

EXAMPLE & TUTORIAL

HOW TO USE «**SEMI-AUTOMATIC FEATURE
ENGINEERING TOOL**» FOR PRODUCING LSM

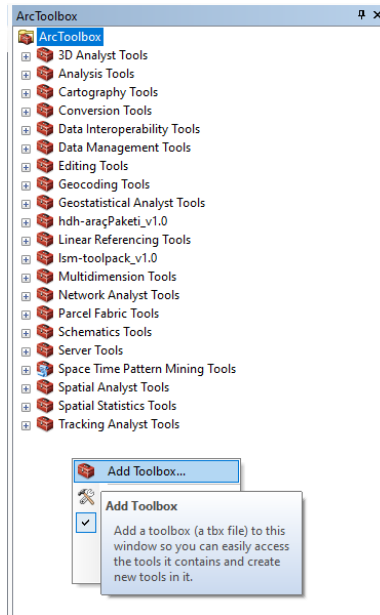
Requirements

- ArcGIS 10.3.1 or later or ArcGIS Pro 1.1 or later (don't have it? try trial edition)
- R Statistical Computing Software, 3.3.2 or later (What is R?). If you're experiencing issues when using a new version of R (i.e.g, R 4.0 and upper) Recommended Version v3.6.3
- ArcGIS R-Bridge -- Recommended Version v1.0.1.239
- Java Runtime Environment for FSelector package
- If you are setting up modules for the first time, you will need an internet connection to install the R-packages on the depository.

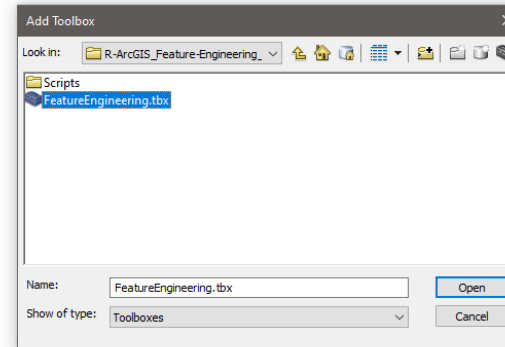
Installation

- Download this [repository](https://github.com/emrehanks/R-ArcGIS_Feature-Engineering_ToolPack) and unzip 'R-ArcGIS_Feature-Engineering_ToolPack-main.zip' into the folder C:\targetfolder.
 - https://github.com/emrehanks/R-ArcGIS_Feature-Engineering_ToolPack
- Open your ArcGIS application:

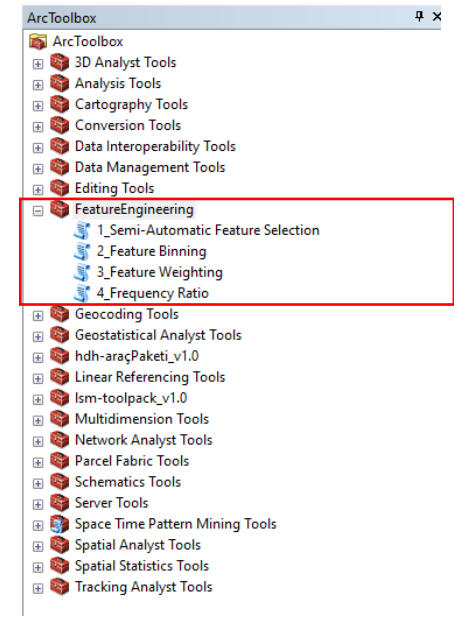
1. Navigate ArcToolBox panel



2. Add Toolbox on ArcMap

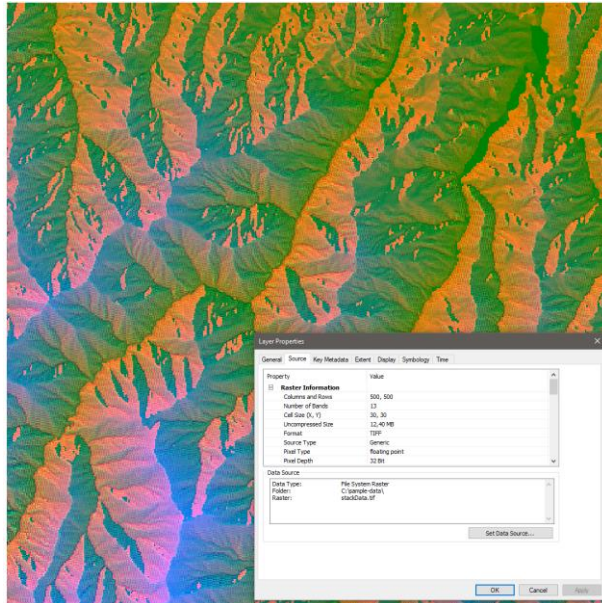


3. Be sure to toolbox added on ArcToolBox panel.

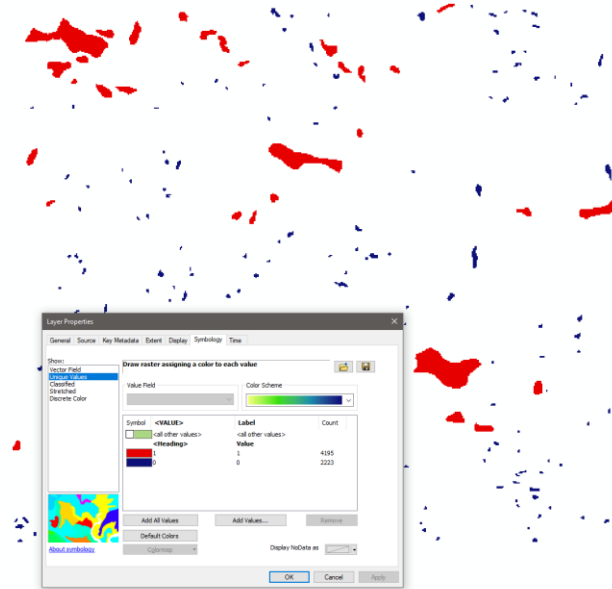


Sample Data

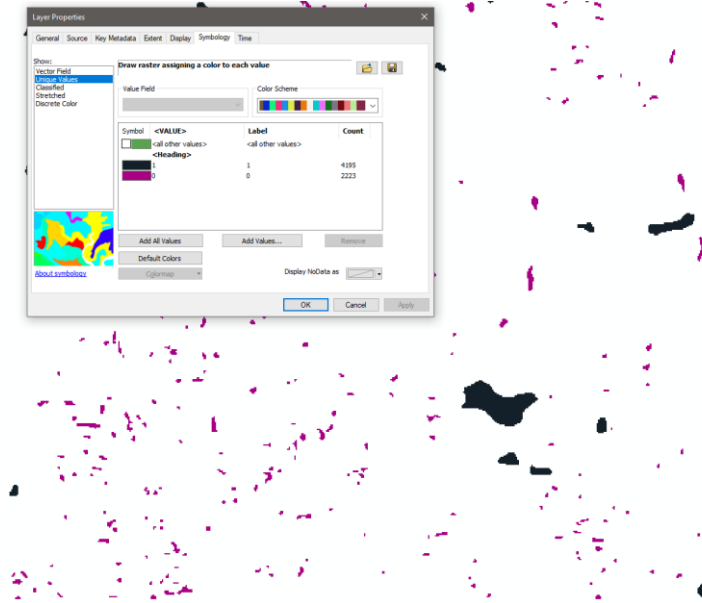
- Download and unzip '[sample-data.zip](#)' into the folder C:\\[targetfolder](#).
- This data contains:



stackData.tif -> Multi-bands geographic image (13 bands, 500*500 and 30m pixel size)
Not: band or factor names was included in this folder as name as «[stackNames.csv](#)»



