



EXPLORING EFFECTS OF TRAINING SET SELECTION ON THE SHALLOW LANDSLIDE SUSCEPTIBILITY MODELING AT REGIONAL SCALE USING MAXENT MACHINE LEARNING TECHNIQUE

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OBJECTIVES

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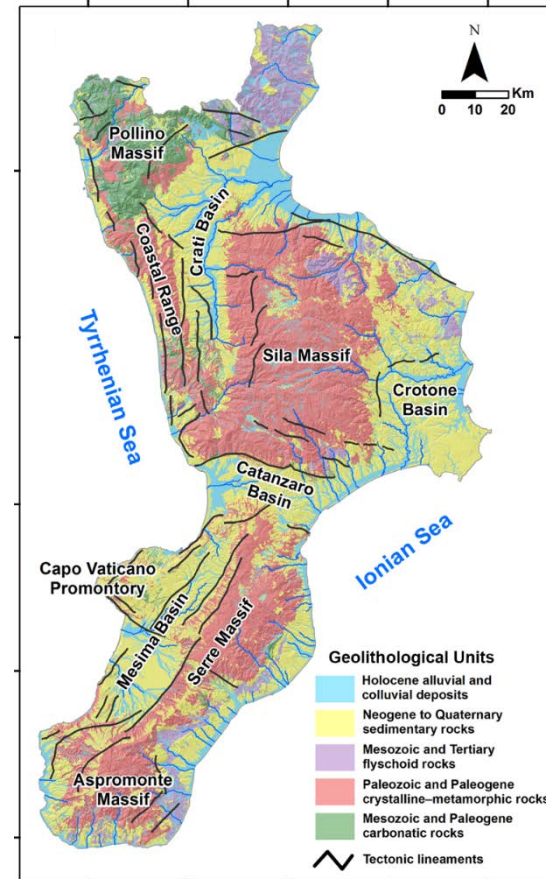
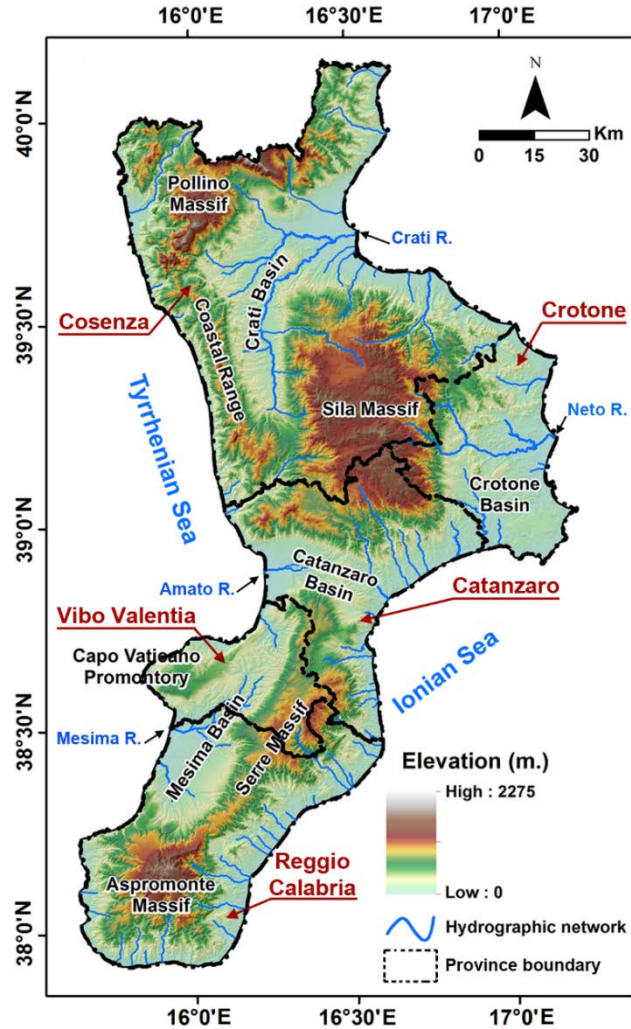
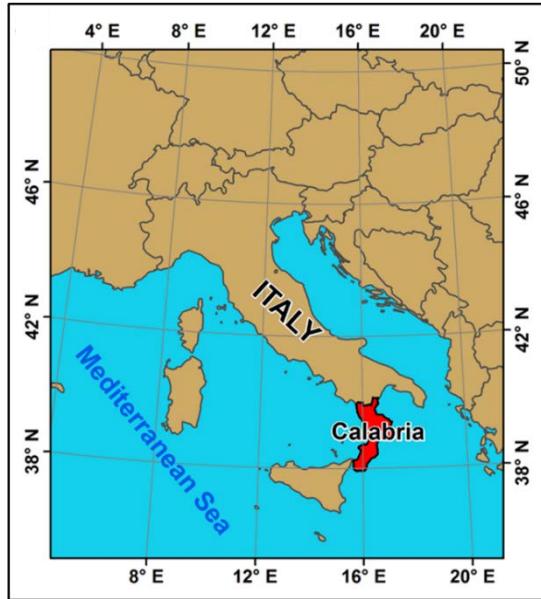
The aims of this study are:

- to evaluate the spatial prediction of shallow landslides at regional scale using Maximum Entropy (MAXENT) method;
- to evaluate training landslides random selection effects on the shallow landslide susceptibility modelling;
- to understand which are the main predisposing factors that control the spatial prediction of shallow landslides.

STUDY AREA

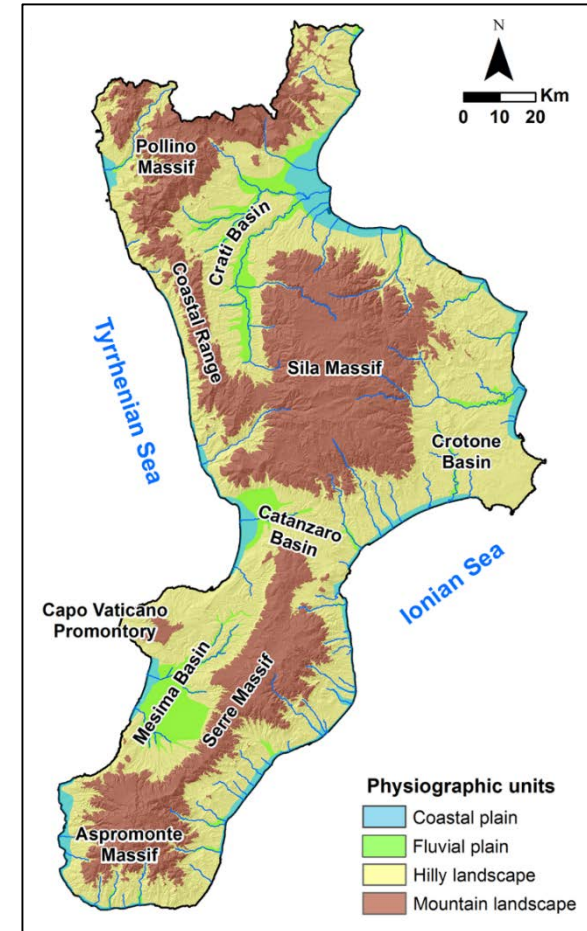
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Calabria Region



