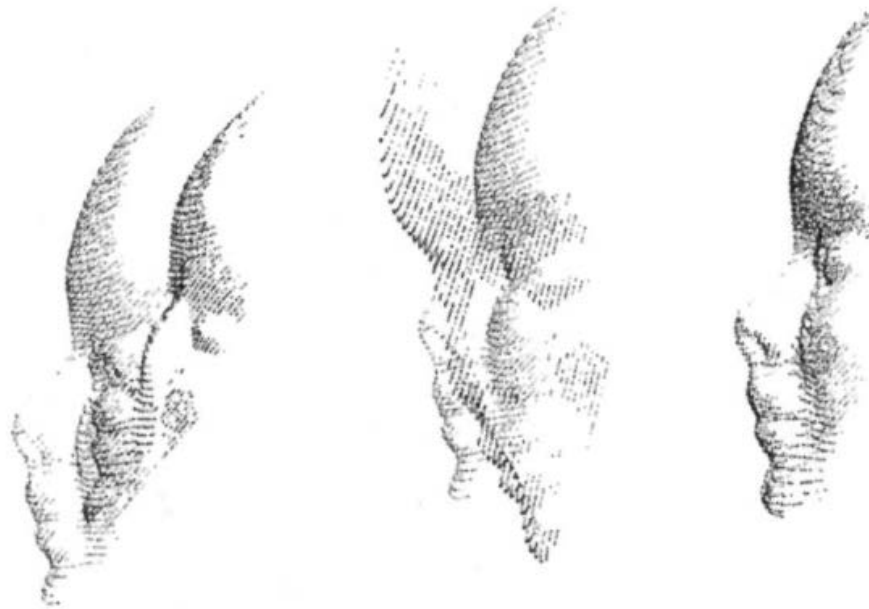
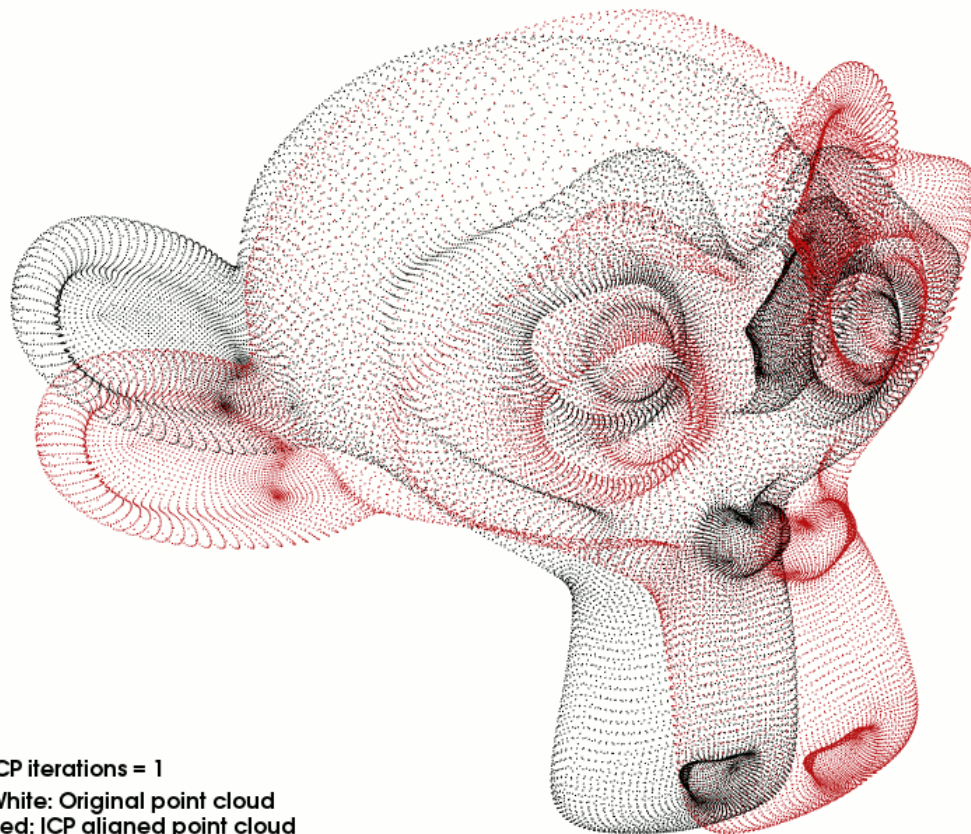


Fig. 4. Iterative point-based registration of phantom face range data

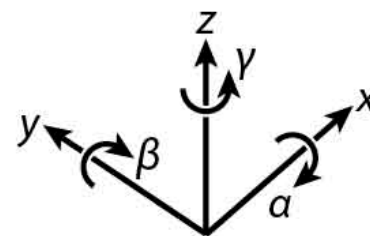
3-D earthquake deformation from repeat LiDAR point clouds

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- ICP minimizes closest point pair distances using iterative **rigid-body transformations**, each one comprising a **translation** $[t_x t_y t_z]$ and a **rotation** $[\alpha \beta \gamma]$

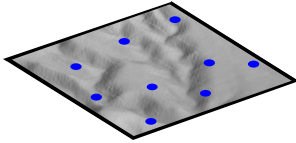


ICP iterations = 1
White: Original point cloud
Red: ICP aligned point cloud

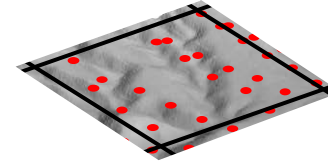
$$\Phi = \begin{pmatrix} 1 & -\gamma & \beta & t_x \\ \gamma & 1 & -\alpha & t_y \\ -\beta & \alpha & 1 & t_z \\ 0 & 0 & 0 & 1 \end{pmatrix}$$



Pre-earthquake point cloud



Post-earthquake point cloud

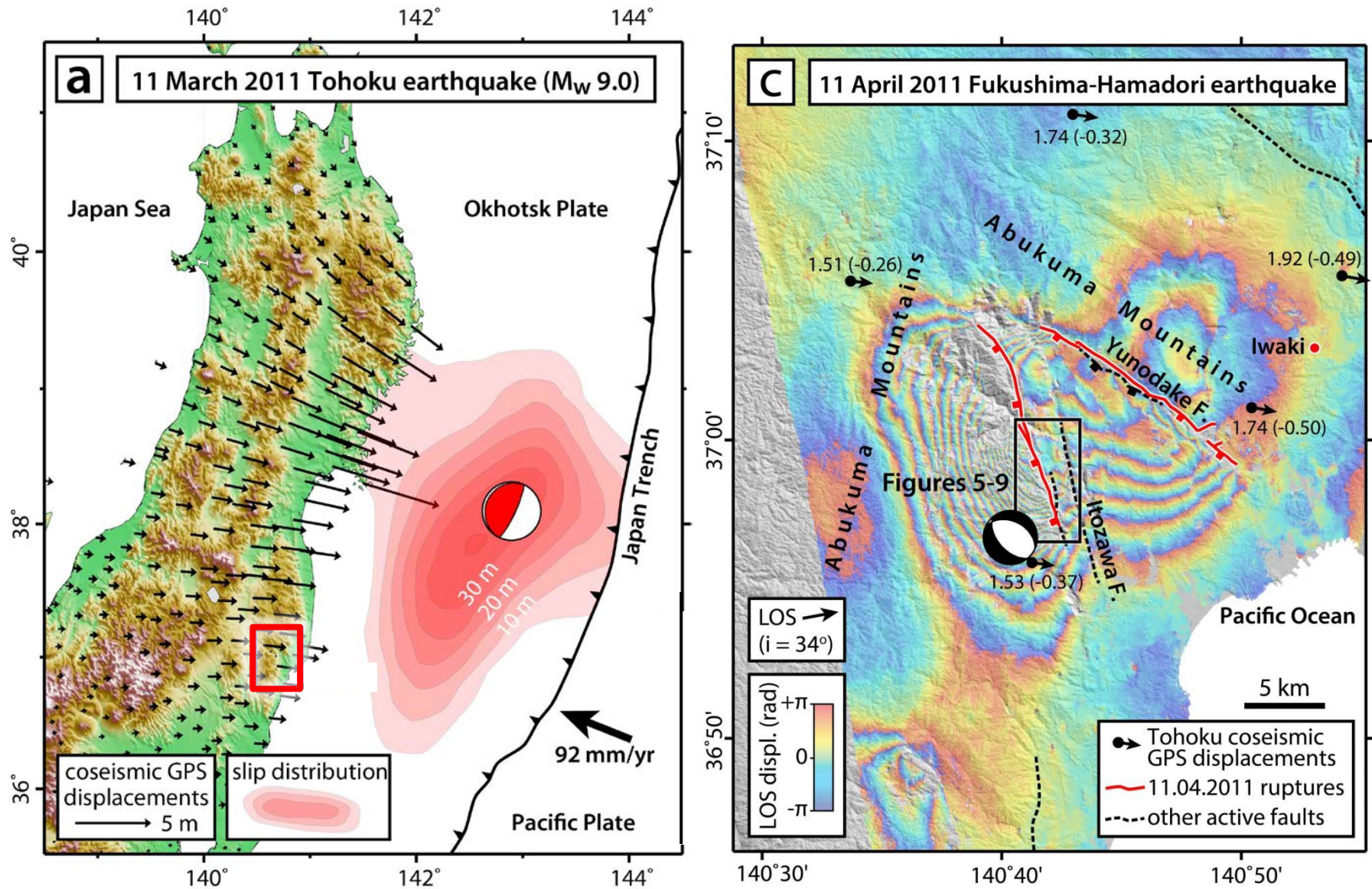


The **Iterative Closest Point** algorithm: a method for registering (aligning) two sets of points

