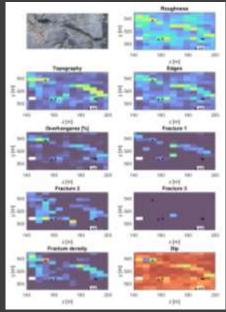
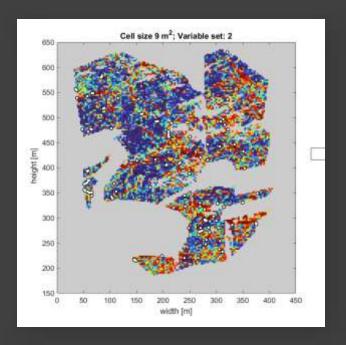
Spatial rockfall susceptibility prediction from rockwall surface classification

Alexander R. Beer¹, Nikolaus Krumrein¹, Sebastian G. Mutz¹, Gregor M. Rink¹, and Todd A. Ehlers¹

¹University of Tübingen, Tübingen, Germany (<u>alexander.beer@uni-tuebingen.de</u>)















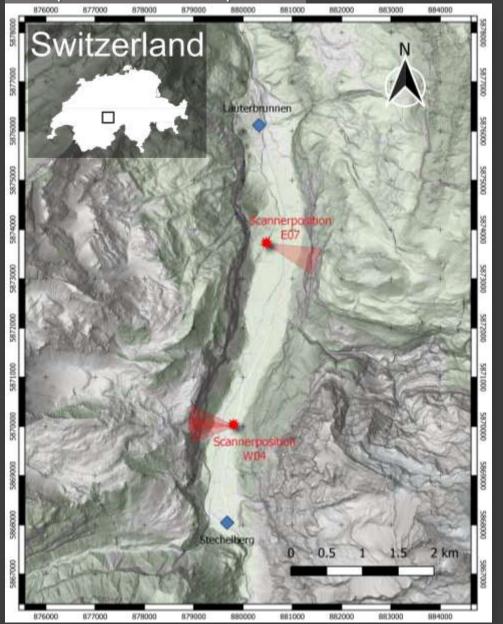
Background

- rockfall both is a major shaping and hazardous process in steep topography
- also abundant permafrost-free over-steepened rockwalls releasing rockfall
- spatial surveys used for fracture pattern, kinematic and rockfall event analyses
- > though, rarely used to predict local rockfall susceptibility vs. observed events

-> How does classification of high-res rockwall surface perform in predicting rockfall events?

TLS surveying

Examples of the scanner positions

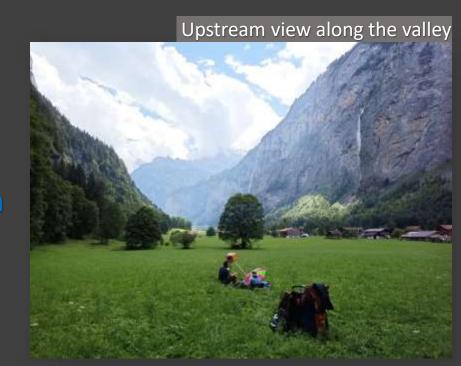


Field site

- Lauterbrunnen valley, Canton Bern, Switzerland
- 5km² of 800m distant, ~vertical limestone walls
- up to 1'000m high, variable fracture patterns

Field work

- TLS from 40 sites using ILRIS-LR
- 900 scans of ~5cmspatial resolution2014-2020