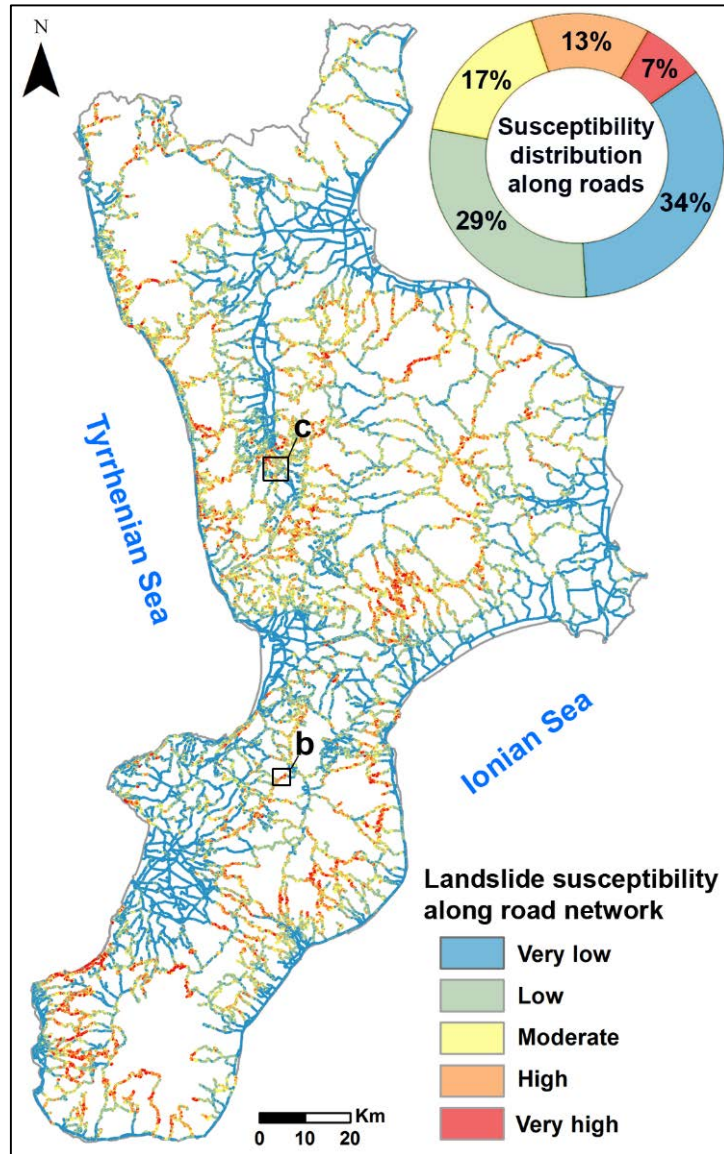
Schematic geological map

Physiographic units map



STUDY AREA

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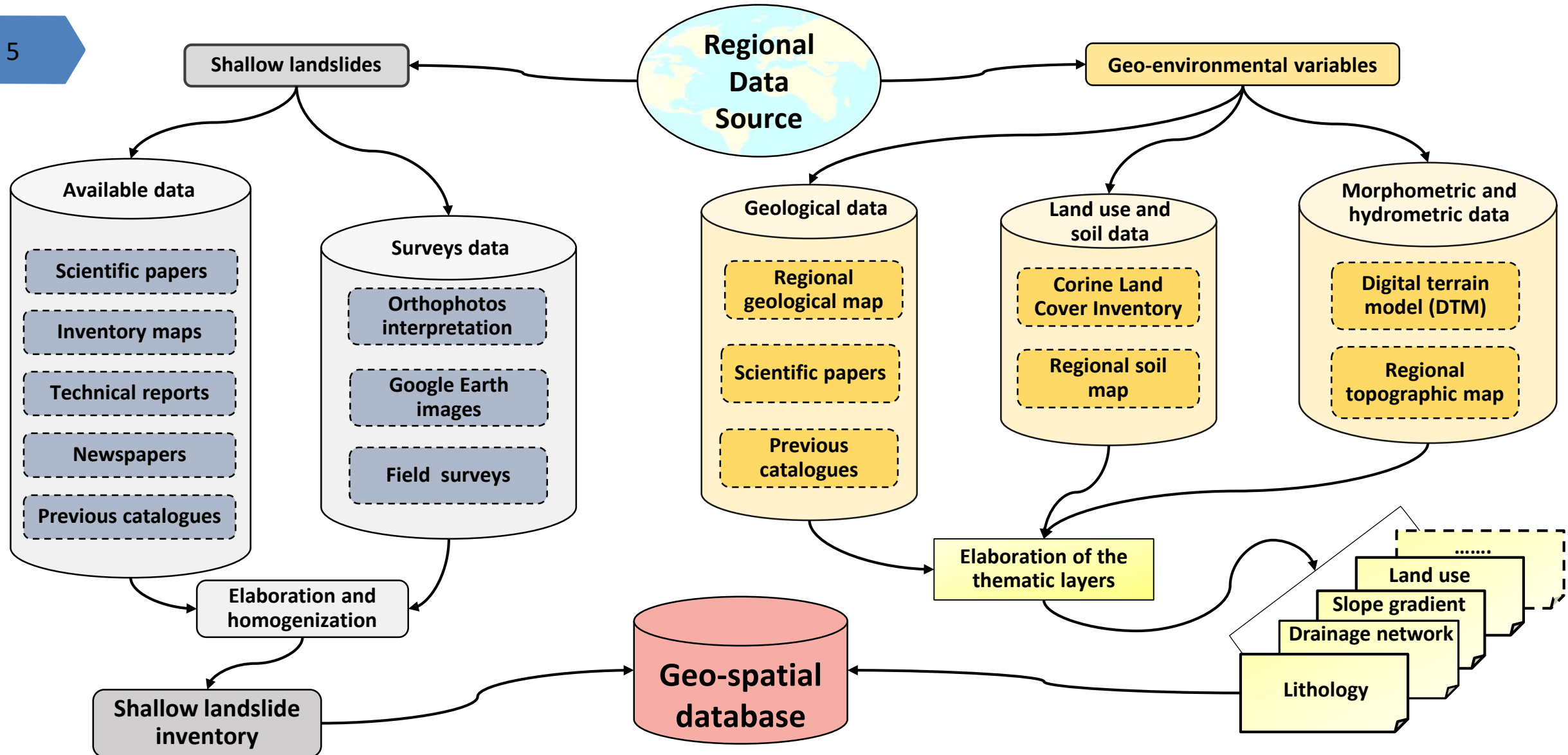
Da Gulla et al., 2021

Examples of landslides occurring in the Region



WORKFLOW OF DATABASE PREPARATION

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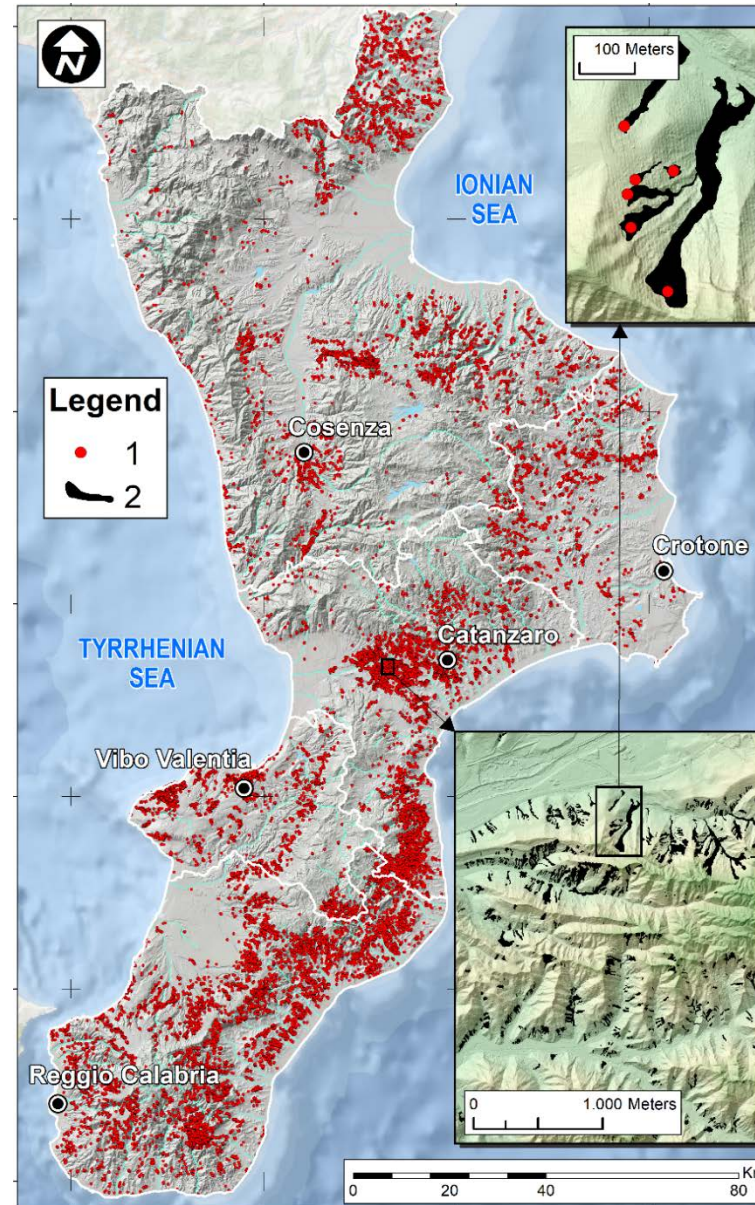