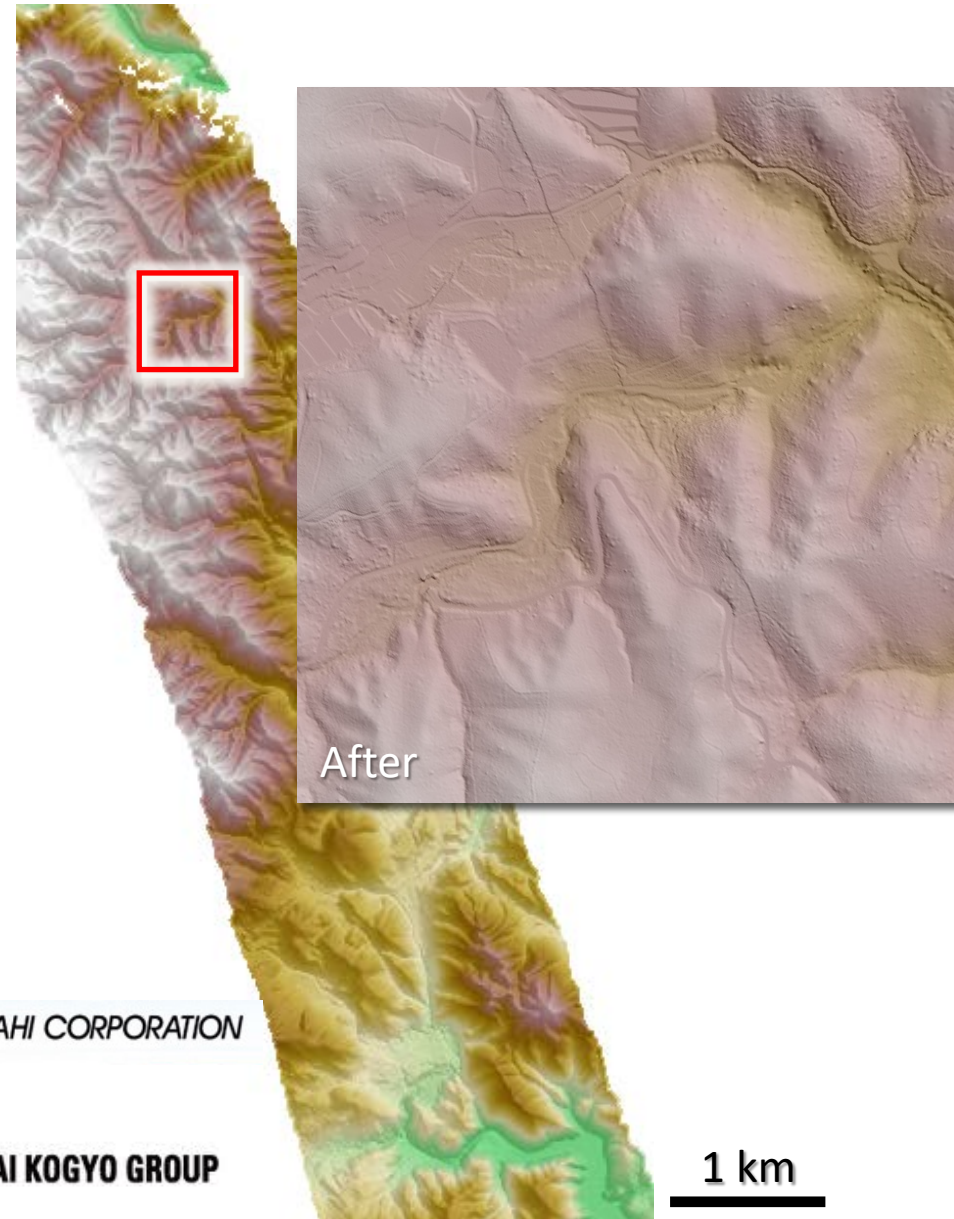
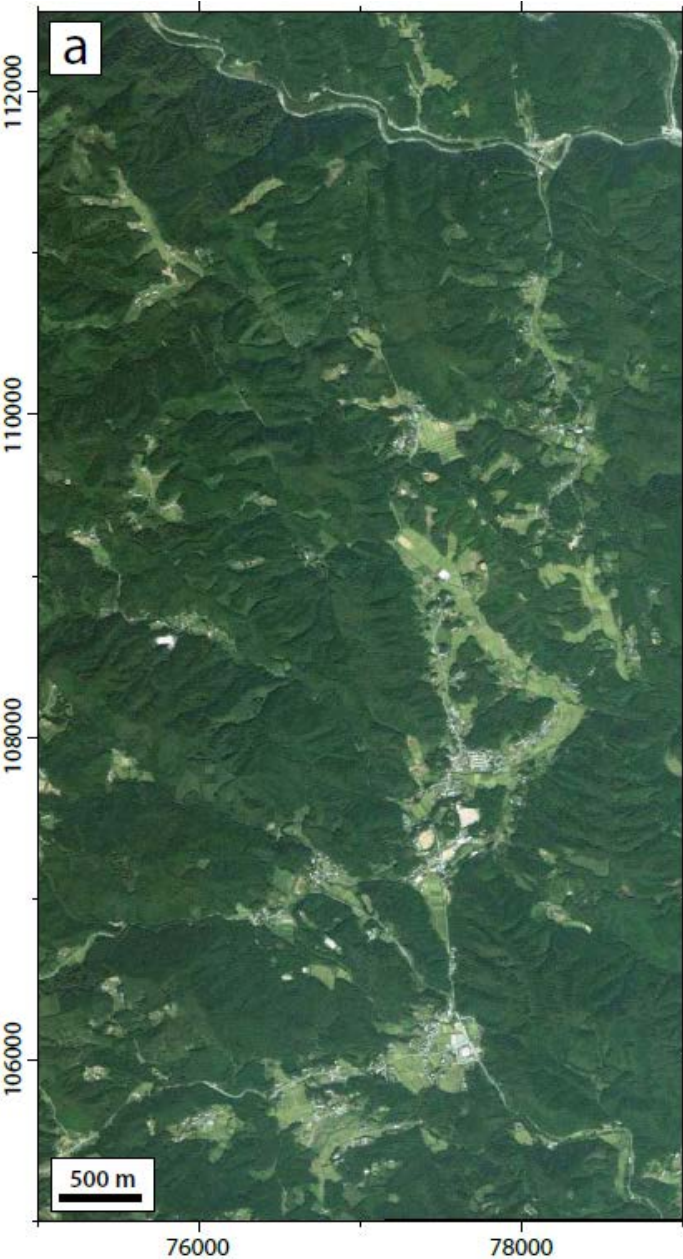
- the two point clouds are first split into square “windows”, 50 m in diameter
- ICP is run separately on each pair of windows. (An additional “fringe” of 5 m is included in the post-event window in order to capture the coseismic displacement)
- ICP finds the displacement and rotation that best aligns the pre-event and post-event point clouds.
- This alignment corresponds to the local coseismic displacement for that window.

see Nissen et al. (2012), Geophys. Res. Lett. and Nissen et al. (2014) EPSL for details
Imagery based methods are more mature and quite effective (mostly 2D)!