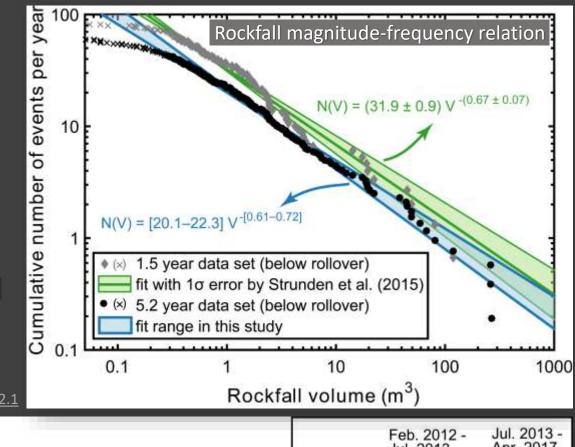


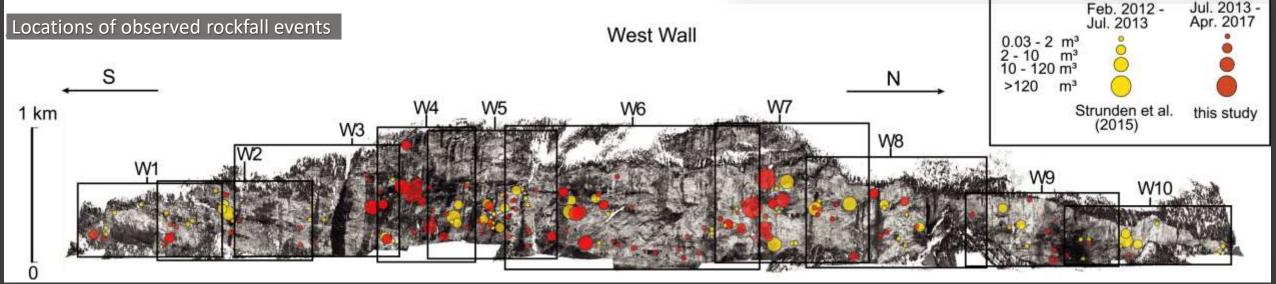
Previous work

TLS epochs change detection

- hand-cut to remove vegetation and fringes
- referenced by fixed points and ICP algorithm
- epoch change in the direction against the wall
- rockfall frequency and shape analyses

Mohadjer et al., 2020, Geology, https://doi.org/10.1130/G47092.1





Methods 1: Rockwall surface analysis

Surface parameters definition

- rasterized 3D data points (3², 5², 10², 15², 25², 40²m²)
- calculated several surface parameters per cell
- grouped them in 6 sets

Surface parameters, grouped in sets

name	entity	definition	variable set affiliation
Roughness	m	average mean distance to a 0.5 m circle around every point (Attachment 1) of a cell	A, B, C, D, E
Edge	1	average normal vector change rate (Attachment 2) of a cell	A, B, C, D, E
Topography	m	mean distance of all points within a cell to the best fitting plane	A, B, C, D, E
Overhangarea	%	area with less than 60 ° of the total surface	B, D, E
Fracture 1; Fracture 2; Fracture 3	m	mean distance between fractures	D, E, F
Fracture density	1/m²	number of fractures within a cell normalized to the cell size	D, E, F
Dip	degrees	average dip direction of a cell	C, E

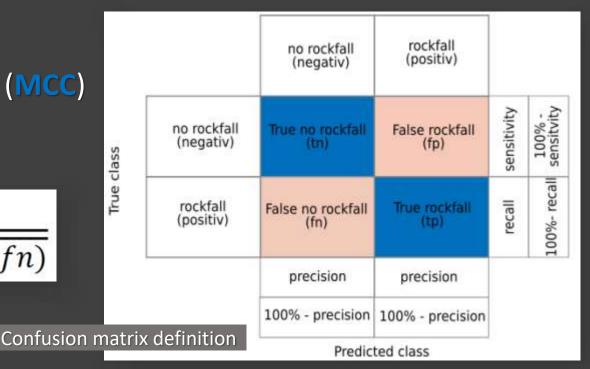
Methods 2: Rockwall surface classification

Bayesian classification procedure

- cells classified as rockfall vs. non-rockfall cells (based on 6a of TLS change-detection)
- implemented Naïve-Bayes-Classifier with 6 parameter sets and 9 variable combinations (distributions and probabilities, including misclassification cost)
- \succ trained on the 6 cell sizes = 324 models on one wall (T1) predicting rockfall susceptibility
- performance visualized by confusion matrix,
 quantified by Mathews correlation coefficient (MCC)