Shallow landslide inventory

Source data used to create the shallow landslides database for the Calabria region. (Da Gullà et al., 2021)

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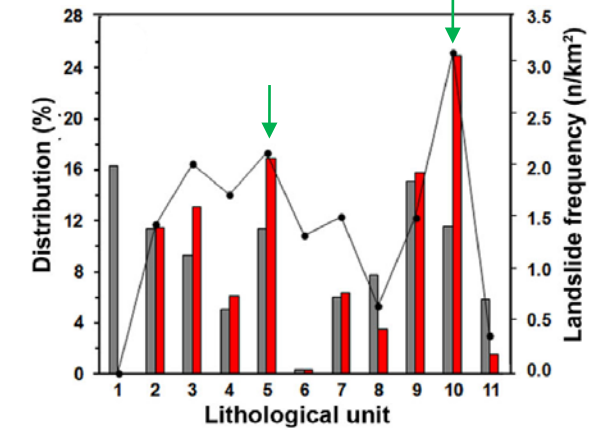
Source data	References	Period/Rainfall event	Number of landslides	Contribution to inventory (%)	
Scientific papers	(Gullà et al. 2008)	Rainfall events 1951 and 1953	7977	36.2	
	(Conforti and Ietto 2020)	From 1990 to 2018	1358	6.2	
	(Borelli et al. 2018)	From 2001 to 2011	750	3.4	
	(Conforti et al. 2016)	From 1998 to 2014	266	1.2	
	(Rago et al. 2013)	From 2000 to 2012	210	1.0	
	(Conforti and Critelli 2012)	From 1991 to 2004	127	0.6	
	(Conforti, Robustelli et al. 2012)	From 1991 to 2004	72	0.3	
	(Conforti and Ietto 2019)	From 1998 to 2011	59	0.3	
	(Borelli, Giofrè et al. 2012)	Rainfall event, winter 2000	33	0.1	
	(Conforti, Filomena et al. 2012)	From 1991 to 2011	29	0.1	
	Inventory map	(Borelli et al. 2015)	From 2008 to 2010	3399	15.4
		(Sorriso-Valvo et al. 2004)	Rainfall event, september 2000	2084	9.5
		(Borelli, Critelli et al. 2012)	From 2000 to 2006	616	2.8
		(Lucà et al. 2011)	From 1990 to 2004	423	1.9
		(Rago et al. 2017)	Rainfall event 30 October–01 November 2015	133	0.6
(Iovine and Merenda 1996)		Rainfall event 1972–73	117	0.5	
(Tansi et al. 2016)		From 2008 to 2012	85	0.4	
(Biondino et al. 2018)		From 2014 to 2017	92	0.4	
Catalogues		Calabria Basin Authorities	From 2000 to 2016	934	4.2
		PhD thesis	(Conforti 2009)	From 2000 to 2008	95
	(Vigliarolo 2009)		From 2000 to 2006	418	1.9
Thecnical reports	Cosenza province	Rainfall event winter 2009–2010	55	0.2	
		Calabria region	Rainfall event winter 2009–2010	153	0.7
	Calabria region	Rainfall event winter 2008–2009	371	1.7	
	CNR-IRPI (Cosenza)	Rainfall events, september 2000 and winter 2008–2009	418	1.9	
	Newspapers	Regional newspapers	Rainfall events, winter 2010–2011 and 2013	64	0.3
		Photo interpretation and field survey	Orthophotos dated 2008, Google Earth satellite images dated, 2010, 2011, 2014, 2015 and 2016.	1690	7.7
	Total shallow landslide cataloged		From 1951 to 2017	22,028	100



Areal distribution and frequency of shallow landslides in the provinces.

Province	Shallow landslides		Landslide frequency (count/km ²)
	count	%	
Catanzaro	6410	29.1	2.7
Cosenza	4083	18.5	0.6
Crotone	1367	6.2	0.8
Reggio Calabria	7136	32.4	2.2
Vibo Valentia	3032	13.8	2.7
Whole Region	22028	100	1.5

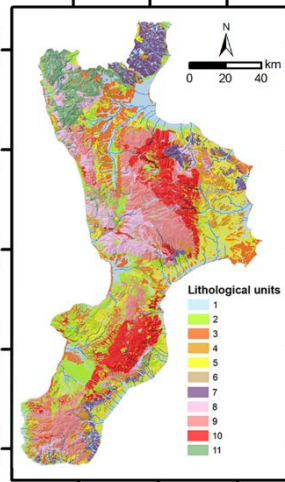
Lithology vs shallow landslides



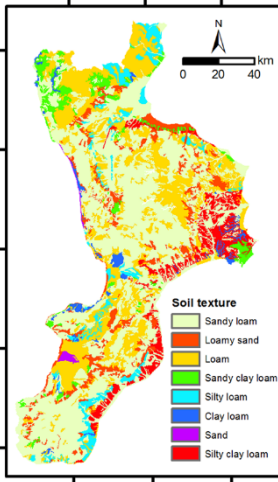
- 1) alluvial deposits, 2) gravel and sand deposits, 3) sandstone rocks, 4) conglomerate rocks, 5) clay and marl rocks, 6) evaporitic rocks, 7) flyschoid rocks, 8) low-grade metamorphic rocks, 9) middle-high-grade metamorphic rocks, 10) intrusive rocks, 11) carbonate rocks

Geo-environmental variables maps

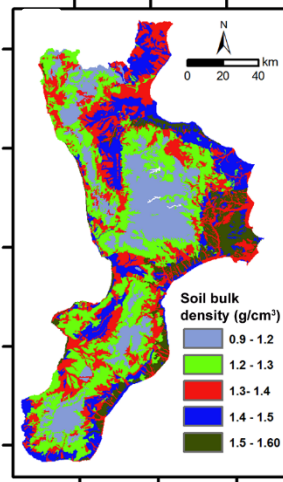
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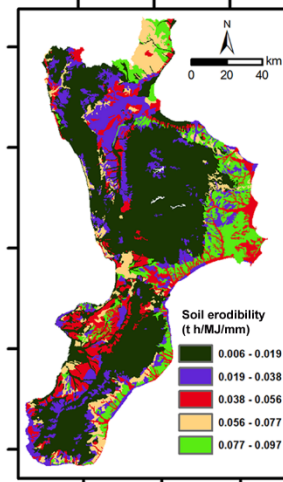
Lithology