11 April 2011 Fukushima-Hamadori earthquake



11 April 2011 Fukushima-Hamadori earthquake



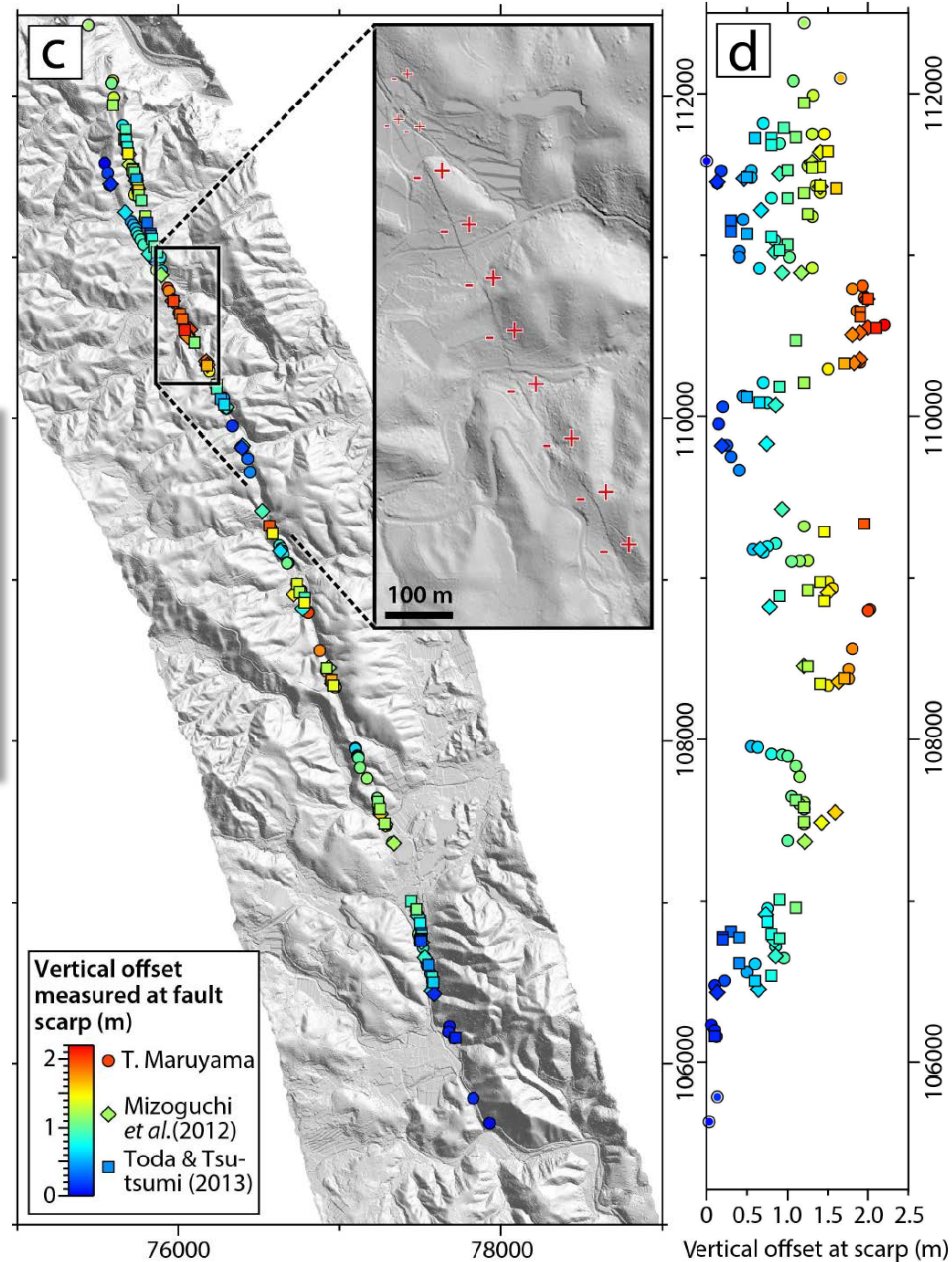
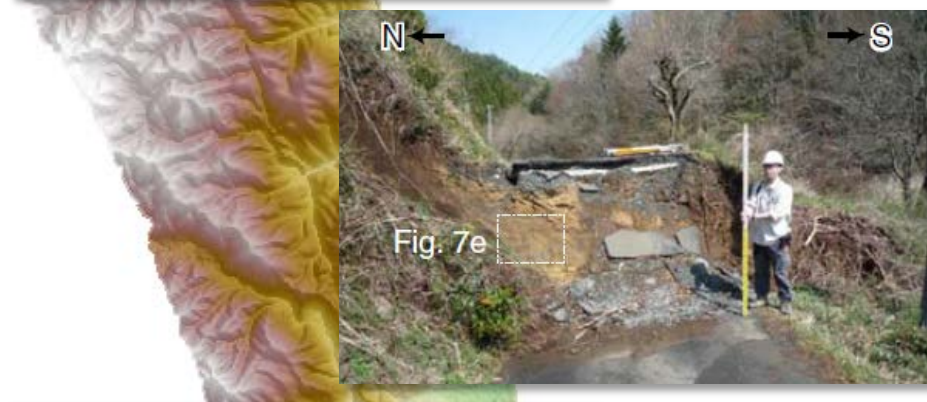
 AERO ASAHI CORPORATION

 KOKUSAI KOGYO GROUP

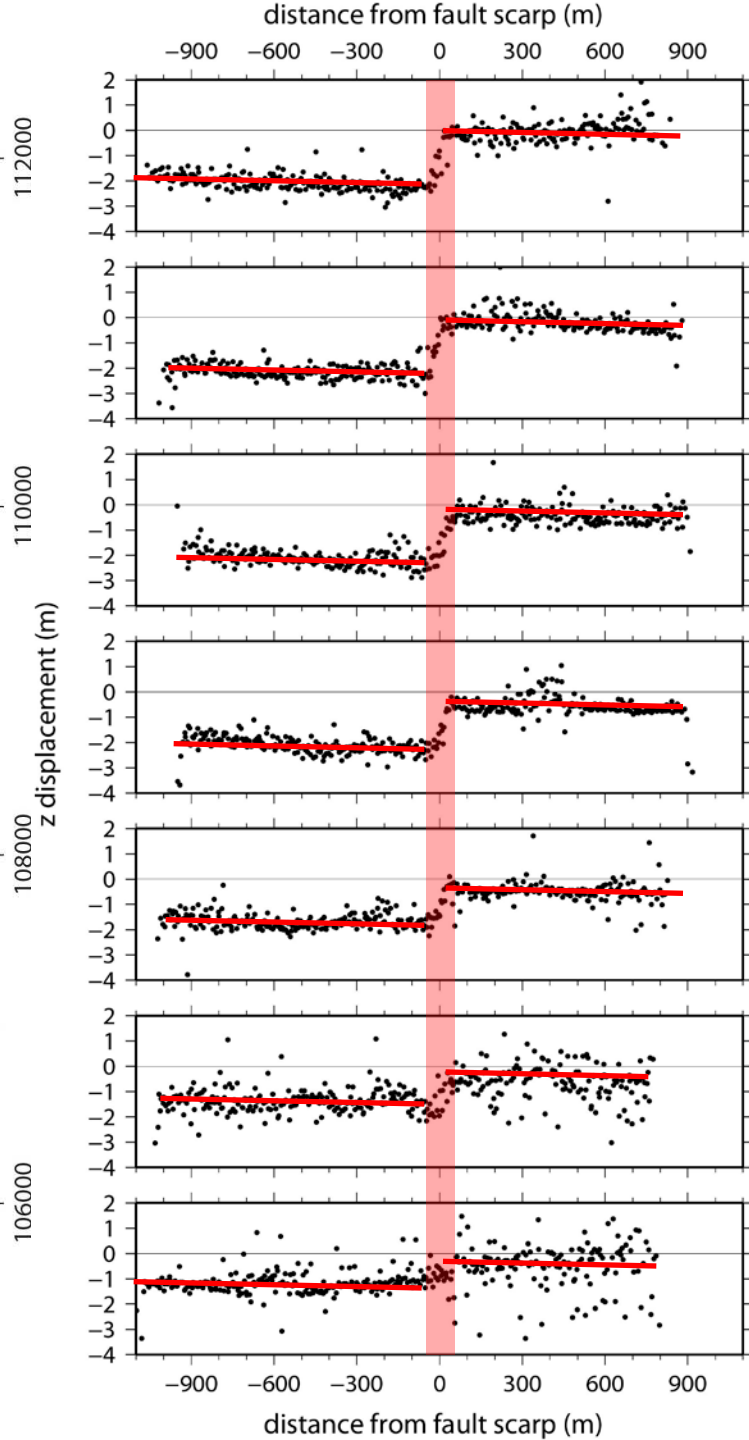
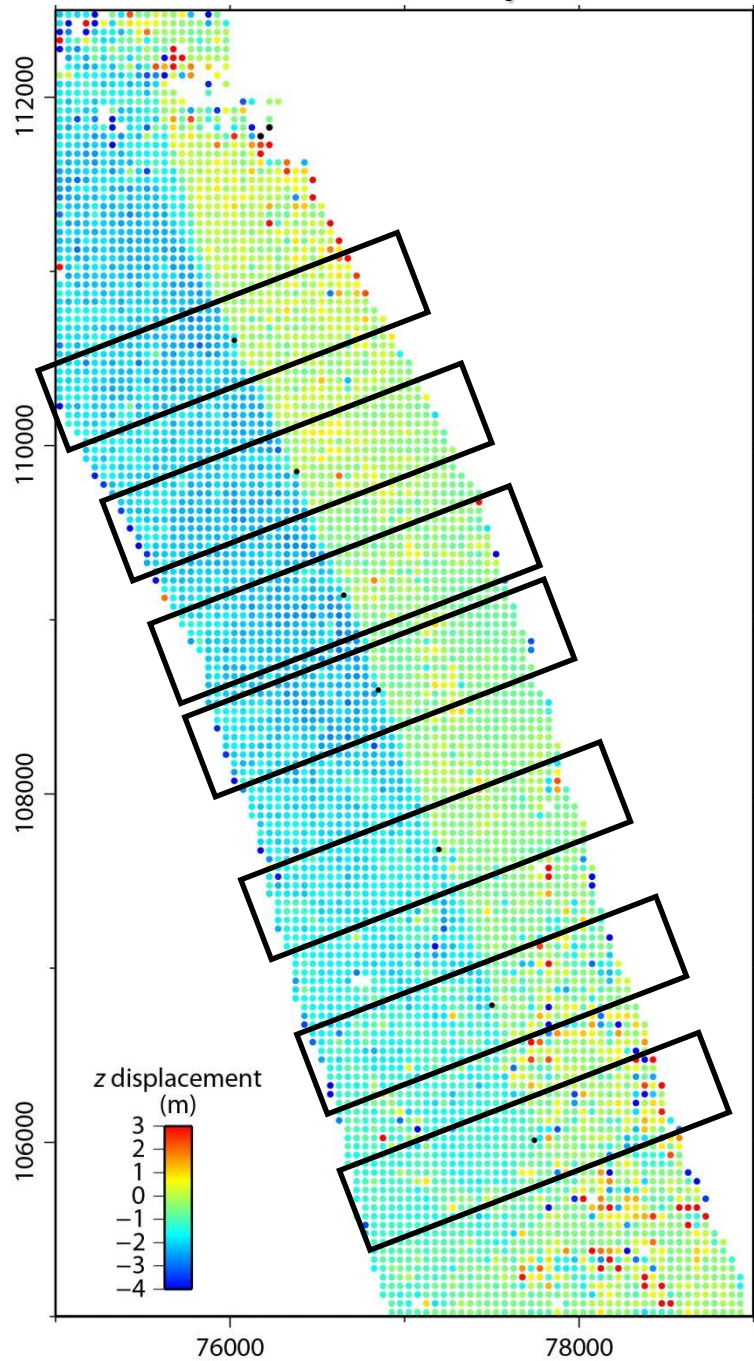
2011 post-event 1 m DEM  AERO ASAHI CORPORATION

