

1 Impact of communal irrigation on the 2018 Palu earthquake-
2 triggered landslides

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30 **Anthropogenic changes to the environment can enhance earthquake-triggered landslides,**
31 **yet their role in earthquake disasters is often overlooked. Coseismic landslides frequently**
32 **involve liquefaction of granular materials, a process that reduces shear strength and**
33 **facilitates downslope motion even on gentle slopes. Irrigation systems can increase**
34 **liquefaction susceptibility and compromise otherwise stable slopes. Here we investigate**
35 **devastating landslides that affected Palu, Indonesia, during the 28th September 2018**
36 **M_w7.5 earthquake. We document fields and buildings translated over 1 km down slopes**
37 **of less than 2° and show landslides were limited to irrigated ground. A liquefied**
38 **detachment was rooted upslope in a conveyance canal that supplied water to the**
39 **irrigation network. A strong correlation between landslide displacement, irrigation**
40 **infrastructure and the highest slopes ($\geq 1.5^\circ$), suggests a causative mechanism that should**
41 **provoke urgent assessment of gently sloping irrigated terrain elsewhere in Sulawesi and in**
42 **tectonically active areas worldwide.**

43 While population growth and urbanisation expose more people to earthquake disasters
44 generally^{1,2}, anthropogenic modification of the environment can increase the risk of specific
45 earthquake hazards. Landslides, the most common secondary effect of earthquakes³, are
46 sensitive to anthropogenic changes to land-use^{4,5}, climate⁶, topography^{7,8} and hydrology⁹,

47 especially when coupled with strong seismicity^{10,11}. Understanding earthquake-triggered
48 landslides in modified environments is vital to ensure appropriate measures are taken during
49 reconstruction and to identify vulnerable areas elsewhere¹².

50 The Indonesian city of Palu, Central Sulawesi, has been identified as vulnerable to
51 earthquakes triggered by the Palu-Koro Fault, which passes immediately west of the urban
52 centre^{13,14,15,16,17} (Fig. 1a,b). Sited within a narrow valley facing an elongate bay, the city is
53 especially vulnerable to tsunami and landslides^{17,18,19,20}. On 28th September 2018, a M_w 7.5
54 earthquake centred 85 km north of Palu^{21,22} initiated a disaster that left 2,081 people dead,
55 1,309 missing and 206,494 displaced²³. Three main landslides, of 0.3-1.4 km² surface area
56 and 1.1 km maximum displacement, destroyed suburban areas on gently sloping alluvial fans:
57 Balaroa in the west, Petobo and Sidera in the east (Fig. 1c). The eastern landslides were tied
58 to a conveyance canal designed to enhance irrigated rice production in Palu valley^{24,25}, which
59 is dry (547-726 mm/a rainfall) compared to surrounding mountains (up to 3997 mm/a)¹⁸.

60 We present analysis of the eastern landslides and irrigation system, based on satellite
61 imagery. The imagery quality, dense human development and raft-like behaviour of landslide
62 blocks allow assessment of irrigation's role in Indonesia's most deadly coseismic landslides.
63 All significant (≥ 10 m displacement) landslides identified within Palu valley are marked in
64 Fig. 1c. The only landslide we do not consider here is Balaroa, because it was isolated, much
65 smaller than the eastern landslides (0.38 km²), adjacent to the surface rupture and not
66 associated with a major irrigation system (Supplementary Discussion 1.1).

67 **Landslide genesis and kinematics**

68 Three hours after a M_w 6.1 foreshock, the 28th September 2018 M_w 7.5 mainshock ruptured
69 >180 km southwards through Palu (Fig. 1b), in places at supershear velocity^{21,22}. Surface
70 offset peaked at about 8 m²¹ along the sinistral Palu-Koro Fault, which bisects Central
71 Sulawesi. The 500 km-long fault slips at 35-39 mm/yr^{13,14}, has a shallow locking depth¹⁴ and

72 clear geomorphic expression^{13,17,26}, particularly along a topographic scarp west of Palu¹³.
73 Significant earthquakes in the region (12th January 1927; 20th May 1938; 14th August 1968;
74 1st January 1996^{27,28}) caused strong ground motions and tsunamis^{19,29}, but there are no reports
75 of landslides similar to those of 2018.

76 Here we use very high resolution DigitalGlobe natural colour satellite imagery acquired on
77 20th February 2018 and on 2nd October 2018, four days after the earthquake (see Methods and
78 Supplementary Figs. 1-3), to evaluate the relationship between: (a) a lateral spread landslide
79 with >150 m displacement (Lolu landslide); (b) two lateral spread/translational landslides
80 with >1 km displacement and cogenetic debris flows (Petobo and Sidera landslides); (c)
81 zones of distributed lateral spreading; and (d) irrigation infrastructure (Fig. 2a, Methods and
82 Supplementary Fig. 4). Landslide fractures were interpreted and objects recognisable in pre-
83 and post-earthquake imagery tracked to map displacement (Fig. 2b-d, Methods and
84 Supplementary Figs. 5-7). We assume tracked objects record landslide motions not
85 superficial liquefied soil flow because of the consistent displacement field and relationship to
86 landslide fractures. Mapped fractures lie 7-8 km east of the surface rupture, and show no
87 signs of being tectonic in origin. Displacement was downslope, orthogonal to tectonic
88 motion, and interpreted to have been gravity driven. Precipitation before the earthquake was
89 average to low (see Supplementary Discussion 1.2).

90 Lolu landslide (Fig. 3a) exemplifies a simple lateral spread above a weak detachment^{30,31}.
91 Maximum displacement is 157 m. An arcuate crown fault system formed in Lolu village at 73
92 m elevation, the landslide terminates in a plantation 760 m west at 63 m elevation, giving an
93 average slope of 0.75°. Above the crown, arcuate cracks extend east to the conveyance canal.
94 The crown is a curvilinear array of normal faults, with individual heaves of ~2 m. Discrete
95 hangingwall and footwall cutoffs suggest a coherent surface layer. Lateral margins are zones
96 of strike-slip (Fig. 2b), imbricate thrusts characterise the toe. Light-coloured lobes along

97 fractures may be liquefied sediment escaping from the basal detachment, and ponded water
98 and sand blows in adjacent fields suggest widespread liquefaction (Supplementary Figs. 4
99 and 8).

100 Petobo landslide (Fig. 3b) is kinematically similar to Lolu. It descends a 1.2° slope from 83 m
101 to 38 m elevation along its 2,190 m length. A scalloped crown preserves individual fault
102 heaves up to 20 m (Fig. 2c). Maximum horizontal translation is 1,096 m – displacement
103 increases systematically from crown and toe to a well-defined peak, lending confidence to
104 our observed maximum (Supplementary Figs. 6 and 7). The landslide's centre, stripped of
105 recognisable objects, represents the exhumed basal detachment. The toe thrust (Fig. 2d) is
106 pinned in an urban area, beyond which there is negligible strain. Cogenetic debris flows mark
107 the southern margin and sand lobes mark the northern strike-slip margin.