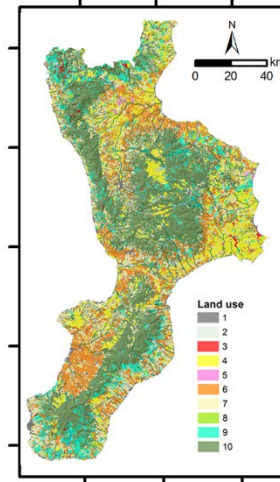
Soil texture



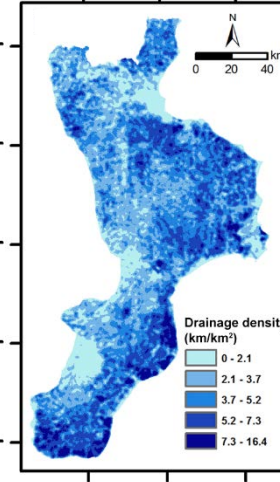
Soil bulk density



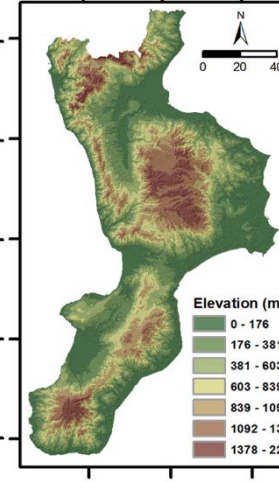
Soil erodibility



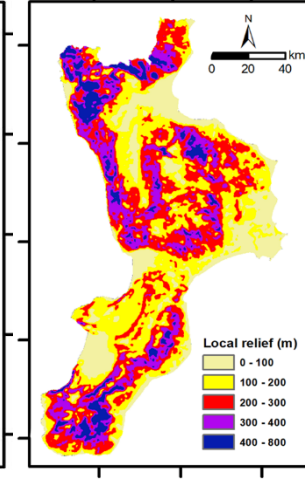
Land use



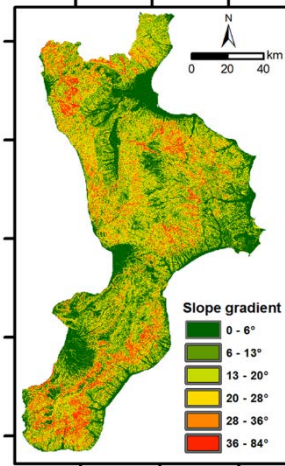
Drainage density



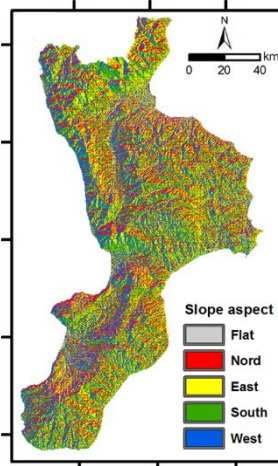
Elevation



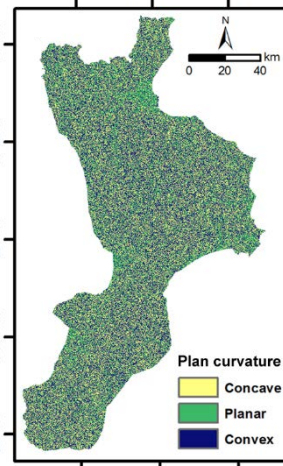
Local relief



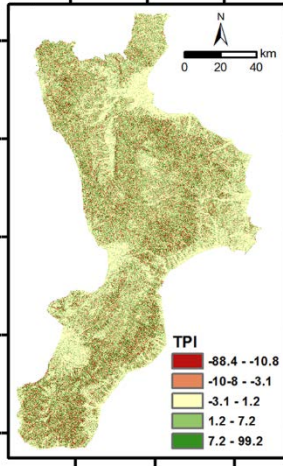
Slope gradient



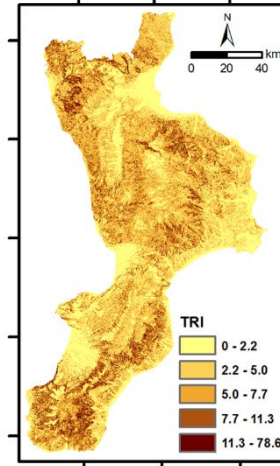
Slope aspect



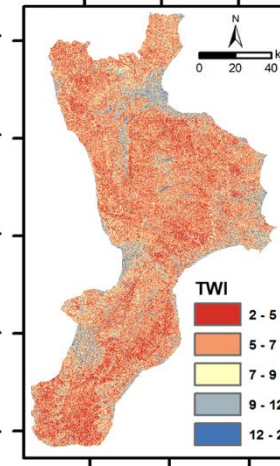
Plan curvature



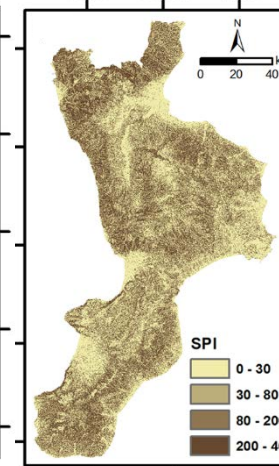
TPI



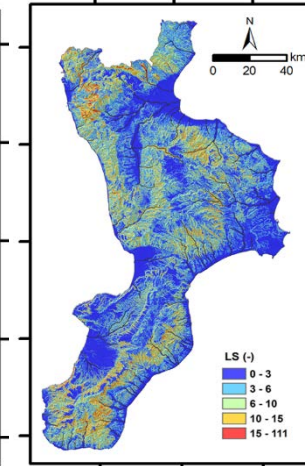
TRI



TWI



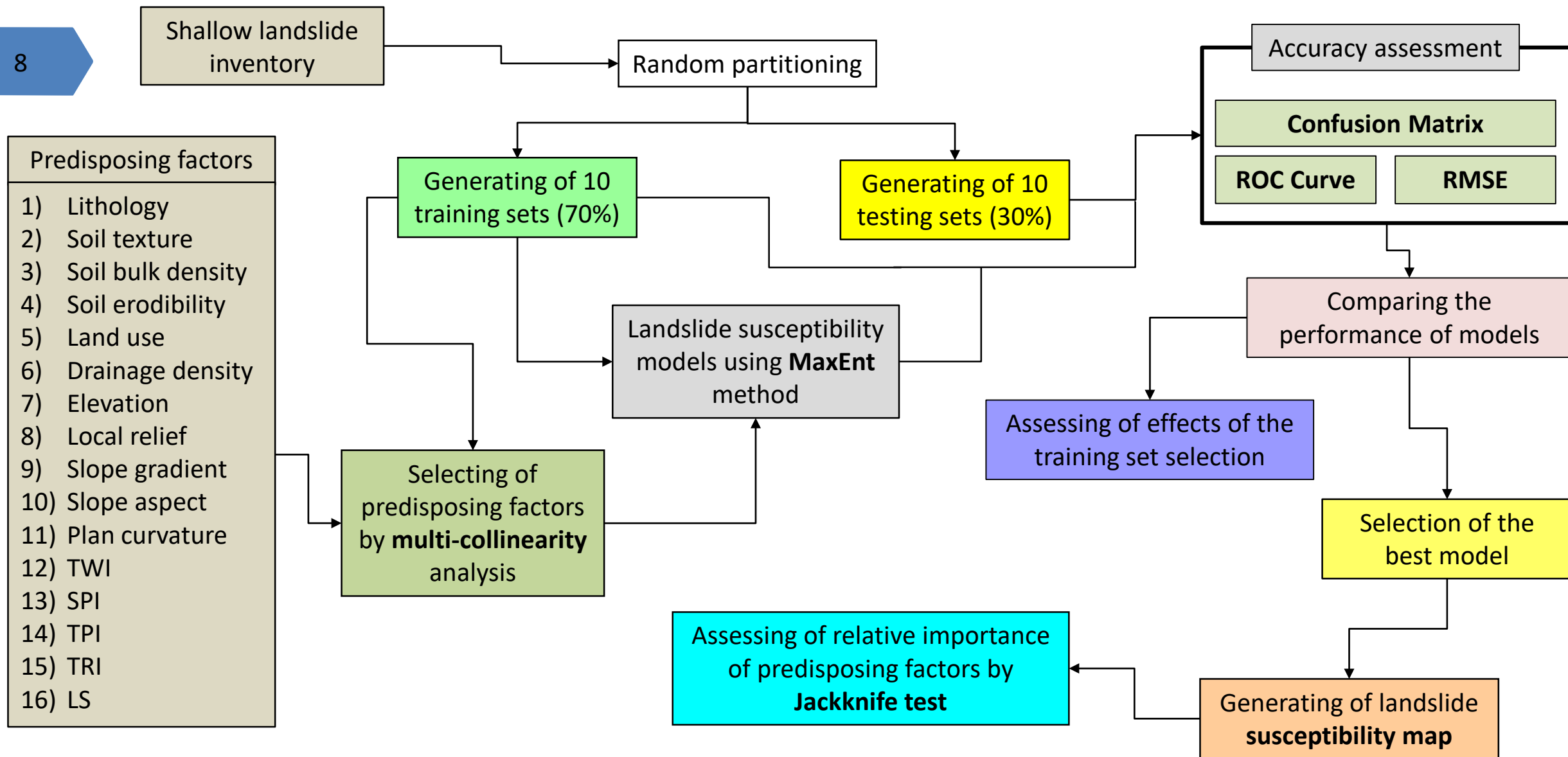
SPI



LS

Flow chart of the landslide susceptibility analysis

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RESULTS