Mathews correlation coefficient

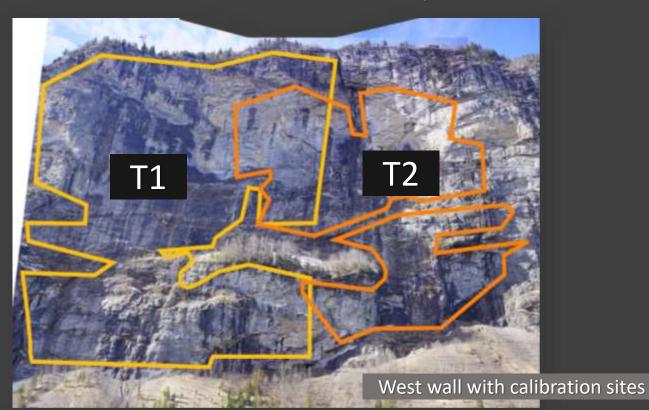
$$MCC = \frac{tp * tn - fp * fn}{\sqrt{(tp + fp) * (tp * fn) * (tn * fp) * (tn * fn)}}$$

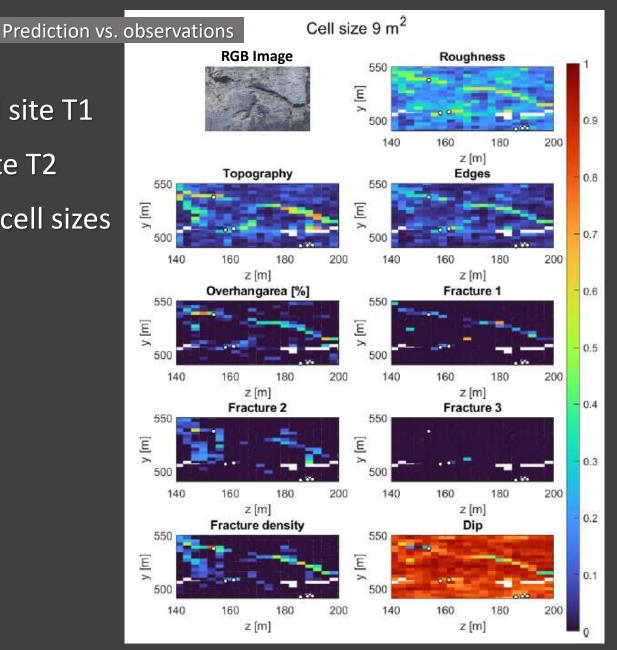


Results 1: Rockwall classification

Naïve-Bayes-Classifier application:

- trained on 6 cell sizes = 324 models on wall site T1
- tested for performance on adjacent wall site T2
- > structures best and well presented in 3²m² cell sizes

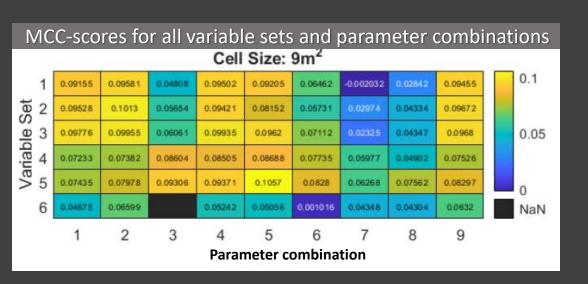




Results 2: Rockfall Classifier evaluation

Classifiers per cell size:

- generally more rockfall predicted than observed (but only 6a observations)
- hence small MCC-values



Confusion matrices of the best classifiers per cellsize Size 9 m2 MCC: 0.1057 Size 25 m² MCC: 0.1373 Size 100 m2 MCC: 0.1451 Variable-Parameter combination: A5 Variable-Parameter combiantion: B1 Variable-Parameter combiantion: C2 1.1% 38.9% Class SSE 2 76 46 42 47.7% 52.3% 26 48 2.6% 6.1% 13.0% 1.6% 4.9% 87.0% 2 2 Predicted Class Predicted Class Predicted Class Size 225 m2 MCC: 0.1798 Size 625 m2 MCC: 0.2511 Size 1600 m2 MCC: 0.3899 Variable-Parameter combiantion: D5 Variable-Parameter combiantion: E9 Variable-Parameter combiantion: F1 36.4% 63.63 30.8% 80.2 SSE 2 Class 53 0.0% 13.1% 50.0% 50.0% 42.2% 6.3% 15.2% 57.8% 51.4% 17.4% 79.8% Predicted Class Predicted Class Predicted Class