108 Sidera landslide (Supplementary Fig. 4), similar to Lolu and Petobo upslope, has evolved into
109 a debris flow downslope where it merges with Paneki River. However, a partially preserved
110 thrust belt at 57 m elevation, 26 m below and 1,950 m WSW of the crown, suggests a
111 landslide slope of 0.76° . Maximum displacement of a recognisable object is 1,099 m
112 (Supplementary Fig. 7), though that object may have been partly moved by entrainment in the
113 debris flow.

114 Lolu, Petobo and Sidera landslides define peaks in a map of displacement based on tracked
115 objects (Fig. 4a and Methods). Additional significant ground displacement around Lolu is
116 expressed by high displacement lobes and nested arcuate cracks. South of Sidera, distributed
117 displacement is associated with widely-spaced fractures and crown cracks. Water ponding
118 and sand blows in both areas suggests extensive liquefaction (Supplementary Fig. 4).

119 **Irrigation infrastructure**

120 There is a striking spatial coincidence between the landslides and the irrigation system that
121 supplies Palu valley (Fig. 2, Methods and Supplementary Figs. 9 and 10). Irrigation water is

122 carried northwards along a 15-25 m wide conveyance canal that traverses west-dipping
123 alluvial fans south of Palu. Near Palu airport the canal turns 90° and narrows as it enters the
124 city. Six 3 m wide secondary canals and many smaller channels flow downslope from >21
125 engineered turnouts along the damaged section of the canal. Water enters fields via
126 distribution channels fed by concrete division boxes or artisanal systems. The density of
127 irrigation infrastructure may be a proxy for shallow water infiltration. Vegetation is healthy in
128 the irrigated area and sparse upslope of the canal (Methods and Supplementary Fig. 11).
129 The conveyance canal forms the upslope limit of all significant landslide faults and surface
130 displacements (Fig. 4a), most crown faults are localised within or at most 125 m west of the
131 canal (Fig. 2 and Supplementary Figs. 4 and 10). Average turnout frequency decreases from
132 one per 210 m at Petobo, to one per 320 m at Lolu/Sidera, to just one per 440 m south of
133 Sidera. Additionally, Petobo landslide is localised where a secondary canal from the SE
134 terminates in a distribution system, and along the axis of the canal to the city centre (Fig. 3b).
135 Small fracture systems upslope of the conveyance canal near Biromaru and the airport are not
136 associated with significant surface displacement.
137 Lolu landslide's crown lies 620 m west of the conveyance canal; however, it nucleated along
138 a smaller channel within Lolu village at the termination of a 3 m wide secondary canal (Fig.
139 3a). The crown is delimited by irrigation turnouts and a concrete division box.

140 **Landslide controls**

141 Landslide activity was limited to irrigated terrain mostly sloping $\leq 1.5^\circ$. Slopes above the
142 irrigation network did not fail, despite being steeper, indicating irrigation water was a more
143 important control on landsliding than slope alone. However, steeper topography (1.5° - 4.0°)
144 close to the main conveyance canal at the top of Petobo and Sidera landslides, may have
145 precipitated localised failure where irrigation water was also present.

146 To evaluate landslide controls, we conducted correlation coefficient and principal component
147 analysis (PCA) using 12 parameters measured across the study area and in ten 0.25 x 8 km
148 swaths (Fig. 4b, Methods and Supplementary Figs. 11 and 12). Parameters included
149 displacement, slope, aspect, vegetation and distance from features such as the earthquake
150 epicentre, surface rupture, canal and irrigation nodes (e.g. turnouts). For landslide
151 displacement, the strongest correlations (0.34-0.41) are displacement azimuth and distances
152 to the canal and irrigation nodes. These correlations rise to 0.66-0.94 in swath 3 (Petobo
153 landslide) and 0.71-0.73 in swath 6 (Lolu landslide). In all landslide swathes, irrigation
154 parameters (distance to canal, nodes and other channels) correlate strongly with ground
155 displacement (Fig. 4c, Supplementary Tables 1-11). Slope is the next most correlated
156 parameter. PCA reveals the first principal component (PC1) contains 62.6% of variance,
157 dominated by irrigation parameters. PC2 (15.8% of variance) is dominated by geographic
158 parameters (distance to Palu River, surface rupture). PC3 (6.1% of variance) is dominated by
159 landslide parameters (displacement amount, azimuth). A composite RGB image of the most
160 correlated parameters plus maps highlighting slope $\geq 1.5^\circ$ and areas >250 m from irrigation
161 channels (Fig. 4b) shows bright regions (A,B,C) representing coincident correlation peaks
162 within the main landslides. Only one site of high displacement (D) lies outside the strongly
163 correlated area. Here, landsliding may be in response to initial failure at A, supported by
164 curvature of the crown faults and WSW-directed displacement vectors in this region.

165 Landslides on gentle slopes can be caused by cyclic shearing-induced liquefaction of water
166 saturated, unconsolidated granular materials^{32,33,34}. Liquefaction increases pore fluid pressure
167 such that the material loses shear strength and cannot support topography. Young, layered,
168 unconsolidated granular sediments below the water table are most susceptible to
169 liquefaction³⁵. Eastern Palu valley is underlain by Quaternary alluvial fans³⁶ sourced from
170 nearby crystalline rocks³⁷. Where the water table is locally elevated by irrigation, the alluvial

171 fans satisfy all conditions for liquefaction. Liquefaction can also be triggered by
172 compressional loading from upslope collapse³⁸ – a feedback that may explain higher
173 displacements below the steeper initiation points. Fluidised sand along landslide margins,
174 water ponding and sand blows (see Methods and Supplementary Figs. 4 and 8), indicate an
175 important role for coseismic liquefaction within the irrigated area only. Additionally, shallow
176 water infiltration can introduce mechanical heterogeneity, promoting gentle slope failure³⁹,
177 suggested by exposure of a stratiform basal detachment in Petobo and Sidera landslides.
178 Turnagain Heights (Anchorage) is the classic example of liquefaction-induced lateral spread,
179 formed during the 1964 M8.5 Alaskan earthquake. A 0.5 km² area experienced 610 m
180 displacement above liquefied sands⁴⁰. Contemporary accounts⁴⁰ suggest morphological
181 similarities to the Palu landslides, though at 1.4 km² surface area and 1096 m displacement,
182 the Petobo landslide alone may be the largest lateral spread ever documented.