Tadashi Maruyama (AIST)

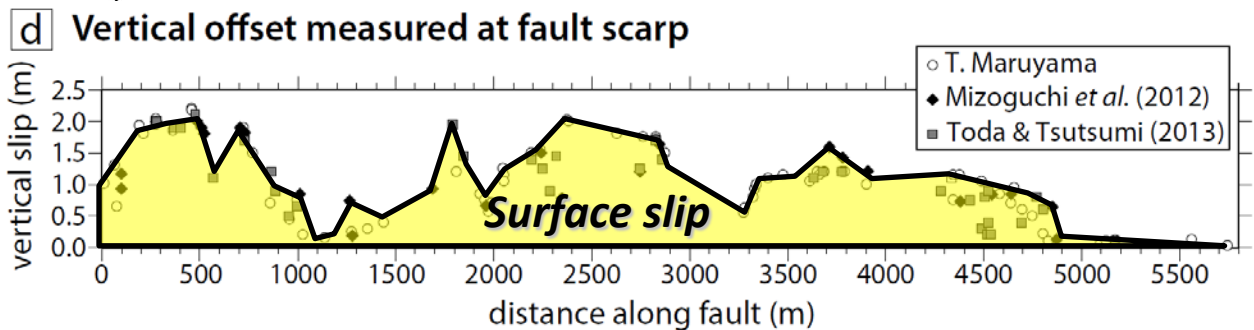
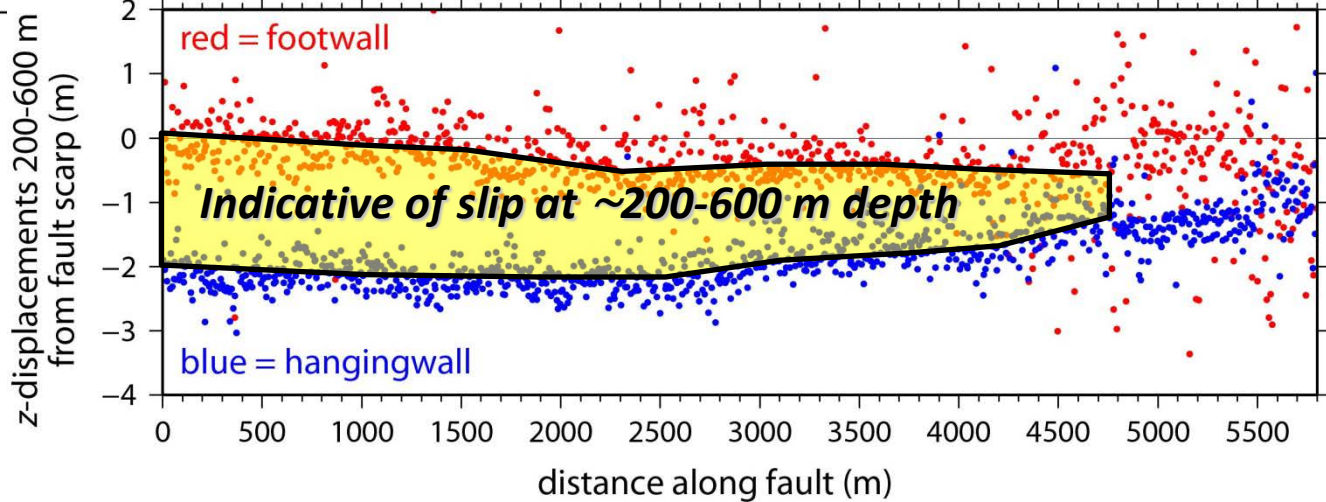
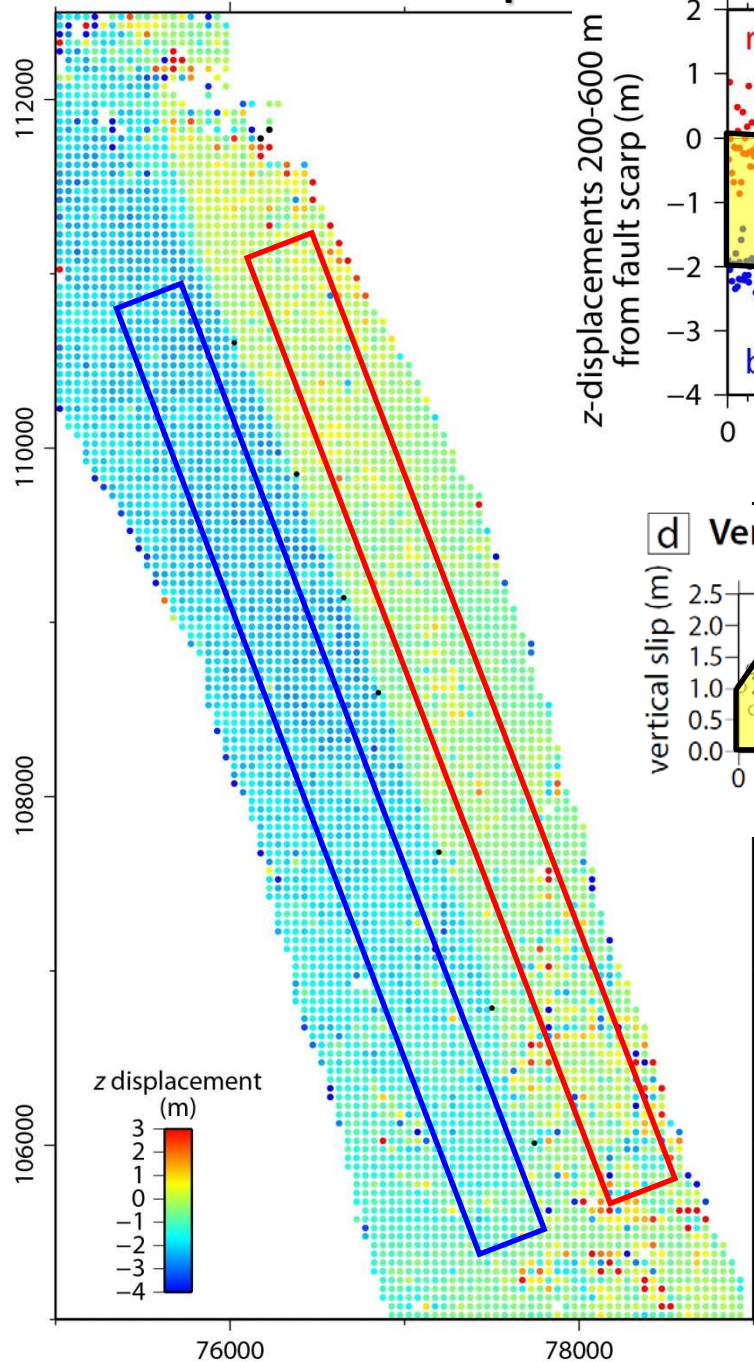
11 April 2011 Fukushima-Hamadori earthquake