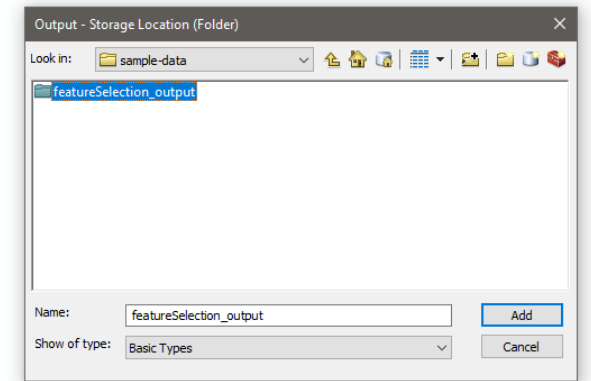
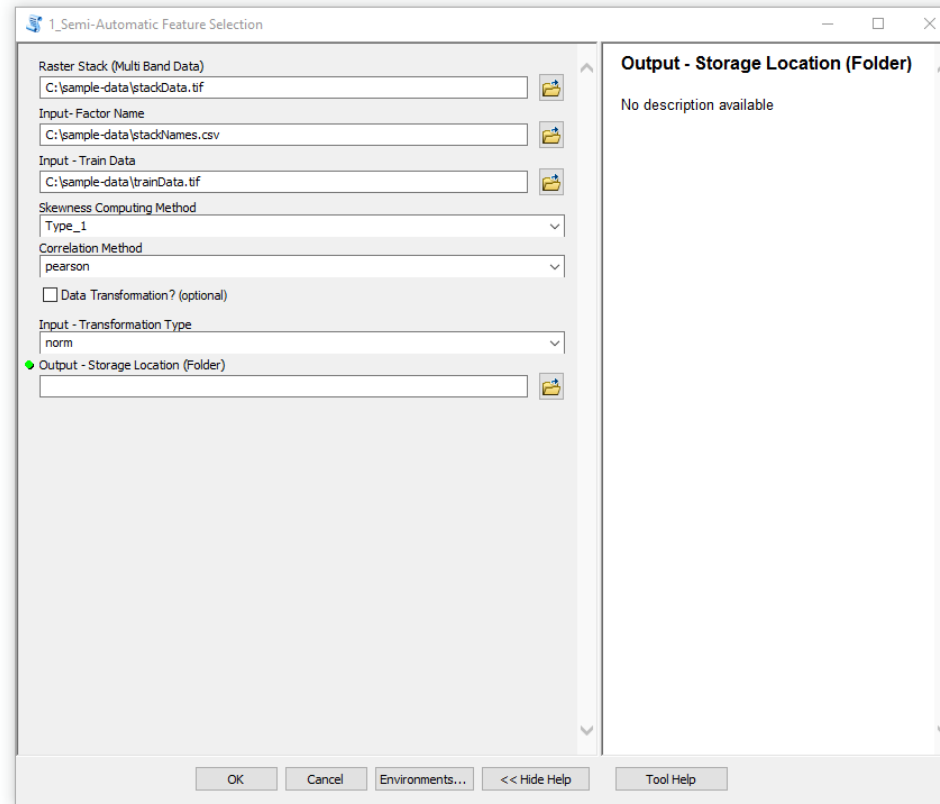
trainData.tif -> Single-bands geographic image (30m pixel size)



validationData.tif -> Single-bands geographic image (30m pixel size)