183 **Implications for landslide mitigation**

184 We conclude that naturally metastable alluvial fans surrounding Palu city were compromised
185 by an elevated water table caused by water infiltration downslope of the conveyance canal
186 and irrigation infrastructure. Irrigation water enhanced mechanical heterogeneity and
187 liquefaction susceptibility, leading to catastrophic failure during the Mw7.5 earthquake. A
188 critical condition for the two very long run-out landslides was the coincidence of the
189 conveyance canal and topographic slopes $\geq 1.5^\circ$.

190 The area south of Sidera (location E, Fig. 4b) requires further investigation since, like Petobo
191 and Sidera, it is marked by widespread liquefaction and slopes $\geq 1.5^\circ$ adjacent to the canal, yet
192 surface displacement was small. Other areas of concern include the wider Lolu region; upper
193 parts of Petobo and Sidera landslides now with free surfaces against topographic depressions;
194 and irrigated areas of southern Palu valley.

195 Irrigation-linked landslides have been documented before^{9,41,42,43}. Loess collapse below a
196 canal during the 1989 Dushanbe earthquake, Tajikistan, was similar to the 2018 Palu
197 landslides in scale, topography, crown fault arrangement, exposure of detachment surfaces
198 and associated debris flows⁴⁴. Palu's landslides thus provide further warning that irrigation in
199 seismically active areas may critically undermine very gentle slopes, particularly in urban
200 settings. Irrigation systems are widespread in seismically active and increasingly urbanised
201 parts of Indonesia⁴⁵, Myanmar⁴⁶ and beyond⁴⁷. Mitigation strategies to prevent future similar
202 landslides might include: minimising water infiltration in areas of steepest slope by utilising
203 dendritic secondary canals originating from fewer turnouts; building tiered, parallel
204 conveyance canals to result in smaller, spaced perturbations of the water table; staggering
205 land use (agriculture, forestry, urban areas) so dry/root-stabilised soil punctuates water
206 saturated areas; and driving stabilising piles across strata of high liquefaction susceptibility in
207 particularly vulnerable urban areas.

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336 **Author contributions**

337 I. M. W. carried out the satellite image interpretation, wrote the manuscript and created the
338 figures. R. H. contributed to image interpretation, worked on image georeferencing,
339 processed the TanDEM elevation model and commented on the manuscript.

340 **Competing Financial Interests**

341 The authors declare no competing interests.

342 **Figure Legends**

343 **Fig. 1: Regional context and overview. a**, Location of Sulawesi, Indonesia. **b**, Central
344 Sulawesi, showing the Palu-Koro Fault, Palu city, the two largest earthquakes of 28th
345 September 2018^{48,49,50} and the M_w 7.5 surface rupture, after ref. ²¹. **c**, Overview of the
346 earthquake-triggered landslides (black areas) of the Palu valley, including Balaroa in the
347 west, and Petobo and Sidera in the east. The main rivers and irrigation channels are shown.
348 Base map is from 12.5 m TanDEM-X topographic data, 2018 surface rupture is drawn after
349 ref. ²¹.

350 **Fig. 2: Landslides and irrigation, eastern Palu valley. a**, Simplified interpretation of
351 landslide domains, landslide-related fractures, urban areas, irrigation channels, topographic
352 contours (from 12.5 m TanDEM-X). Location in Fig. 1c. See Methods for description and
353 Supplementary Fig. 4 for full interpretation. **b**, Northern strike-slip margin, Lolu landslide.
354 Housing estate (red roofs) was originally square in plan. **c**, Extensional domain of Petobo
355 landslide, detaching along the conveyance canal. **d**, Toe thrust system of Petobo landslide,
356 highlighting shortening and coherent thrust slices. Representative tracked objects circled in
357 yellow, displacement in metres. Locations in Figs. 2a. and 3. Imagery ©2018 Google and
358 DigitalGlobe.

359 **Fig. 3: Structural maps of two representative landslides. a**, Lolu landslide, showing pre-
360 earthquake irrigation infrastructure, roads, buildings and topographic contours. Post-

361 earthquake roads and building positions shown in darker grey. **b**, Petobo landslide, showing
362 pre-earthquake irrigation elements, urban areas and topographic contours. See Supplementary
363 Fig. 6 for individual building tracking in Petobo landslide. In both maps, representative
364 displacement vectors are shown at the same scale as the map. Map locations shown in Fig.
365 2a. Full vector dataset shown in Supplementary Fig. 7.

366 **Fig. 4: Landslide displacement and controlling parameters. a**, Ground displacement
367 (black dots: observation points) and irrigation infrastructure. **b**, The wider irrigated area
368 (extent in Fig. 1c), combining five highly correlated parameters: (i) RGB image of
369 displacement (red), distance from irrigation nodes (green) and distance downslope of
370 conveyance canal (blue); (ii) slopes $\geq 1.5^\circ$ (white), $< 1.5^\circ$ (black); (iii) regions >250 m from
371 irrigation channels/rivers (yellow). Strong correlations expressed by bright areas (A, B, C).
372 D: anomalously high displacement, E: anomalously low displacement. **c**, Absolute correlation
373 coefficients between displacement and 11 other parameters for ten 0.25 x 8 km east-west
374 swaths.

375 **Additional information**

376 Supplementary information is available for this paper.

377

378 Impact of communal irrigation on the 2018 Palu earthquake-
379 triggered landslides

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383

384 **Methods**

385 **Data sources.** This study is based on two very high resolution 8-band panchromatic natural
386 colour satellite images from DigitalGlobe⁵¹ that cover the study area before and after the 28th
387 September 2018 earthquake (Supplementary Fig. 1a). Images were acquired by the
388 WorldView-4 satellite, and have variable spatial resolution, expressed as ground sample
389 distance (GSD). Off-nadir angle (ON) is less than 24° in all scenes. The main pre-earthquake
390 image was acquired on 20th February 2018 (scene ID: 1030010078CD4A00, GSD: 0.56 m,
391 ON: 18.5°), and the post-earthquake image was acquired on 2nd October 2018 (scene ID:
392 1040010042376D00, GSD: 0.31 m, ON: 2.3°). Additional scenes acquired on 26th September
393 2017 (scene ID: 1030010073524100, GSD: 0.51 m, ON: 18.5°), 5th August 2017 (scene ID:
394 103001006F222600, GSD: 0.54 m, ON: 23.7°), and 26th February 2011 (source not known)
395 were used in places to help map the pre-earthquake irrigation network. DigitalGlobe imagery
396 was accessed using freely available Google Earth Pro software
397 (<https://www.google.com/earth/download>) during October-November 2018. Scenes were
398 viewed from directly above and terrain modelling was disabled to minimise interpreter errors
399 related to perception and intuitive responses⁵². In addition, Google Street View
400 (Supplementary Fig. 1b) was used to inspect pre-earthquake details of the irrigation
401 infrastructure adjacent to main roads and to help identify the pre-earthquake location of