2005-2011 vertical displacements

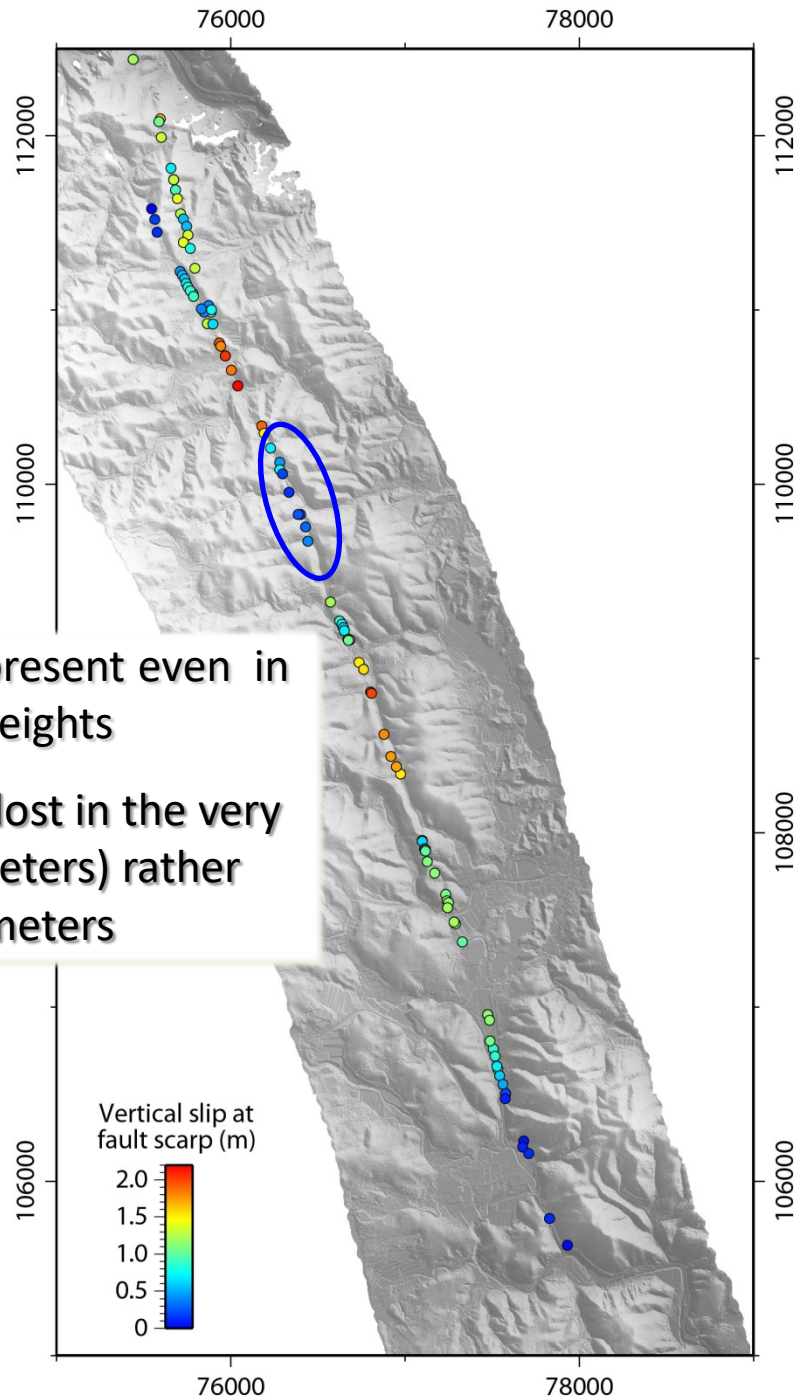
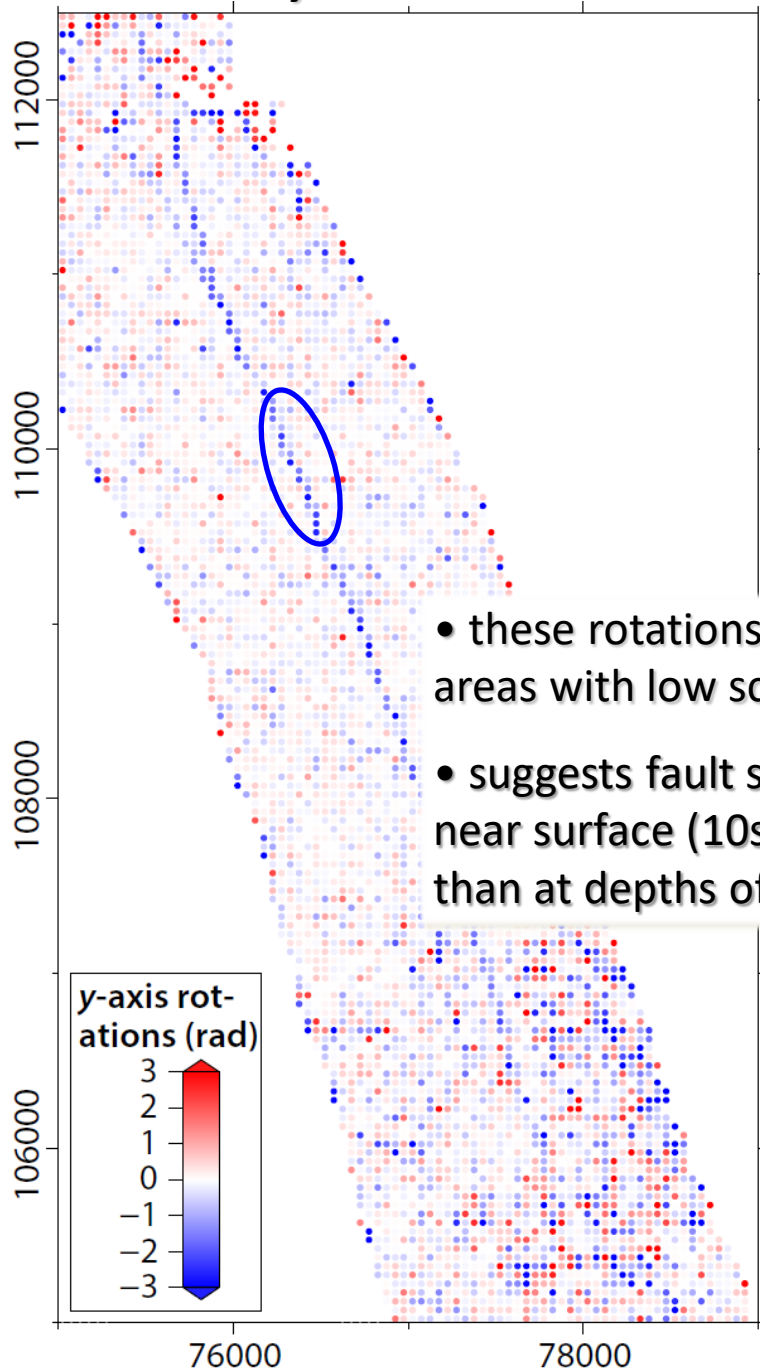