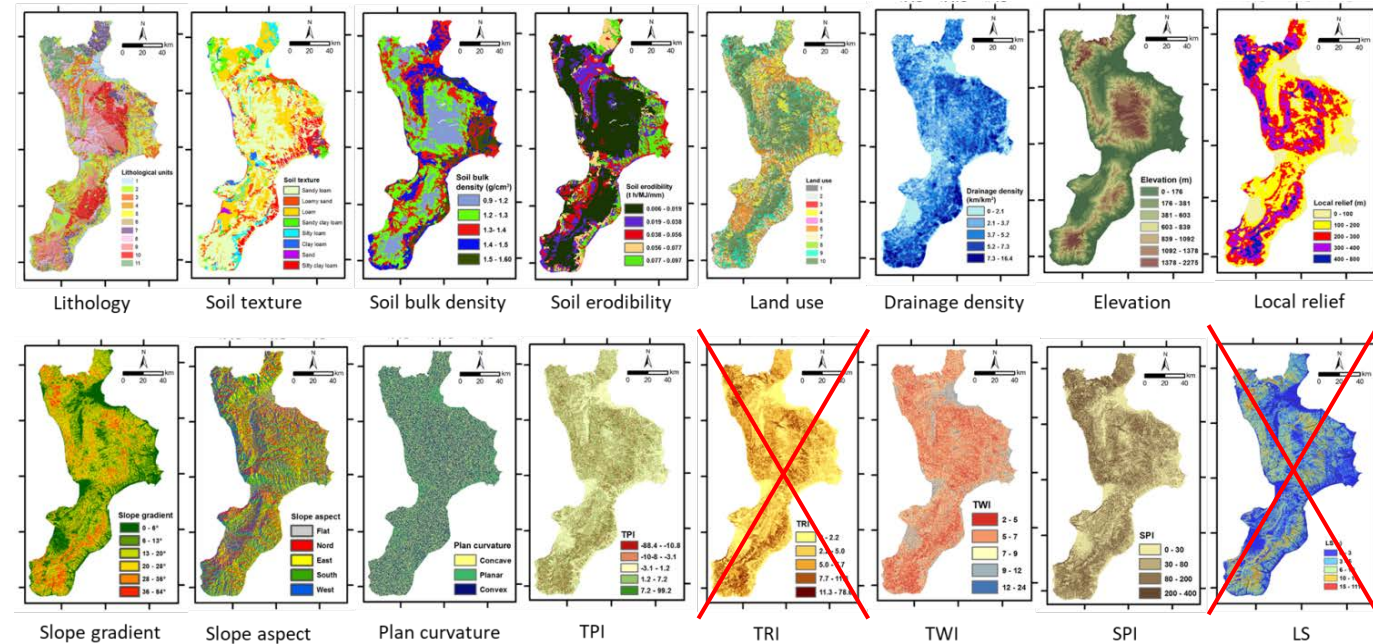
Multi-collinearity analysis

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Predisposing factor	Before		After	
	Tolerance	VIF	Tolerance	VIF
Lithology	0.827	1.210	0.830	1.205
Soil texture	0.797	1.255	0.802	1.247
Soil bulk density	0.380	2.635	0.391	2.557
Soil erodibility	0.460	2.174	0.471	2.124
Land use	0.854	1.171	0.855	1.169
Drainage density	0.749	1.335	0.749	1.334
Elevation	0.340	2.938	0.343	2.914
Local relief	0.386	2.588	0.395	2.530
Slope gradient	0.074	13.484	0.431	2.321
Slope aspect	0.992	1.008	0.993	1.008
Plan curvature	0.998	1.002	0.998	1.002
TPI	0.710	1.408	0.735	1.360
TRI	0.063	15.847	-	-
TWI	0.462	2.163	0.484	2.067
SPI	0.944	1.059	0.966	1.035
LS	0.151	6.610	-	-

VIF= variance inflation factor

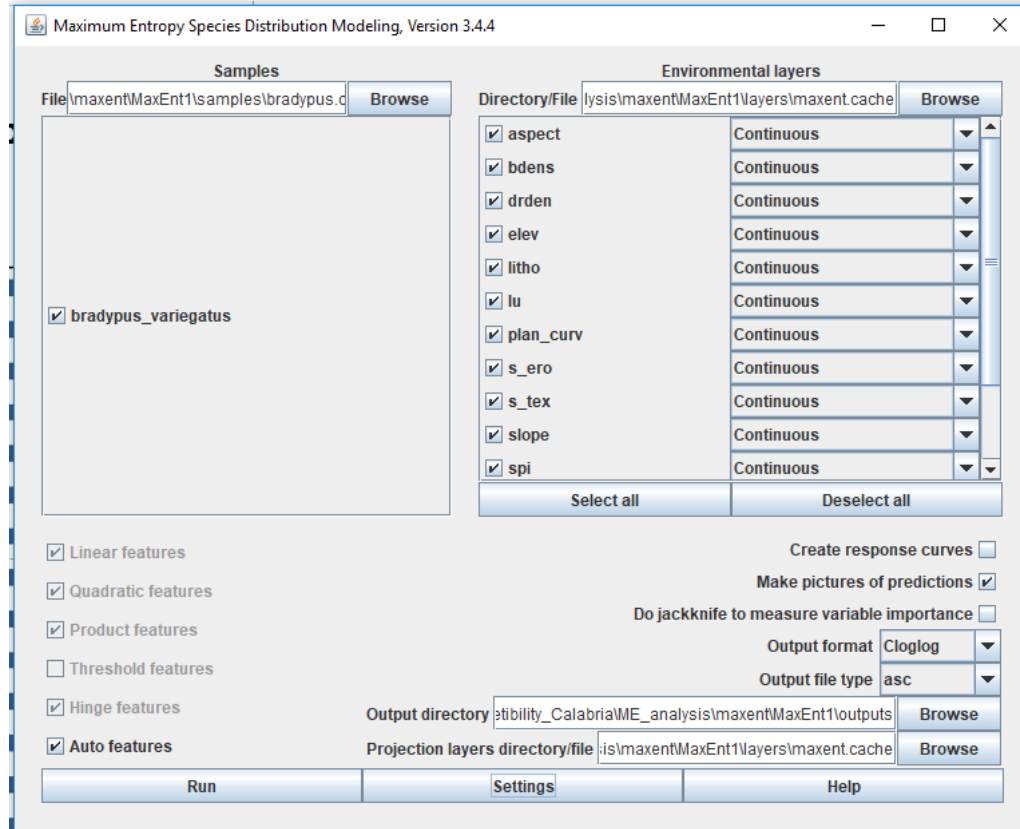
With *Tolerance* > 0.2 and *VIF* < 5, the variables are independent of each other.