Rockfall susceptibility evaluation 1

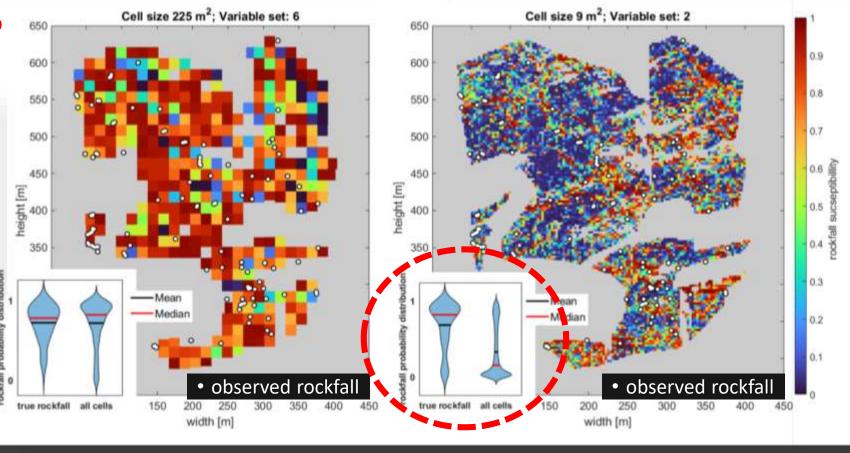
Check:

- best variable set: Roughness, Edges, Topography, and Overhang-Area (not fractures)
- best parameter combination: kernel density estimation, uniform probability

probability 0.73 vs 0.3



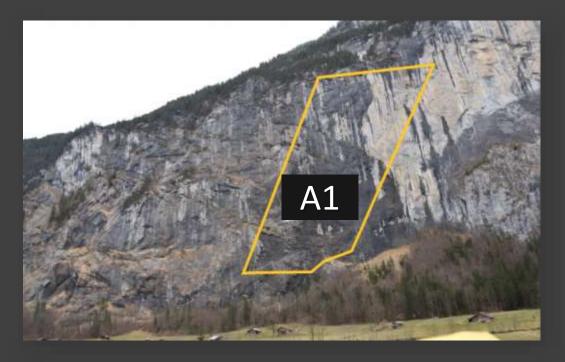
Application of the best classifier at wall T2 for different cell sizes



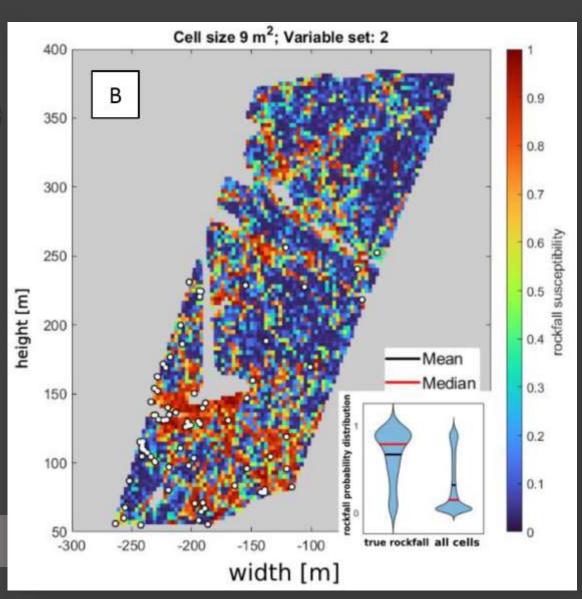
Rockfall susceptibility evaluation 2

Transfer:

apply method to different rockwall A1(other valley side, less fractured, smooth face)



Application of the best classifier at wall A1



Wrap-up

- NB classifier simple and fast for non-contact rockfall susceptibility mapping
- generally transferable (for $\leq 10^2$ m³), since only orientation-dependent
- small cellsize better; fracture sets not useful in cell-based approach

- fractures and overhangs well detected
- > next: also predict rockfall types