2005-2011 vertical displacements



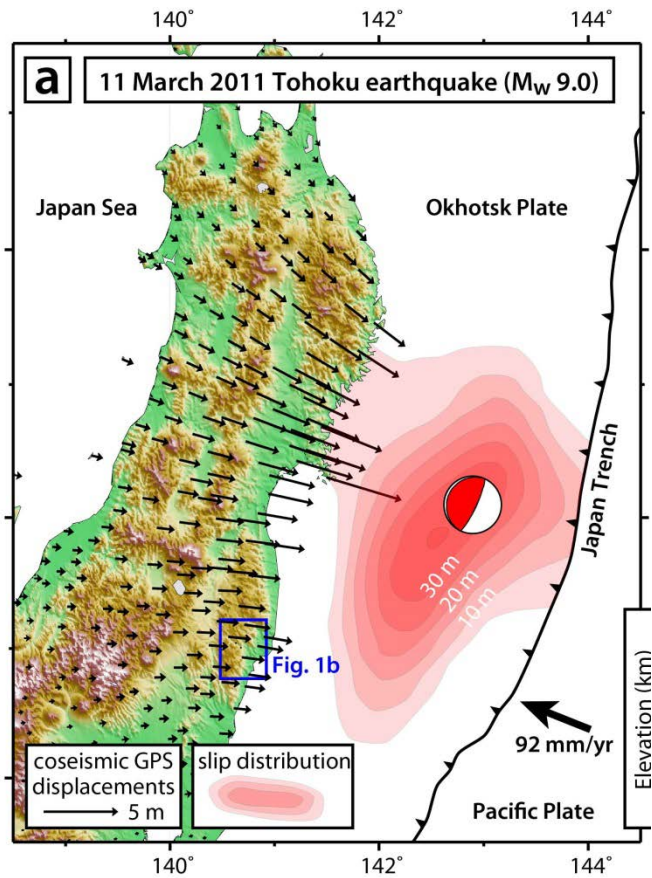
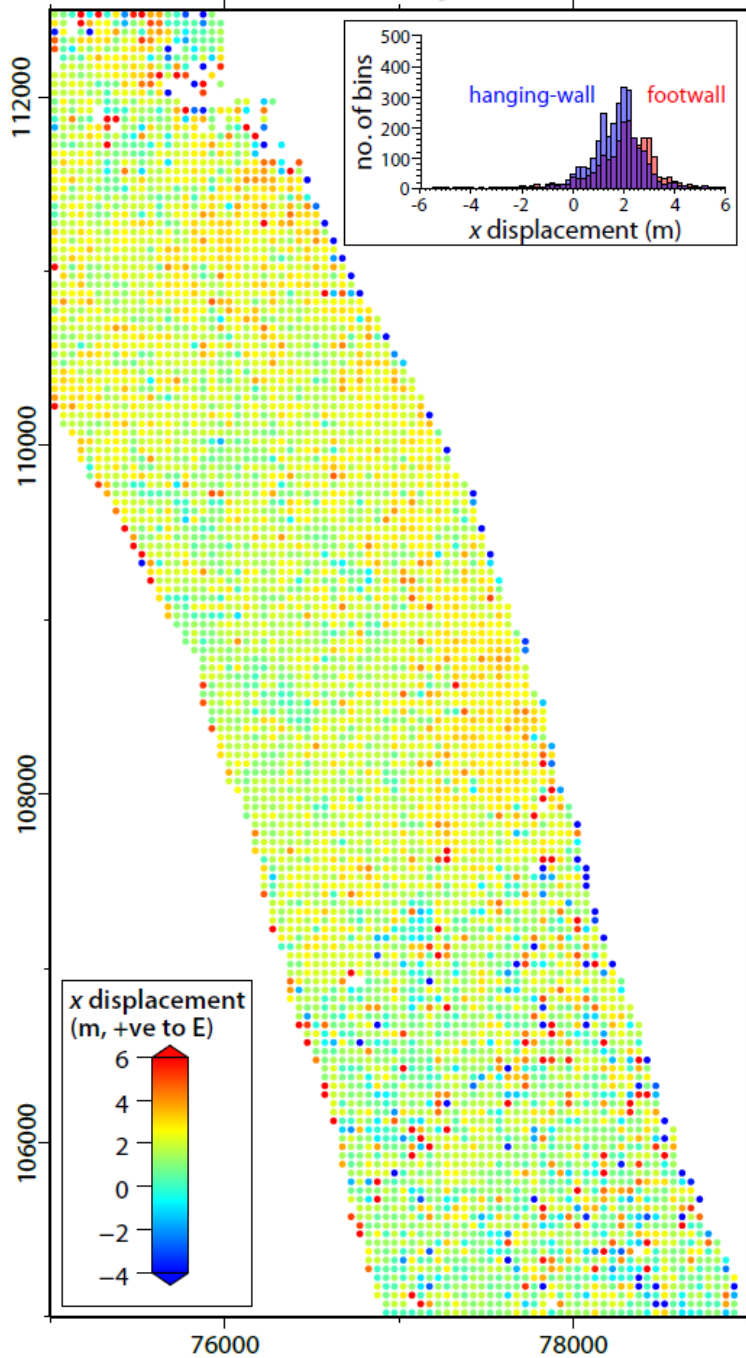
- In many places, only a small proportion of the slip makes it to the surface
- However, slip at depths of a few hundred meters appears to vary smoothly

2005-2011 y-axis rotations



- these rotations are present even in areas with low scarp heights
- suggests fault slip is lost in the very near surface (10s of meters) rather than at depths of kilometers