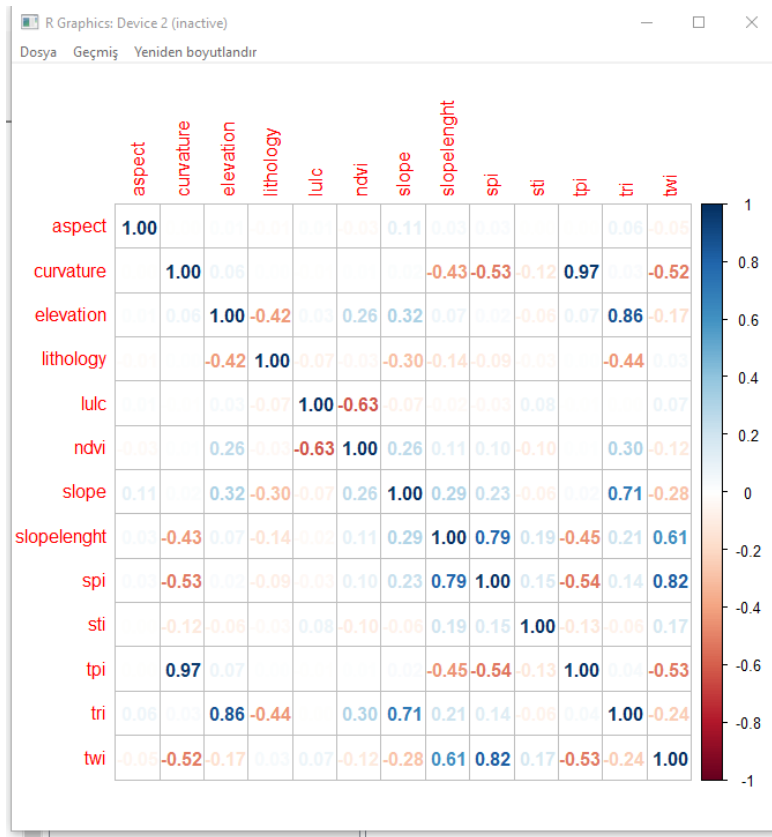
1_Semi-Automatic Feature Selection

1. Import stackData.tif
2. If available, import stackNames.csv.
3. Import train image file **Not:**
Landslide area must be labeled as '1' and non-landslide areas must be labeled as '0'.
4. Select type of skewness function.
5. Select type of Correlation method
6. Give the output location.
7. Select a folder and click Add
8. Click «OK»



1_Semi-Automatic Feature Selection

- Note that the data with correlation and click Ok on information window
- In this tutorial; The data namely **curvature**, **tri**, and **spi** were identified as the most correlated features.



```
rgdal: version: 1.5-27, (SVN revision 1148)
Geospatial Data Abstraction Library extensions to R successfully loaded
Loaded GDAL runtime: GDAL 3.2.1, released 2020/12/29
Path to GDAL shared files: C:/Users/emreh/OneDrive/Documents/R/win-library/3.6/sf/gdal
GDAL binary built with GEOS: TRUE
Loaded PROJ runtime: Rel. 7.2.1, January 1st, 2021, [PJ_VERSION: 721]
Path to PROJ shared files: C:/Users/emreh/OneDrive/Documents/R/win-library/3.6/rgdal/proj
PROJ CDN enabled: FALSE
Linking to sp version:1.4-5
To mute warnings of possible GDAL/OSR exportToProj4() degradation,
use options("rgdal_show_exportToProj4_warnings"="none") before loading sp or rgdal.
Overwritten PROJ_LIB was C:/Users/emreh/OneDrive/Documents/R/win-library/3.6/rgdal/proj
Attaching package: 'dplyr'

The following objects are masked from 'package:raster':

  intersect, select, union

The following object is masked from 'package:spatialEco':

  combine

The following objects are masked from 'package:rgeos':

  intersect, setdiff, union

The following object is masked from 'package:gridExtra':

  combine

The following objects are masked from 'package:stats':

  filter, lag

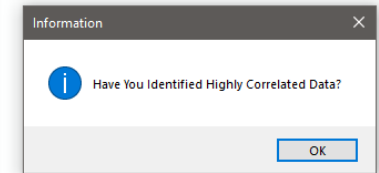
The following objects are masked from 'package:base':

  intersect, setdiff, setequal, union

Attaching package: 'SparseM'

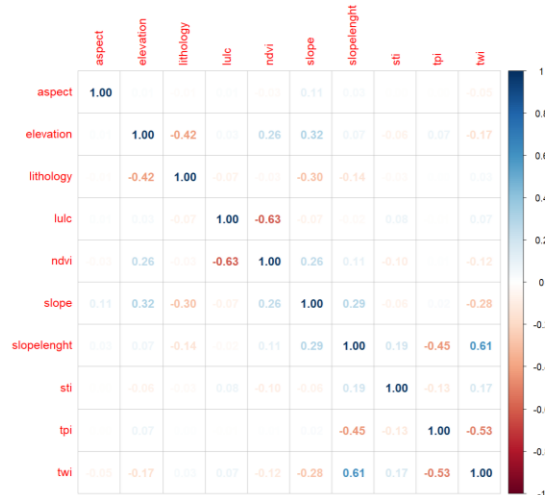
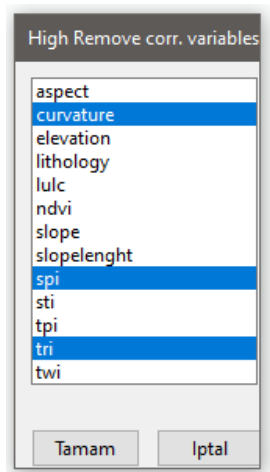
The following object is masked from 'package:base':

  backsolve
```