402 translated buildings. General observations of the irrigation infrastructure and the geology of
403 the alluvial fans were made during fieldwork in 2009, 2010 and 2011. Pre-earthquake
404 vegetation/moisture indices were mapped using Sentinel 2B satellite imagery acquired on 27th
405 September 2018, in green/red/near infra-red bands 3 to 8 (central wavelengths 559 to 833
406 nm). Bands have 10-20 m spatial resolution⁵³. Sentinel data were processed using ArcGIS
407 software. We used topographic data from the German Aerospace Center's TanDEM-X
408 satellite, which flew alongside the TerraSAR-X satellite to produce a high resolution global
409 digital elevation model (DEM) by synthetic aperture interferometry⁵⁴. The DEM has 0.4
410 arcseconds (~12 m) horizontal resolution and 2 m vertical accuracy where slope is < 20%.
411 Full-resolution TanDEM-X data have been available for scientific use upon application since
412 2016⁵³, and a 90 m derivative has been freely available via the German Aerospace Centre
413 since 2018: (<https://geoservice.dlr.de/web/dataguide/tdm90/>). We were awarded use of four
414 ~12 m resolution 1°x1° tiles centred on 120°E, 1°S that cover Palu and Central Sulawesi.
415 Shaded relief images and topographic contours used in the figures were derived from the
416 TanDEM-X data. All geospatial data were processed in ArcGIS software and projected to a
417 UTM zone 50S coordinate system, WGS1984 datum and Transverse Mercator projection.

418 **Positional accuracy of interpreted images.** Horizontal positional accuracy in Google Earth
419 is often poorly constrained; estimates range from 39.7 m RMSE (root mean square error) in
420 2008⁵⁵ to 1.80 m in 2013⁵⁶ and ~1 m in 2015⁵⁷ for the natural colour imagery, and 2.64 m
421 RMSE for the terrain model⁵⁸. Digital Globe⁵¹ report <5 m positional accuracy for their
422 WorldView-4 satellite imagery. To test the positional accuracy of the two main images used
423 they were calibrated against an independently acquired Global Positioning System (GPS)
424 track (Supplementary Fig. 2). A track acquired using a Garmin Oregon 550T handheld GPS
425 receiver was selected because it crossed the study area, including regions of low deformation,
426 broadly from NW to SE. The track was acquired during fieldwork on 25th September 2010,

427 driving from north to south on the left side of the road from Palu to south of Sidera. From
428 points placed at 100 m intervals along the track, offset from the centreline of the left-hand
429 carriageway was measured in N-S and E-W directions. For the post-earthquake image,
430 calibration measurements were only made where landslide displacement was <3 m – in most
431 places along the GPS track landslide displacement was negligible. In all cases the GPS tracks
432 are a good fit to the imagery, considering inherent GPS positional uncertainty, plotting on or
433 very close to the road. For the 20th February 2018 scene, mean offset is 5.9 m east, standard
434 deviation (SD) is 2.14 m (n=96); and 3.94 m north, SD 1.13 m (n=32). For the 2nd October
435 2018 scene, mean offset is 0.28 m west, SD 1.46 m (n=60); and 0.28 m south, SD 0.68 m
436 (n=32) (Supplementary Fig. 3). This test shows that the post-earthquake imagery has a mean
437 positional accuracy with respect to the GPS network close to the image's GSD (0.31 m). The
438 pre-earthquake imagery is systematically offset to the NE by about 10 times its GSD.

439 For this study, positional consistency between satellite scenes is more important than high
440 accuracy on a global framework. The GPS track test highlights a systematic positional offset
441 of the pre-earthquake imagery accessed via Google Earth at the time of analysis, which is not
442 present in the post-earthquake imagery. To ensure landslide displacements are considered
443 within a fixed local reference frame, a grid of 12 tracked objects was picked on the upper
444 alluvial fan east of the irrigation canal, away from all observed landslide activity. These
445 objects were widely and evenly spaced, and selected because they could be tracked with
446 confidence in both pre- and post-earthquake satellite images and were far from any landslide-
447 related features. All 12 objects showed a systematic offset of the pre-earthquake image
448 towards the ENE with respect to the post-earthquake image (mean azimuth 073° , SD 4.45°).
449 Offset ranged from 3.46 m to 7.16 m (mean 5.45 m, SD 0.99 m) east; and from 0.82 m to
450 2.60 m (mean 1.64 m, SD 0.43 m) north (Supplementary Fig. 2e,f). These values are similar
451 to the independent GPS track offset, within SD (E-W) and slightly larger than SD (N-S).

452 To remove the systematic offset between landslide displacement vectors measured from the
453 pre- and post-earthquake images, a correction vector composed of the inverse mean offset
454 values from the 12-point reference frame, i.e. 5.45 m west and 1.64 m south, was applied to
455 the start point of all measured displacement vectors (Supplementary Fig. 3c). This correction
456 reduced displacement vectors in the stable eastern area and in the region downslope of the
457 landslides to <2 m, which is close to the accuracy of the object picks. Corrected vectors were
458 then plotted in ArcGIS and used to derive a displacement raster (Fig. 4a)

459 **Landslide-related structural mapping.** With the advent of very high resolution satellite and
460 aerial imagery, it is possible to map landslide-related fissures and other earthquake-related
461 structures remotely^{59,60}, a critical step towards understanding landslide evolution⁶¹.

462 Landslide-related fractures at the surface are dominated by tensile cracks because of the low
463 tensile strength of geomaterials, and can be recognised as dark lines in optical imagery⁶².

464 Dilational tension gashes, shear fractures and larger structures analogous to normal, thrust
465 and strike-slip faults^{63,64} were mapped from the post-earthquake scene (Supplementary Fig.
466 4). Fractures appear as dark or light lines ~10-100 m long, sometimes arranged into hard-
467 linked 'faults' ≤ 1.3 km long. They displace roads and boundaries and can be marked by light
468 patches interpreted as sand blows. En-echelon arrays, wing cracks, relay ramps and tip splays
469 help distinguish fractures from cultural linear features such as tracks. Comparison with pre-
470 earthquake imagery helped to confirm an earthquake origin for very small or poorly imaged
471 linear features. Structures were mapped as vector polylines directly in Google Earth, exported
472 to KMZ format and imported into ArcGIS software. Picking accuracy is $\pm 1-5$ m.