RESULTS

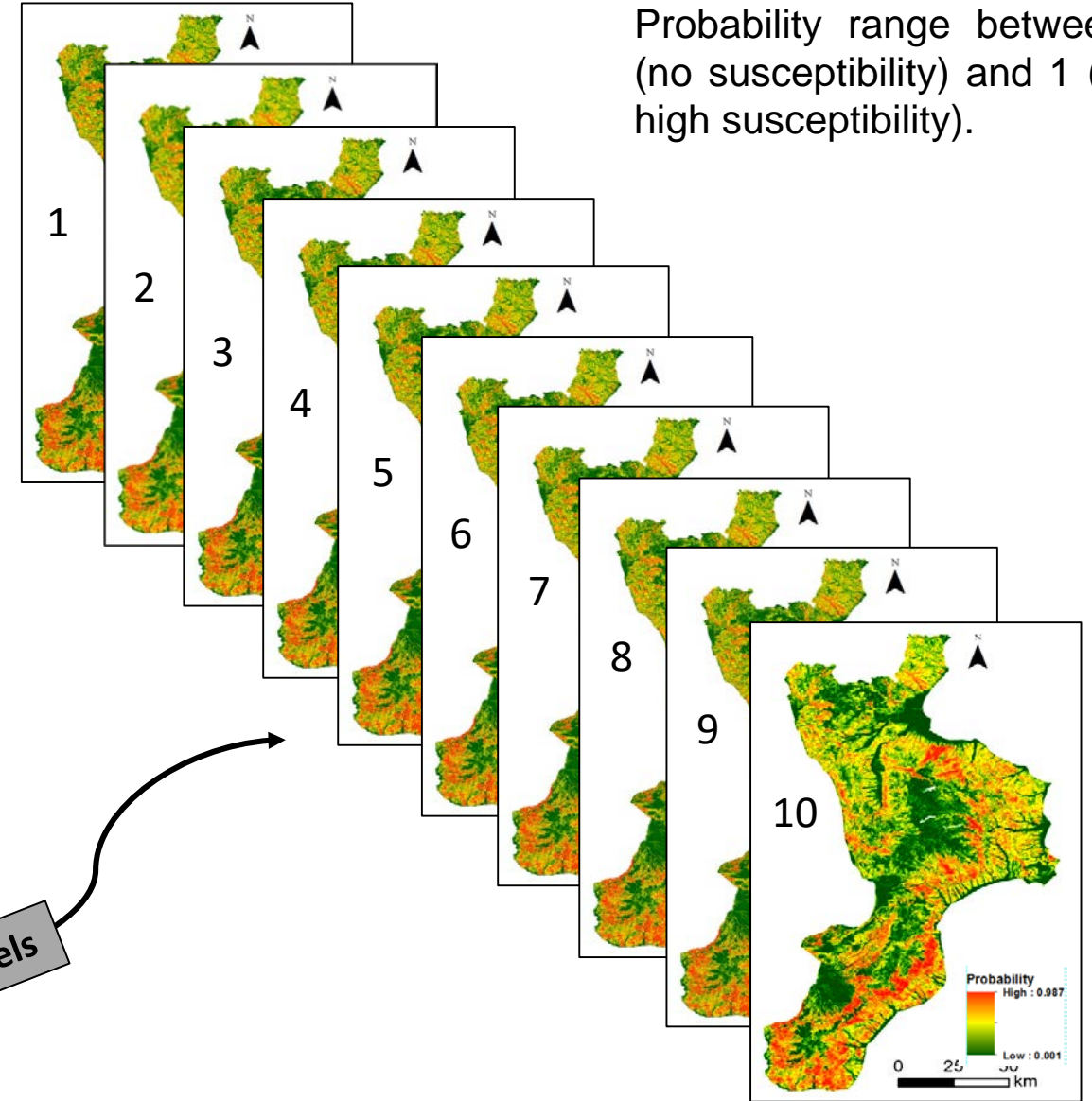
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Building of the 10 shallow landslide susceptibility models using Maximum Entropy method



MaxEnt 3.4.4 software

Models



Probability range between 0 (no susceptibility) and 1 (very high susceptibility).

Phillips et al. (2006) - Maximum entropy modeling of species geographic distributions. *Ecological Modelling* 190, 231–259.

RESULTS

Accuracy of models obtained by 10 replicates of training dataset

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Training dataset

Best model

Testing dataset

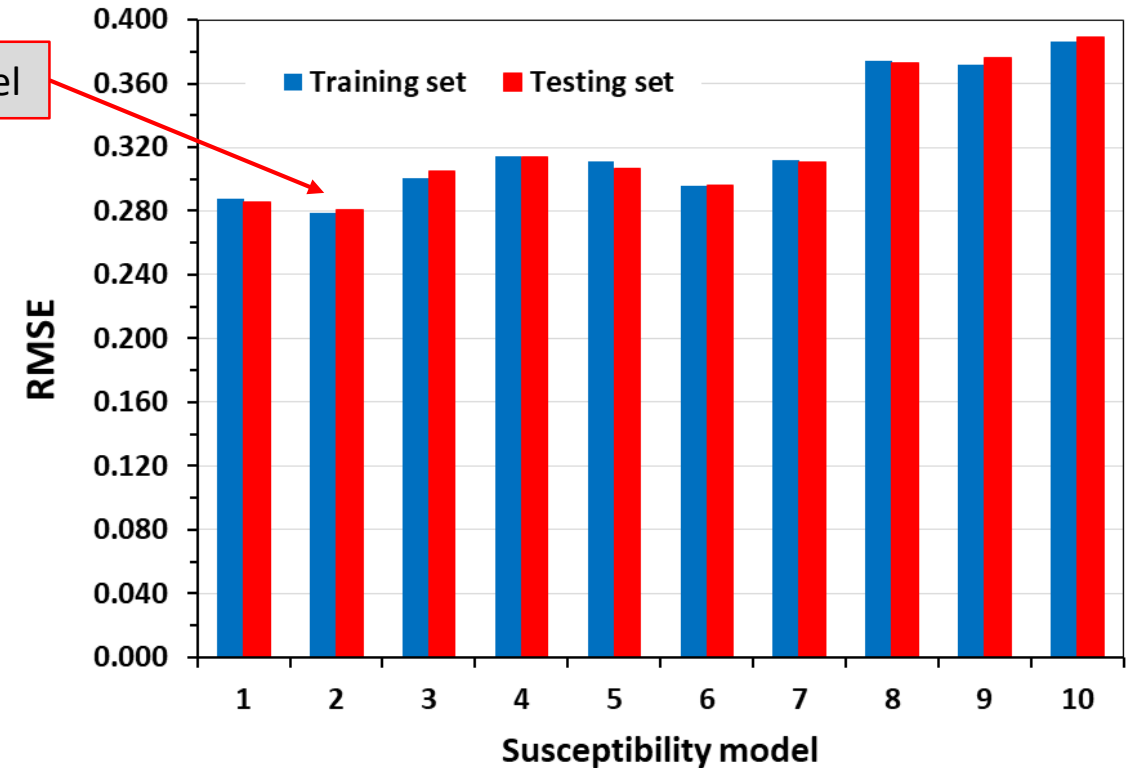
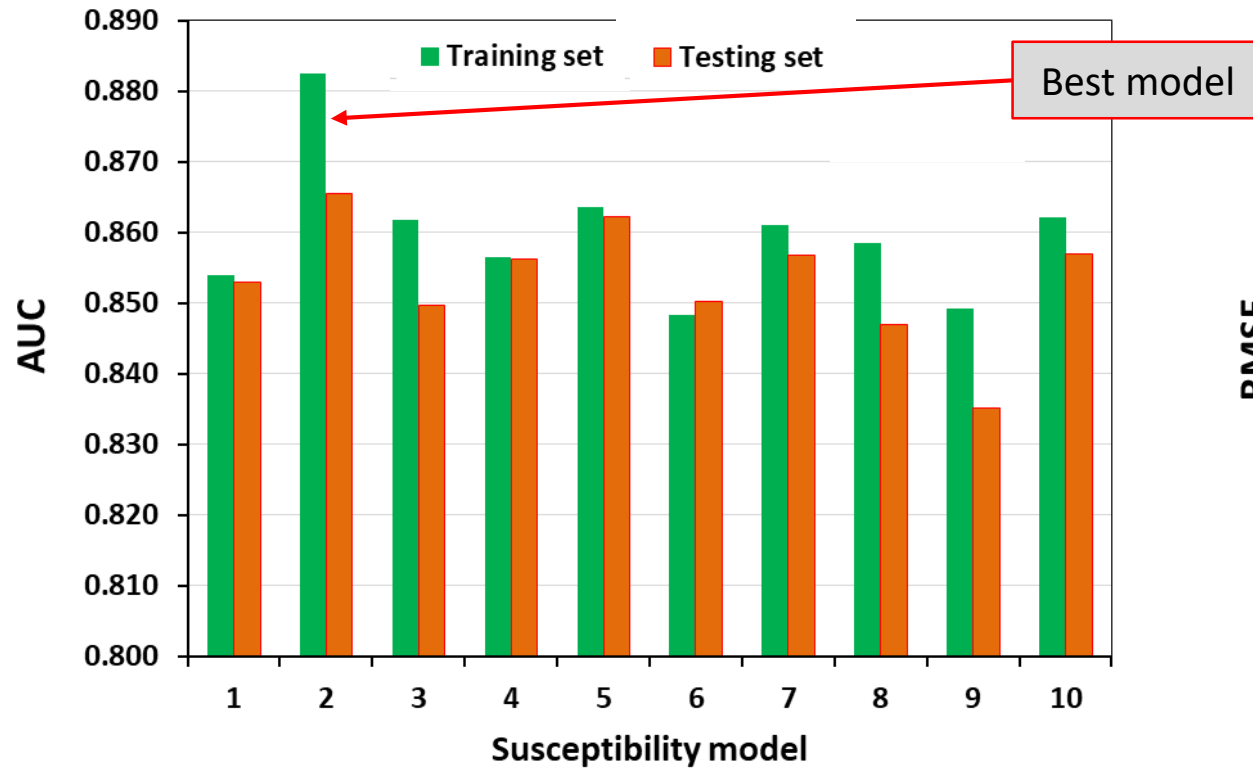
Model	Accuracy parameters			
	Sensitivity	Specificity	Accuracy	Kappa index
1	0.797	0.824	0.810	0.620
2	0.813	0.853	0.832	0.664
3	0.794	0.858	0.823	0.645
4	0.785	0.832	0.807	0.613
5	0.802	0.848	0.823	0.646
6	0.782	0.836	0.807	0.614
7	0.800	0.849	0.823	0.645
8	0.800	0.832	0.815	0.631
9	0.784	0.820	0.801	0.601
10	0.800	0.844	0.820	0.641
Min	0.782	0.820	0.801	0.601
Max	0.813	0.858	0.832	0.664
Mean	0.796	0.840	0.816	0.632
S.dev	0.010	0.013	0.010	0.020
C.V.	1.213	1.513	1.186	3.089