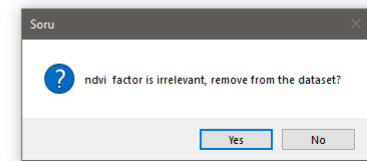
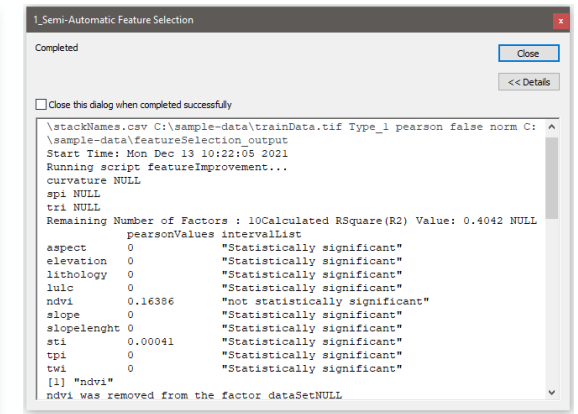
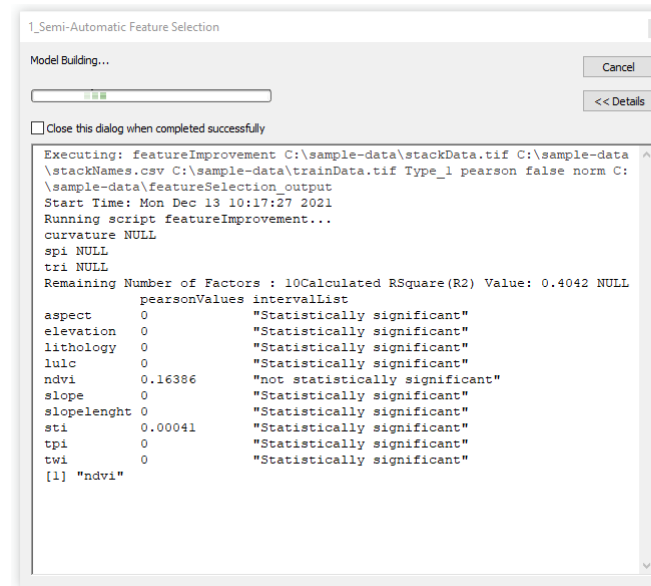


1_Semi-Automatic Feature Selection

1. You can select multiple features by «ctrl and click».
2. The second information graph about correlation statistic was shown as below:

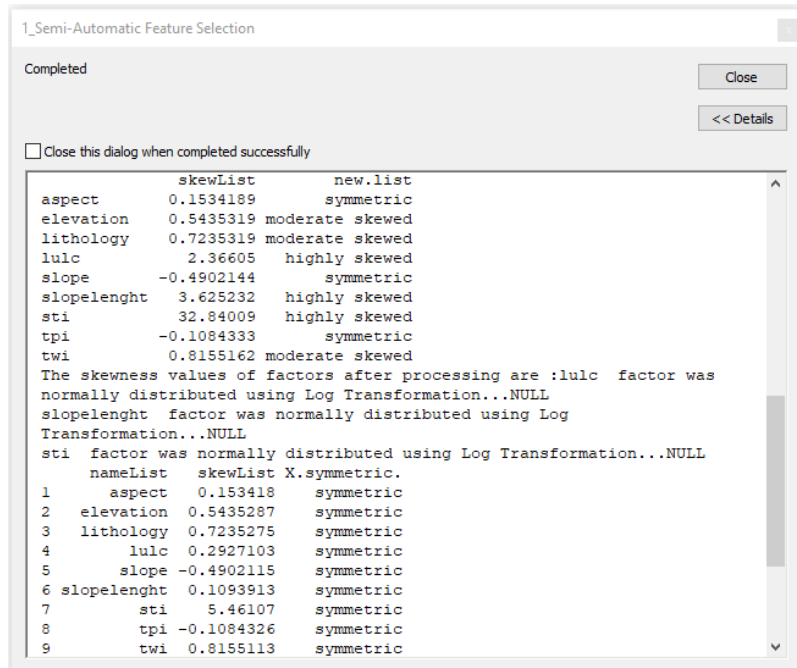


3. Pearson Correlation reported that the NDVI feature is not statistically significant. The tool is awaiting confirmation from us whether to delete this factor or not.