473 **Displacement measurement.** Finite displacement of the ground surface was measured by
474 tracking the position of 8,737 objects recognisable in both pre- and post-earthquake satellite
475 imagery (Supplementary Figs 5-7). Manual object tracking was employed because digital
476 image correlation⁶⁵ breaks down in areas of extreme damage in the post-earthquake

477 scene^{66,67}. Tracked objects included points with high X-Y positional certainty, mostly roof
478 features (e.g. ridge/hip/valley line intersections), wall vertices, small building centrepoints,
479 field boundary intersections, track/road intersections, small bridge abutments and if necessary
480 small isolated trees. Distinctive objects or associations of objects chosen were those that
481 could be recognised in both post- and several scenes of pre-earthquake imagery
482 (Supplementary Figs. 5 and 6). Objects at ground level were preferentially picked to
483 minimise parallax error related to satellite view angle. Tracked objects were picked with
484 about 30-100 m X-Y spacing, higher in regions of change and around fractures and faults.
485 Higher densities were also picked ~250 m either side of the main conveyance canal to
486 maximise certainty about the location of displacement onset. Lower densities resulted from
487 an absence of recognisable objects, for example in woodland and in the Sidera debris flow.
488 Accuracy of tracking points is in principle the same as the satellite image minimum ground
489 sample distance (0.31 m), but given issues of parallax, lighting variation and the vagary of
490 features such as footpaths or road margins, accuracy is probably closer to ± 1 m. Accuracy in
491 some areas of high displacement is locally worse, perhaps $\pm 5-10$ m, where collapsed
492 buildings can be identified only as associations of colour patches representing clusters of
493 building materials. In such cases the mid-point of buildings and colour patches was picked.
494 Lines linking the pre-earthquake position (start) and post-earthquake position (end) of each
495 tracked object were drawn directly in Google Earth, exported to KMZ format and imported
496 into ArcGIS software. For correlation coefficient and principal component analysis (Fig. 4b,c
497 and Supplementary Figs. 11-12), an additional 631 displacement measurements were made in
498 the wider study area. No significant ground displacement was observed beyond the focal
499 study area, so the additional data was gathered at a coarser spacing of about 250 m. These
500 data were processed as described above.

501 **Mapping of liquefaction features.** Liquefaction is commonly manifested as ejected
502 sediment and water, areas of ground distortion, fissures and settlement^{35,68,69}; these features
503 are visible in high resolution satellite imagery. Ground and aerial photograph-based studies
504 from the 2010-2011 Christchurch, New Zealand, earthquake sequence⁷⁰⁻⁷² were used as a
505 reference for identifying liquefaction effects in the post-earthquake imagery of Palu. Pounded,
506 ejected water appears as irregular, often rounded patches of uniformly dark colour usually
507 with a concentric fringe of lighter colour. Mottling in fields is often associated with ponding,
508 and may represent partially drained water ponds and patches of residual sediment or stressed
509 vegetation. Discrete sand blows from fractures appear as pale yellow patches or fan-like
510 flows. All liquefaction-related features were compared to pre-earthquake imagery
511 (Supplementary Fig. 8), mapped as vector points directly in Google Earth, exported to KMZ
512 format and imported into ArcGIS software for map production (Supplementary Fig. 4).

513 **Displacement raster.** Corrected displacement vectors based on traced objects were plotted as
514 points at the origin (pre-earthquake) X-Y coordinates. A natural neighbour triangular
515 irregular network (TIN) was produced based on values from the corrected displacement field,
516 and this was used to produce a 5 m resolution gridded raster surface representing
517 displacement (Fig. 4a). The raster's colourbar histogram was manually adjusted to emphasise
518 variation in displacement above the ± 1 m picking accuracy and below the 25 m threshold
519 where displacement becomes extreme.

520 **Slope and aspect rasters.** The TanDEM-X digital elevation model was used to derive
521 topographic slope and aspect rasters. Processing included filtering to remove off-terrain
522 objects (trees, buildings) using a search neighbourhood of 30 cells and a slope threshold of
523 1.5° , to produce a digital terrain model (DTM). The DTM was cleaned with a low-pass filter
524 to remove high frequency noise. For the slope raster, additional processing included
525 averaging each cell to the mean of a search radius of 2 cells (~24-25 m) to slightly smooth

526 anomalous high amplitude features (mainly cultural) remaining in the terrain without
527 reducing resolution. A slope raster was derived from that output. For Fig. 4b, a binary
528 colourbar was applied with a threshold of 1.5°.

529 **Irrigation network mapping.** The irrigation network (Supplementary Fig. 4) was mapped
530 using the post-earthquake image, plus pre-earthquake images, particularly scenes acquired on
531 20th February 2018 and 26th September 2017, to fill gaps where the land surface was badly
532 damaged. Positional accuracy is similar to that for structural picking (\pm 1-2 m) but may be
533 slightly worse because of minor, uncorrected positional differences between the scenes used.
534 The primary and all secondary conveyance canals, most significant distribution channels and
535 much of the water control and monitoring infrastructure was mapped (Supplementary Fig. 9),
536 but many smaller downslope features and channels were omitted because of their high
537 density and difficulty distinguishing them in satellite imagery. Turnouts can be recognised as
538 9x3 m open-sided buildings on the west side of the main canal that house valve controls and
539 culvert inlets which project into the canal. Downslope of the valve house, one or more 3 m
540 wide conveyance canals or substantial distribution channels carries water westwards. Canal
541 level checks are identified by slightly smaller open-sided buildings built adjacent to bridges
542 across the canal. Distribution boxes can be recognised where conveyance canals or main
543 distribution channels split into two or more distribution channels, and are marked by a small
544 open-sided building containing valve controls. Ground truthing of mapped irrigation features
545 was possible using road-level quasi-continuous photographic imagery accessible via Google
546 Street View, packaged within Google Earth (<https://www.google.com/earth/download>).
547 Google Street View imagery was acquired in the Palu region during December 2015, and
548 allowed close inspection of 6 turnouts, 4 checks, 3 division boxes and one sluice in the
549 landslide-affected area, as well as long sections of the main canal, distribution channels and
550 the Sapu river weir in the south. Example imagery is included in Supplementary Fig. 9.

551 Drainage lines and points were traced directly in Google Earth, exported to KMZ format and
552 imported into ArcGIS software.

553 **Spatial correlation and principal component analysis (PCA).** To evaluate the spatial
554 correlation between landslide displacement and 11 other parameters derived from the area
555 east of the Palu River, we used the PCA toolbox in ArcGIS. The toolbox produces covariance
556 and correlation matrices, eigenvectors and the principal components that show variation in
557 the dataset. PCA has been used to analyse interdependence in complex datasets⁷³, including
558 geospatial data⁷⁴. Each of the 12 parameters was normalised to a 12.5 m resolution raster
559 scaled 0-1. For PCA, the 9.8 x 4.6 km detailed study area was expanded to 15.7 x 8.5 km,
560 clipped in the west by the Palu River, which forms the ultimate downslope limit of the
561 earthquake-triggered landsliding. PCA was applied to the full composite raster and
562 individually to ten E-W swaths of 250 m width (Fig. 4b) that cross regions of high, low and
563 negligible landslide activity. The 12 parameters (see Supplementary Fig. 11) were: **1:**
564 **Displacement azimuth**, derived from the measured displacement vectors (000° - 359°) and
565 recalculated such that 270° had a maximum value of 1, 000° and 180° became 0.5, and 090°
566 became 0. A 12.5 m raster was generated from the interpolated data. **2: Displacement**
567 **amount**, calculated from the measured displacement vectors. To normalise, displacements of
568 <3 m were assigned a value of 0, to acknowledge the high measurement uncertainty in that
569 population. Higher displacements were assigned as follows: 3-<4 m (0.1); 4-<5 m (0.2); 5-<6
570 m (0.3); 6-<8 m (0.4); 8-<10 m (0.5); 10-<12 m (0.6); 12-<14 m (0.7); 14-<16 m (0.8); 16-
571 <18 m (0.9); >18 m (1.0). All displacements above 18 m resulted in surface breaks and
572 building destruction, so a value of 1 indicates onset of significant landsliding. A 12.5 m raster
573 was generated from the interpolated data. **3: Modified Soil-Adjusted Vegetation Index**,
574 (MSAVI2) is a standard measure of vegetation cover and health corrected for soil exposure,
575 using multispectral satellite imagery⁷⁵:

$$MSAVI2 = \frac{(2 \times NIR + 1 - \sqrt{(2 \times NIR + 1)^2 - 8 \times (NIR - RED)})}{2}$$

577 where NIR = near infra-red band, and RED = red band. MSAVI2 was normalised so that
578 1=healthy vegetation and 0= no vegetation. We used Sentinel 2B scene
579 S2B_MSIL1C_20180927T022319_N0206_R103_T50MRD_20180927T055542, acquired on
580 27th September 2018, the day before the Palu earthquake. Sentinel 2B bands 4 (red, 664.9 nm
581 wavelength) and 8 (near infra-red, 832.9 nm wavelength) have spatial resolutions of 10 m. **4:**
582 **Distance downslope of the conveyance canal**, used to generate a 12.5 m raster with values
583 from 1 (closest) to 0 (farthest). All points upslope of the canal were assigned values of 0. **5:**
584 **Distance from irrigation channels**, used to generate a 12.5 m raster with values from 1
585 (closest) to 0 (farthest). **6: Distance from earthquake epicentre** (0.256°S, 119.846°E) was
586 used to generate a 12.5 m raster with values from 1 (closest) to 0 (farthest). **7: Distance**
587 **downslope of major irrigation nodes** along the mapped irrigation system (e.g. turnouts,
588 division boxes) was used to generate a 12.5 m raster with values from 1 (closest) to 0
589 (farthest). All points upslope of the conveyance canal were assigned values of 0. **8: Distance**
590 **from the Palu River**, immediately west of the study area, was used to generate a 12.5 m
591 raster with values from 1 (closest) to 0 (farthest). **9: Distance from the 2018 surface**
592 **rupture**, digitised after ref. ²¹, was used to generate a 12.5 m raster with values from 1
593 (closest) to 0 (farthest). **10: Sentinel 2B bands 843**. Bands 8 (near infra-red, NIR), 4 (red)
594 and 3 (green) from Sentinel 2B scene
595 S2B_MSIL1C_20180927T022319_N0206_R103_T50MRD_20180927T055542 were
596 combined to generate a greyscale image, used to generate a 12.5 m raster with values from 1
597 (high NIR values) to 0 (high green values). **11: TanDEM-X topographic aspect**, the
598 processed digital terrain model was used to generate a 12.5 m aspect raster with values from
599 1 (south) to 0 (north). **12: TanDEM topographic slope**, the processed digital terrain model

600 was used to generate a 12.5 m slope raster with values from 1 (19°) to 0 (0°). Full results of
601 the correlation coefficient and principal component analysis are presented in Supplementary
602 Tables 1-11. Correlation coefficients between displacement and the other 11 parameters
603 within each of the ten swaths were used to produce the graph in Fig. 4c. Graphical
604 representations of principal components 1, 2 and 3 are presented in Supplementary Fig. 12,
605 alongside a composite RGB raster of all three.

606 **Data availability.**

607 Geospatial data generated during this project, including landslide displacement
608 measurements, liquefaction indicators and digitised irrigation infrastructure, are available at:

609 https://royalholloway.figshare.com/articles/Geospatial_Data/9205184. The post-earthquake
610 satellite scene is available via Google Earth:

611 <https://www.google.com/earth/download/gep/agree.html>, and via DigitalGlobe:

612 <https://discover.digitalglobe.com/>. Sentinel 2B data can be downloaded from the Copernicus

613 Open Access Hub: <https://scihub.copernicus.eu/dhus/#/home>. TanDEM-X 90 m data can be
614 downloaded via the German Aerospace Centre:

615 (<https://geoservice.dlr.de/web/dataguide/tdm90/>), and the application page for ~12 m data is:

616 <https://tandemx-science.dlr.de/cgi-bin/wcm.pl?page=TDM-Proposal-Submission-Procedure>.