Model	Accuracy parameters			
	Sensitivity	Specificity	Accuracy	Kappa index
1	0.795	0.820	0.807	0.614
2	0.811	0.838	0.824	0.647
3	0.800	0.833	0.816	0.632
4	0.776	0.832	0.802	0.603
5	0.807	0.832	0.819	0.637
6	0.798	0.820	0.809	0.617
7	0.794	0.826	0.809	0.618
8	0.789	0.806	0.797	0.594
9	0.805	0.792	0.798	0.596
10	0.808	0.826	0.817	0.634
Min	0.776	0.792	0.797	0.594
Max	0.811	0.838	0.824	0.647
Mean	0.798	0.823	0.810	0.619
S.dev	0.010	0.014	0.009	0.018
C.V.	1.315	1.705	1.127	2.920

RESULTS

Accuracy of models for each selection of training set and testing set

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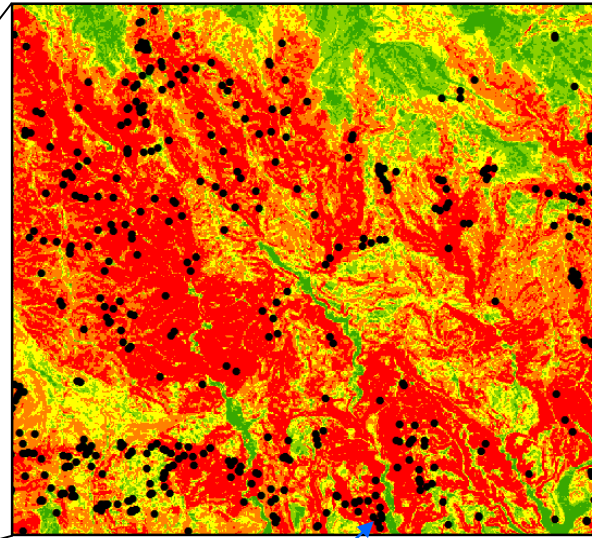
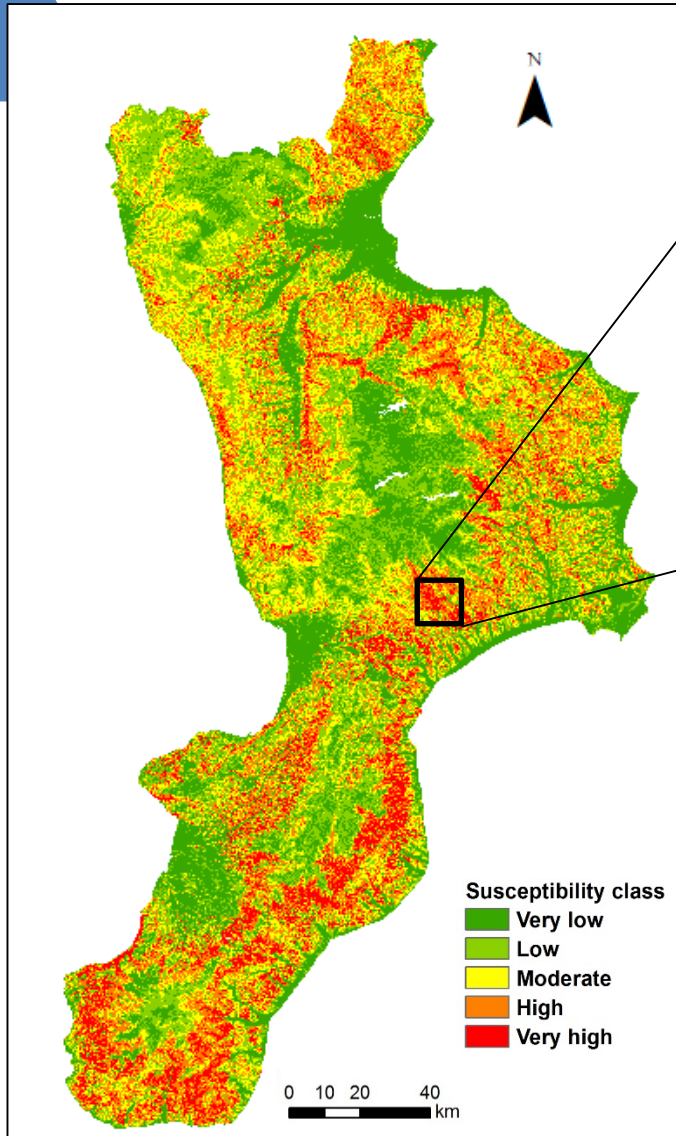
	Min	Max	Mean	S.dev	C.V.
Training sets	0.848	0.882	0.860	0.010	1.110
Testing sets	0.835	0.866	0.853	0.009	1.010

	Min	Max	Mean	S.dev	C.V.
Training sets	0.279	0.386	0.323	0.039	12.064
Testing sets	0.281	0.389	0.324	0.040	12.321

RESULTS

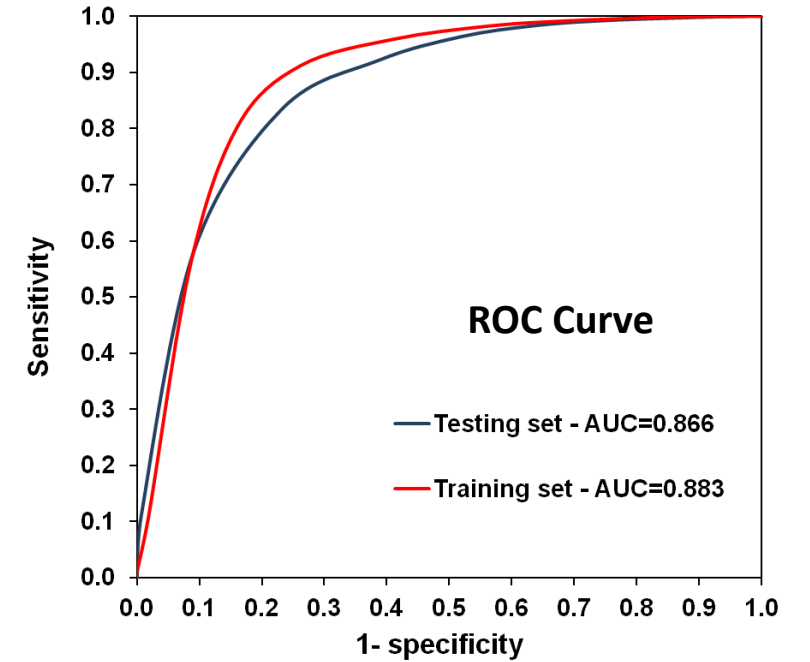
Shallow landslide susceptibility map

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Shallow landslide

Validation of susceptibility map



Percentage area of the susceptibility classes in the map, percentage of shallow landslides distribution and related landslide frequency falling in the susceptibility classes.