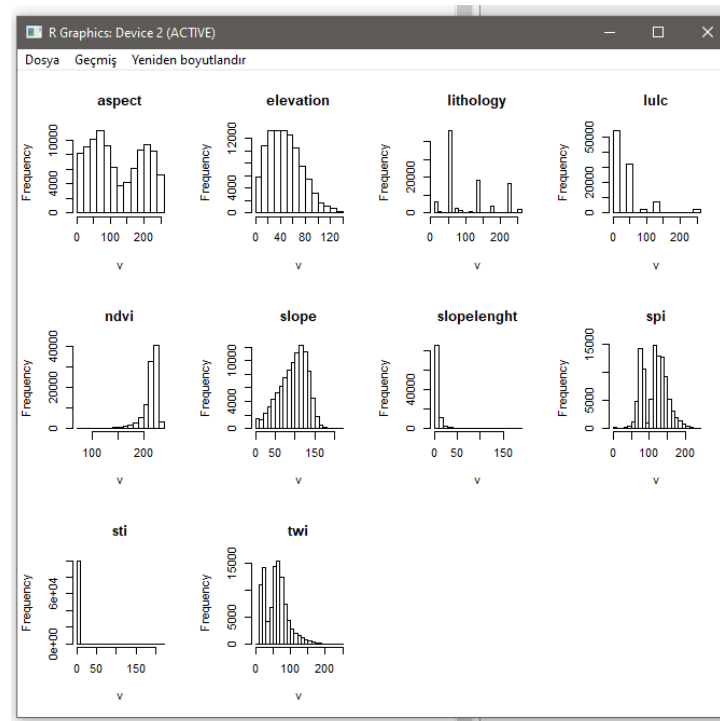


1_Semi-Automatic Feature Selection

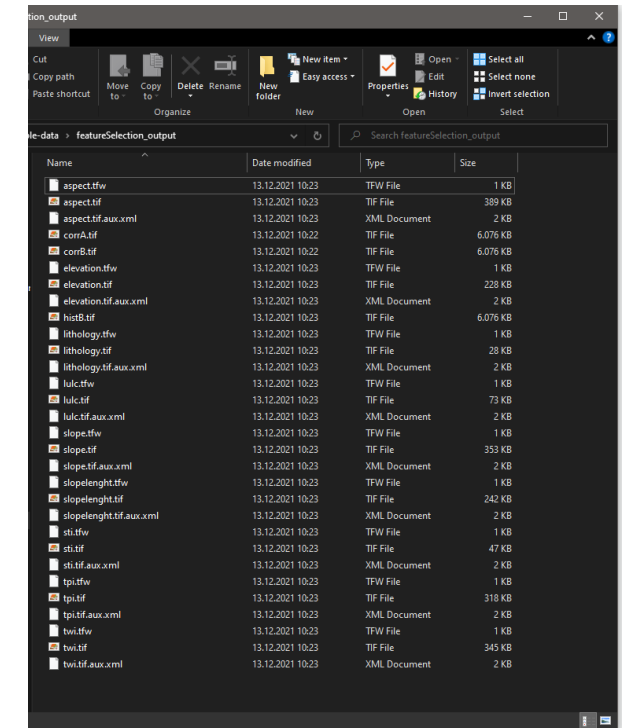
1. The before and after analyzed of the skewness score report is also shown in the dialog panel:
Not: Do not fill the check bar (i.e., Close this dialog when completed successfully) if you want to note and keep all scores for future use.



2. Histogram plots of features after skewness computing method processing.

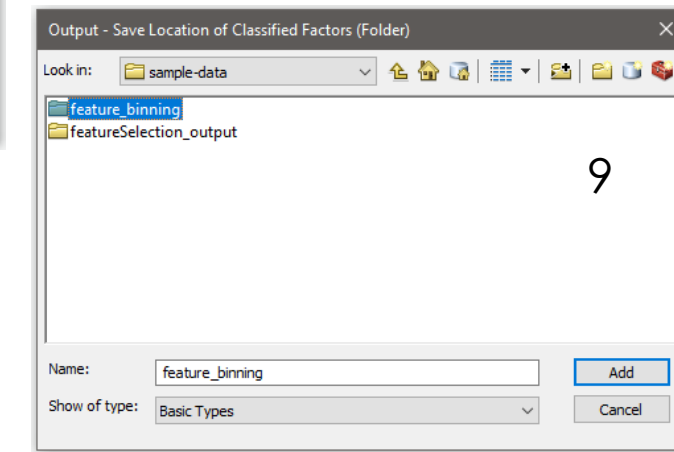
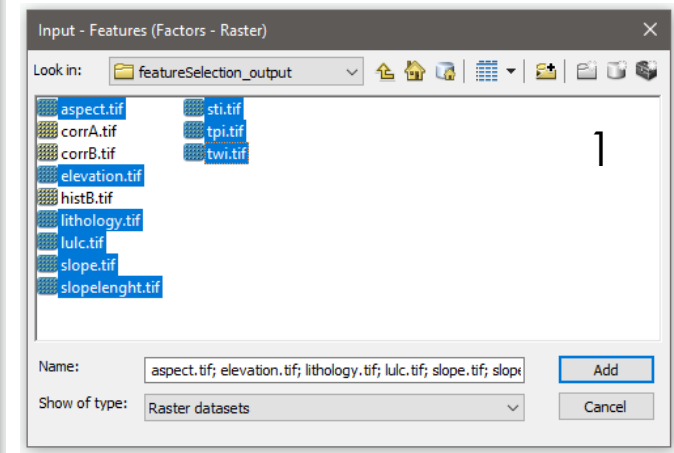
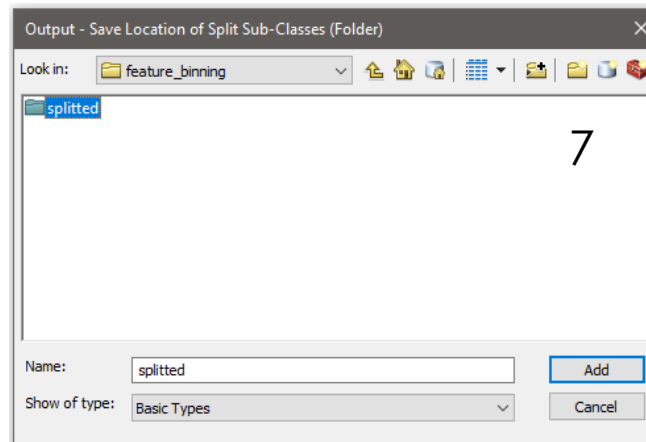
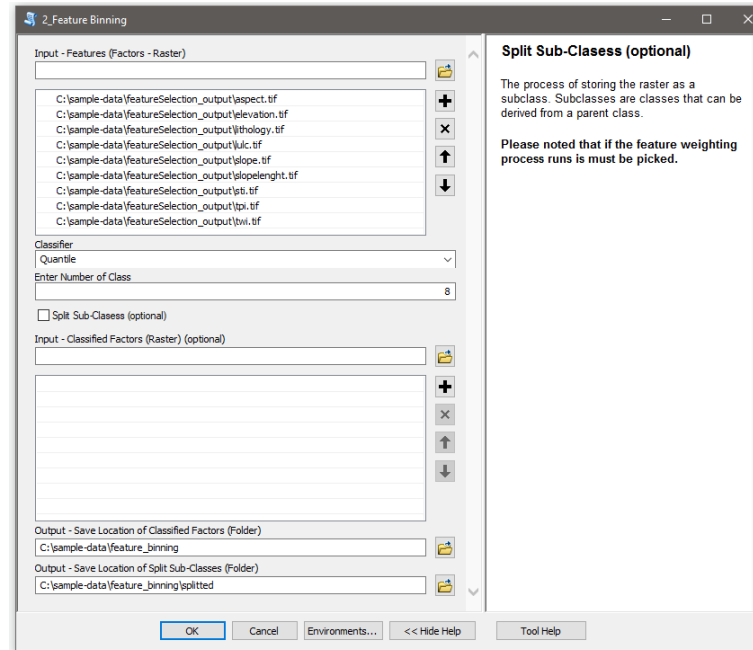