2005-2011 E-W displacements



Degradation of the 2010 El Mayor-Cucapah earthquake scarp

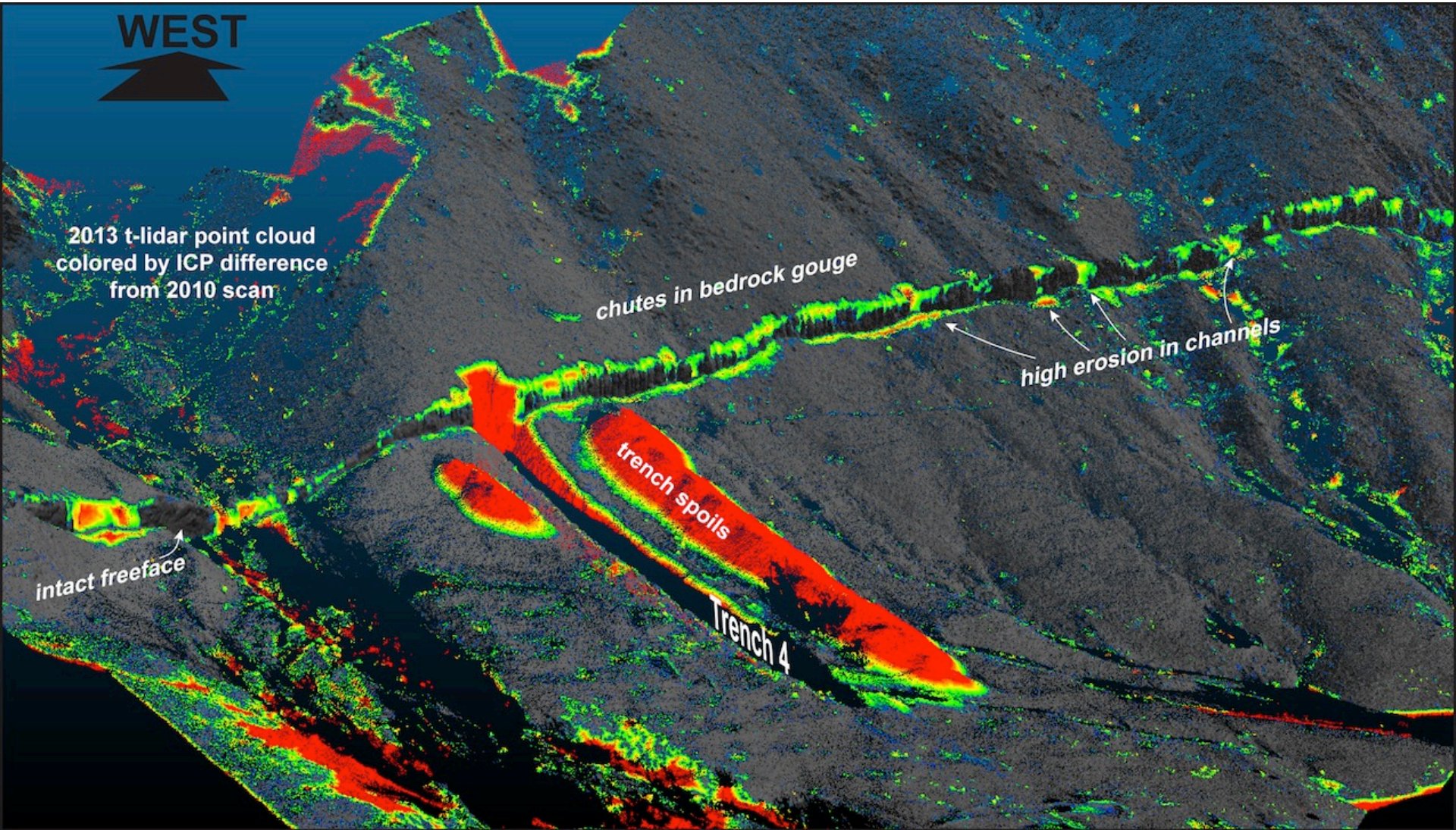
April, 2010



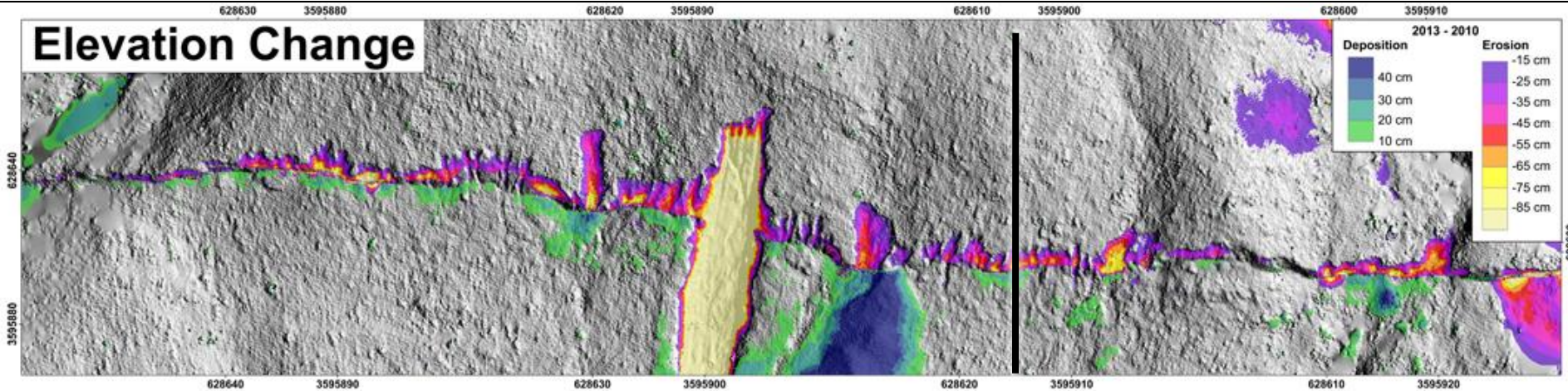
November, 2013



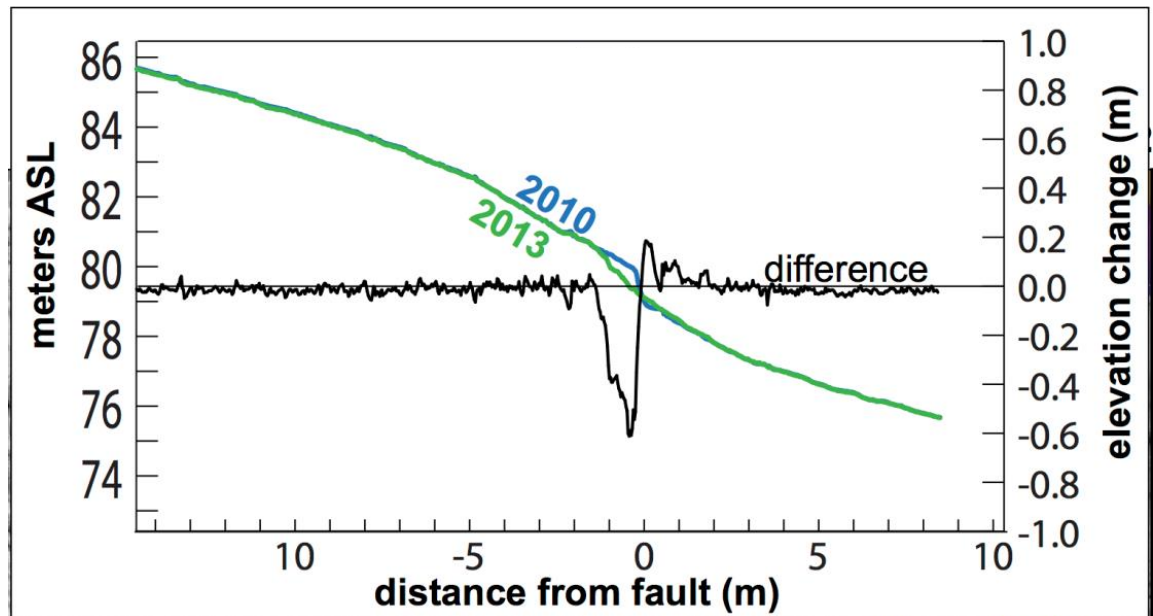
Degradation of the 2010 El Mayor-Cucapah earthquake scarp



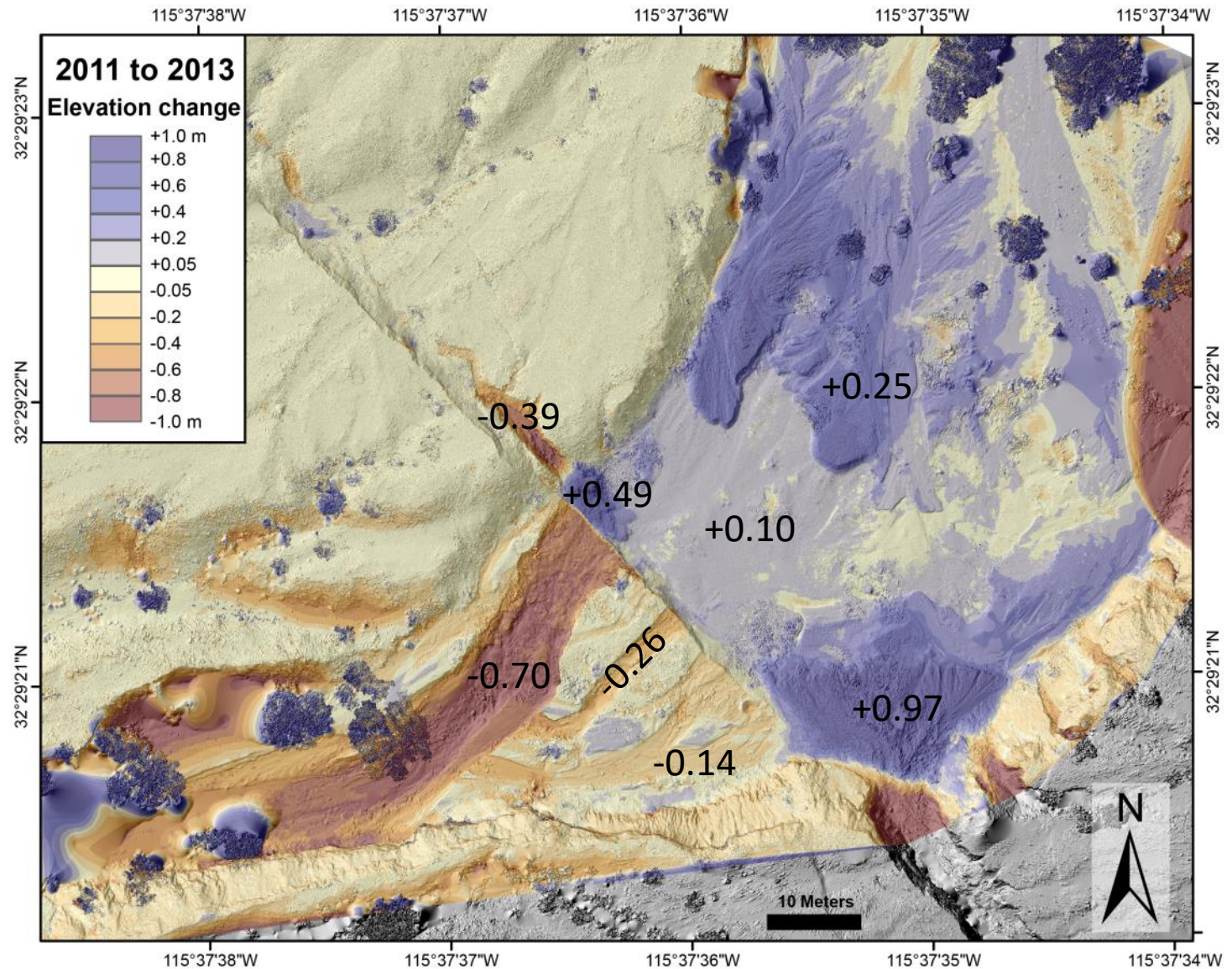
Degradation of the 2010 El Mayor-Cucapah earthquake scarp