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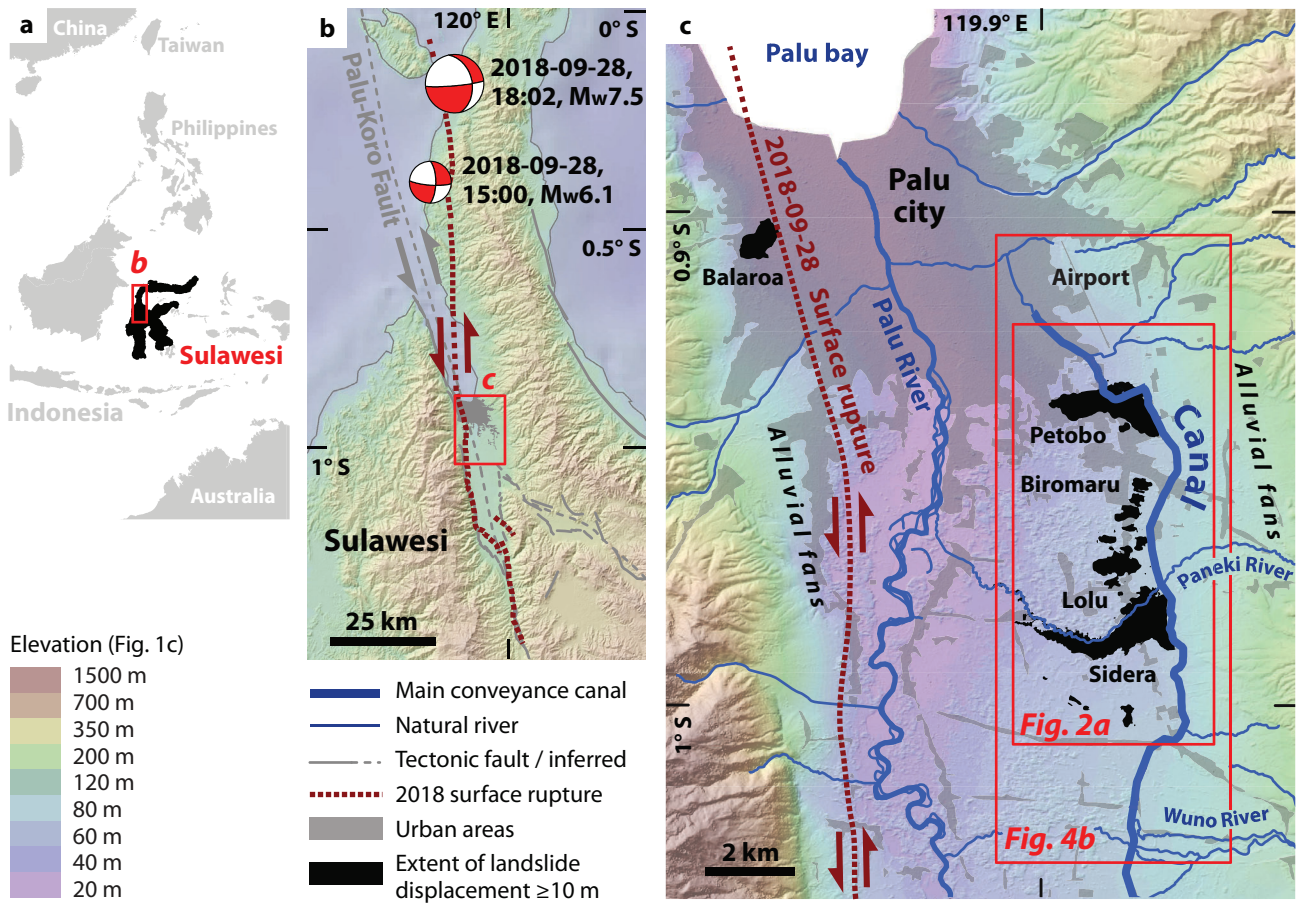
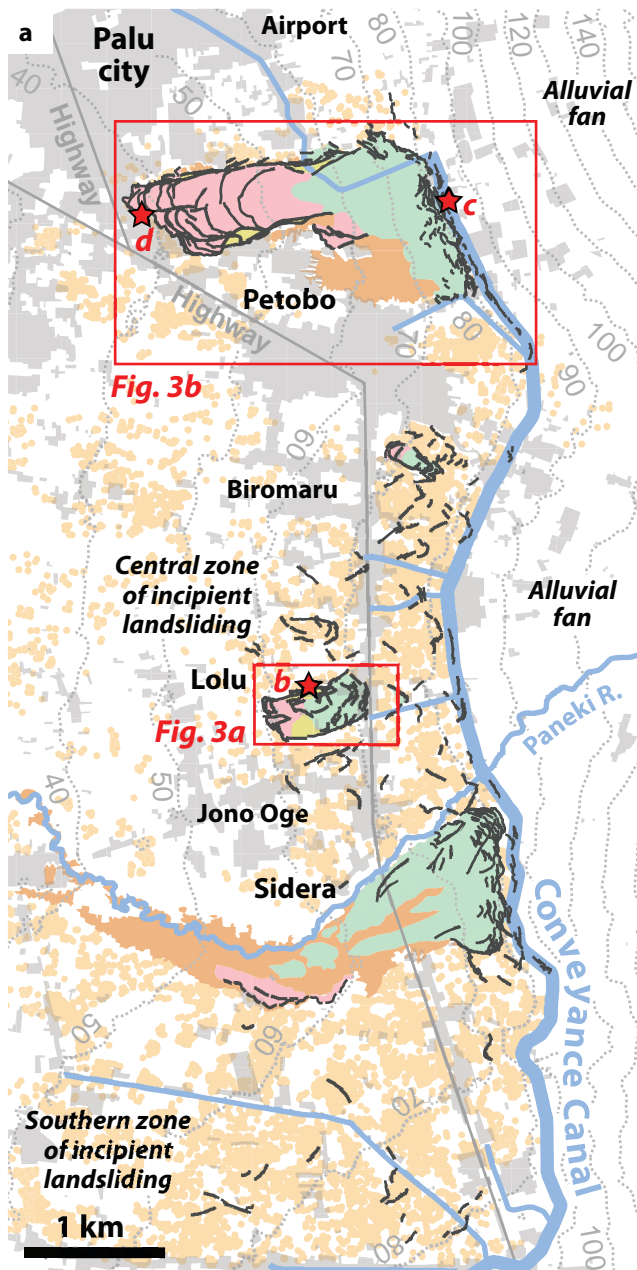
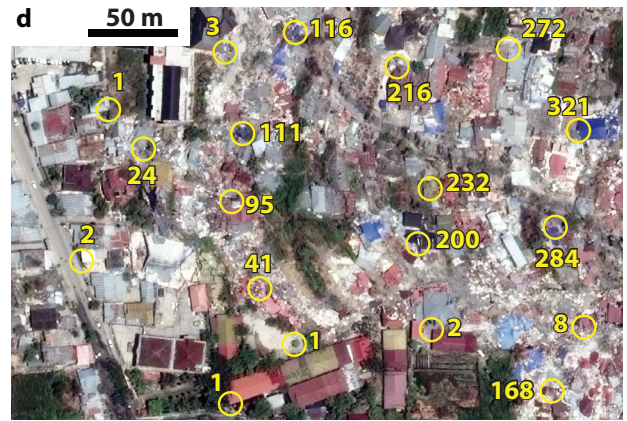


Figure 1



- Main conveyance canal, width ~ 15 m
- Secondary conveyance canal, width ~3 m / river
- Major landslide-related fault/fracture
- Main road
- ⋯ Topographic contour (elevation in m)



- Significant or extreme extension
- Dominant lateral translation
- Zone of intense thrusting
- Fluidised sediment flow
- Urban areas (pre-earthquake extent)
- Liquefaction indicators