Probability range	Susceptibility class	Area (%)	Shallow landslides (%)			Landslide frequency
			Training set	Testing set	All dataset	
0.00 – 0.18	Very low	22.8	0.7	0.7	0.7	0.03
0.19 - 0.41	Low	25.5	3.9	3.5	3.8	0.15
0.42 - 0.61	Moderate	23.6	12.1	9.8	11.4	0.48
0.62 - 0.81	High	18.6	28.9	29.1	28.9	1.55
0.81 - 0.98	Very high	9.6	54.4	56.9	55.2	5.78

83.3%

86.0%

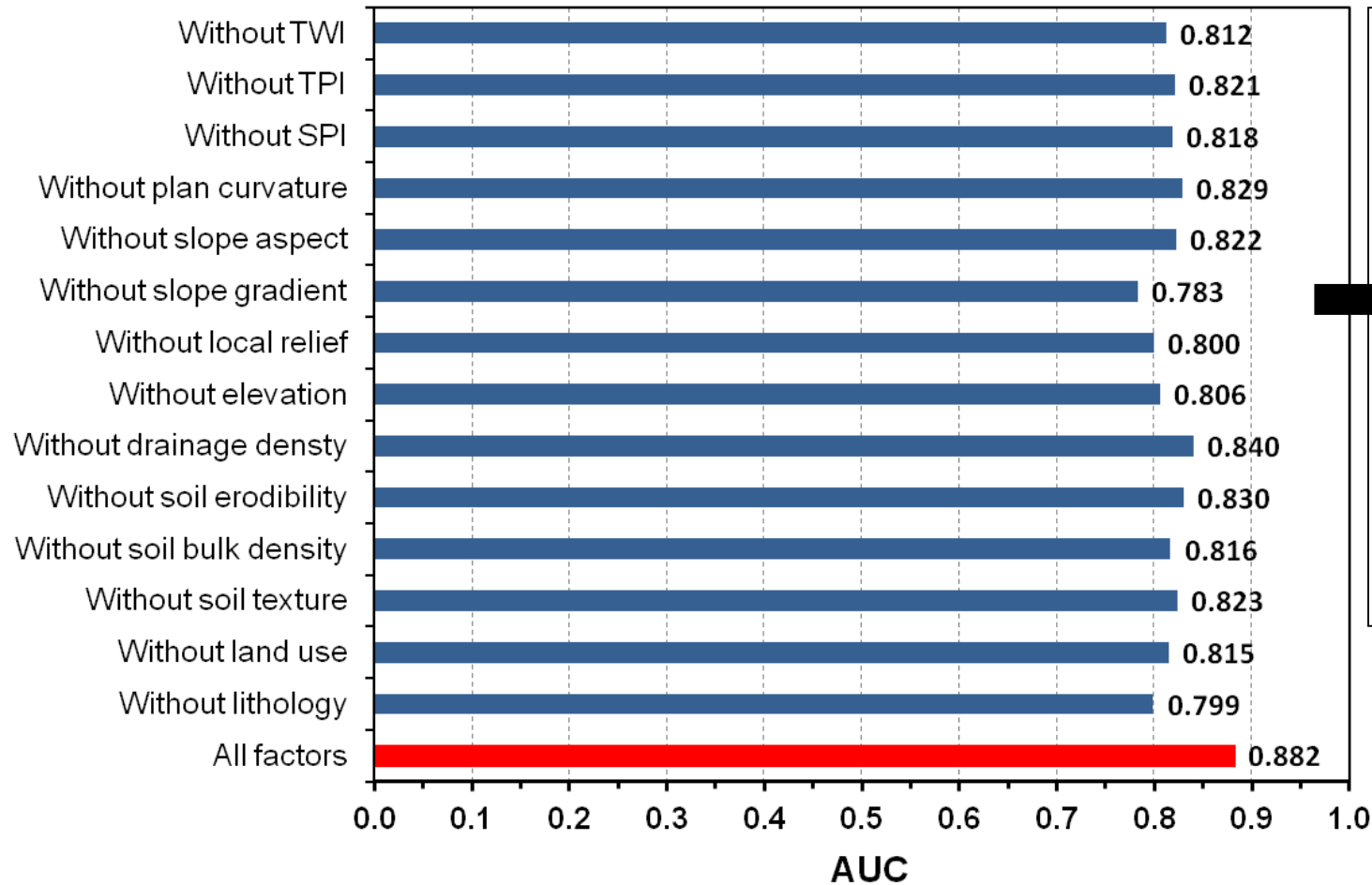
84.1%

RESULTS

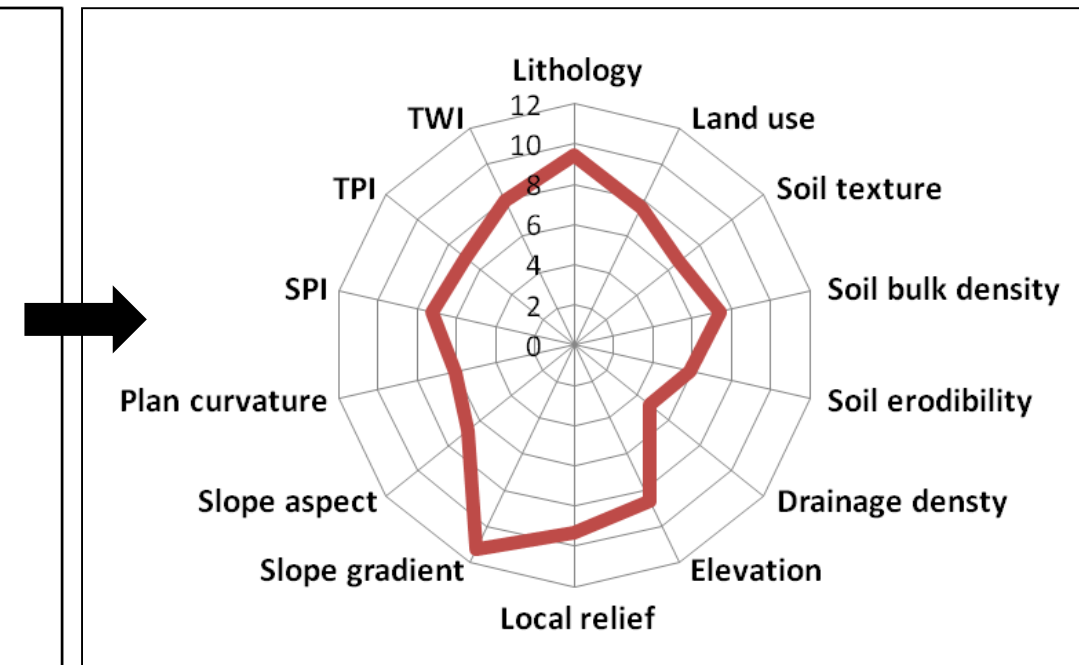
Sensitivity of predisposing factors used in the shallow landslide susceptibility model

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Jack-knife test



Relative importance (RI)



$$RI = \frac{(AUC_{all} - AUC_i)}{AUC_{all}} \times 100$$

CONCLUSIONS

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- The data collected in this study contribute to build a regional shallow landslide database that can be used both for land use planning and hazard-risk assessment at regional scale;
- The results indicate that MaxEnt method is able for mapping shallow landslide prone-areas within Calabria region;
- The MaxEnt method appeared to be less sensible to the change of random training sets;
- The results highlighted also that the selection and the sensitivity evaluation of the predisposing factors is a key requirement for modeling of landslide susceptibility;
- Others methods should be tested in order to find the most appropriate model to produce landslide susceptibility maps, as well as evaluate the effects of the random selection of the training sets.