10 m



Degradation of the 2010 El Mayor-Cucapah earthquake scarp



Aligning point clouds and topographic change detection

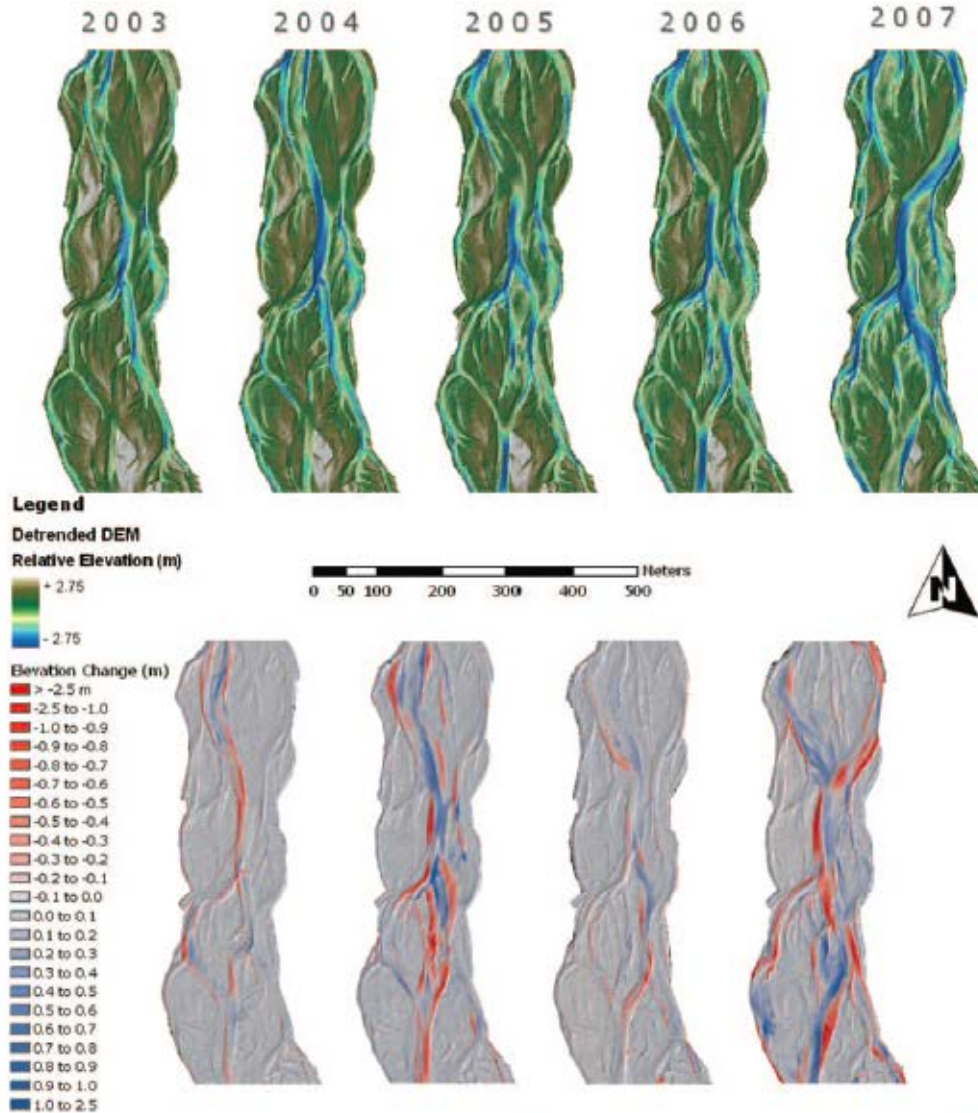


Figure 2. Detrended DEMs and DoD for 2003 to 2007. Note that the hillshades from the more recent year in the DoD are shown behind the DoD for context. This figure is available in colour online at www.interscience.wiley.com/journal/esp

RTK dGPS surveys tied to base stations occupying the same known point.

Point clouds are in exactly the same reference frame from the start.

DEMs are generated and pixel values subtracted: “DEM of Difference” of DOD

Wheaton *et al.* (2010), Accounting for uncertainty in DEMs from repeat topographic surveys: improved sediment budgets, *Earth Surface Processes and Landforms*

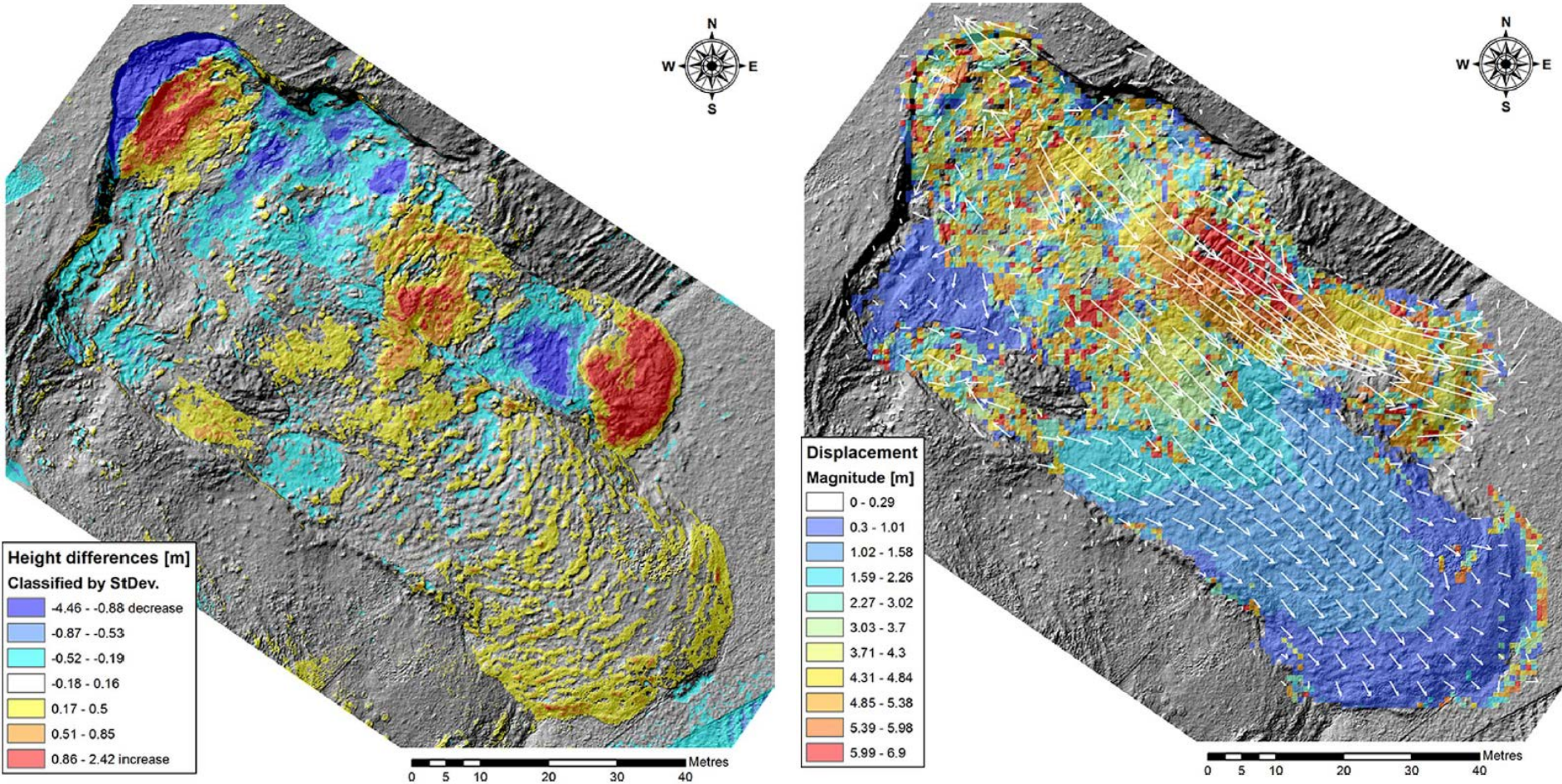
Aligning point clouds and topographic change detection

Repeat SfM surveys tied to ground control points surveyed with real-time kinematic GPS (2 – 4 cm accuracy).



Lucieer *et al.* (2015), Mapping landslide displacements using Structure from Motion (SfM) and image correlation of multi-temporal UAV photography, *Progress in Physical Geography*

Aligning point clouds and topographic change detection



- DEM of Difference (left) and horizontal displacement field from pixel cross-correlation (right)
- Caltech COSI-Corr package: Co-registration of optically-sensed images and correlation
http://www.tectonics.caltech.edu/slip_history/spot_coseis/index.html

Lucieer *et al.* (2015), Mapping landslide displacements using Structure from Motion (SfM) and image correlation of multi-temporal UAV photography, *Progress in Physical Geography*