Figure 2

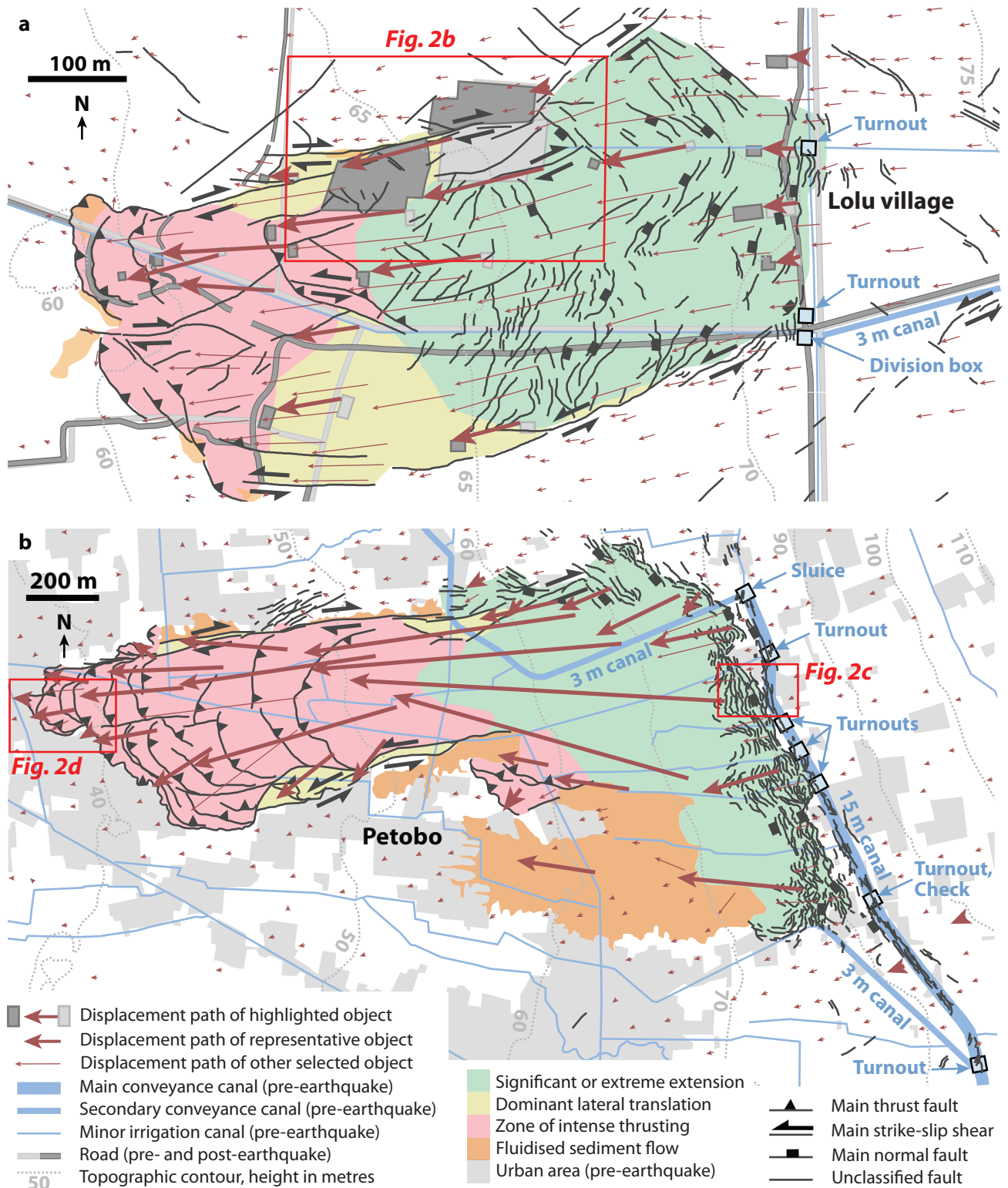


Figure 3

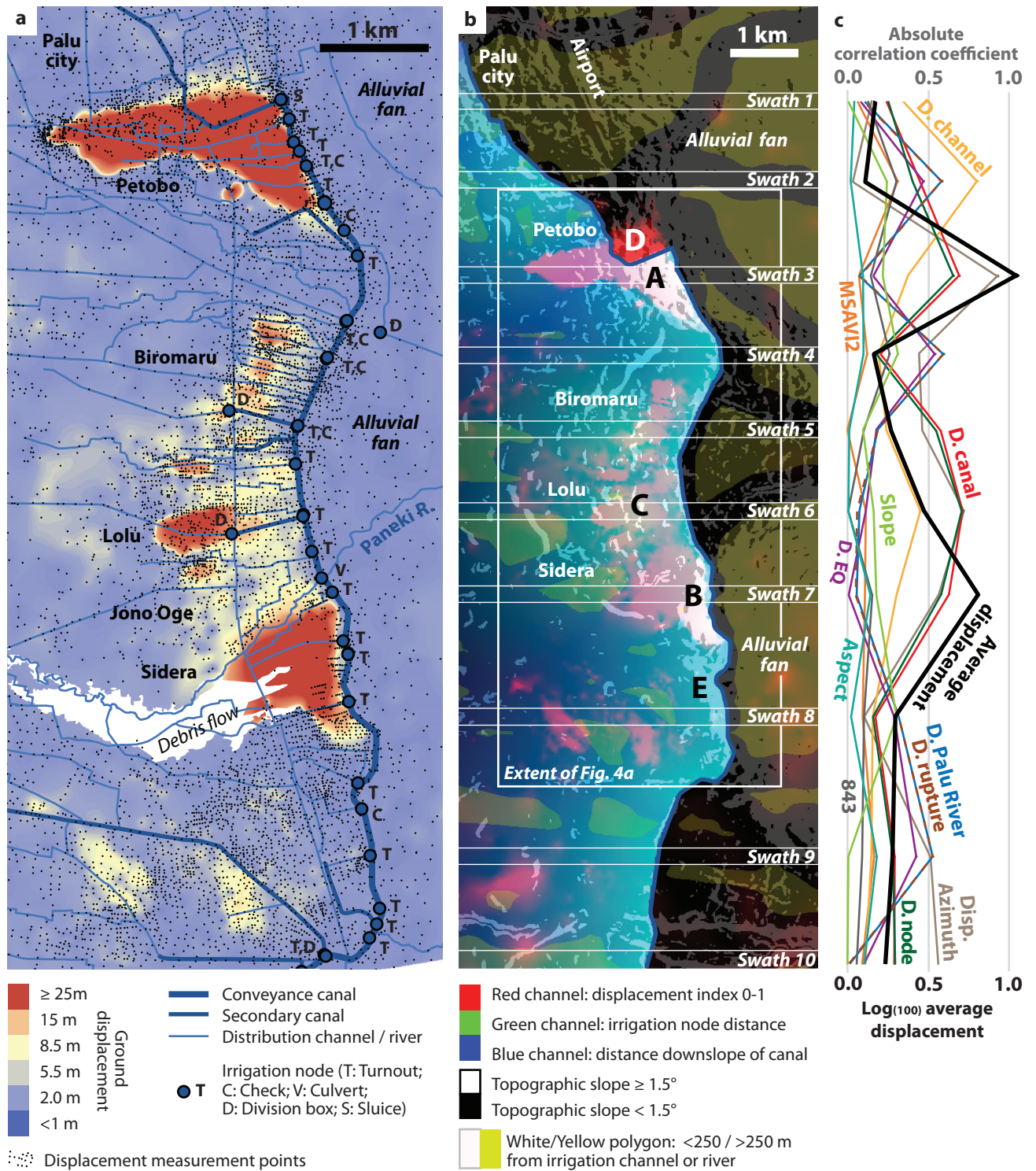


Figure 4