2. All feature selection processes successfully completed. Users also can access all process output from the selected folder.



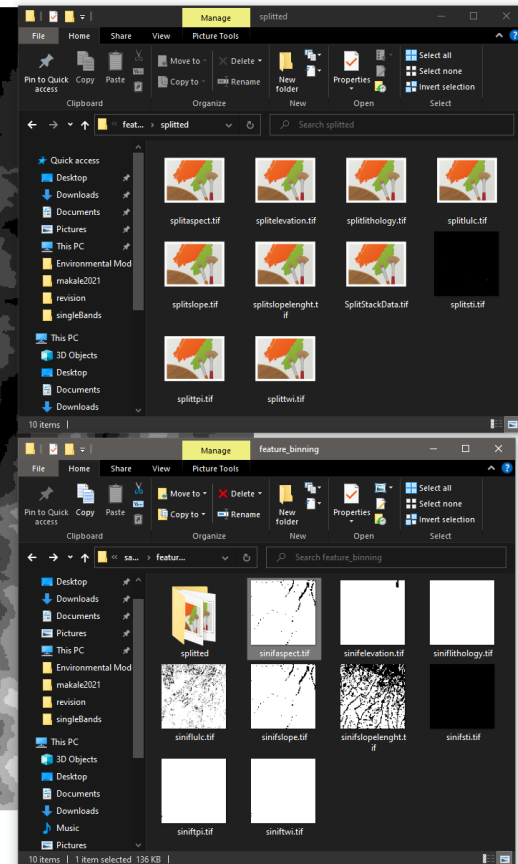
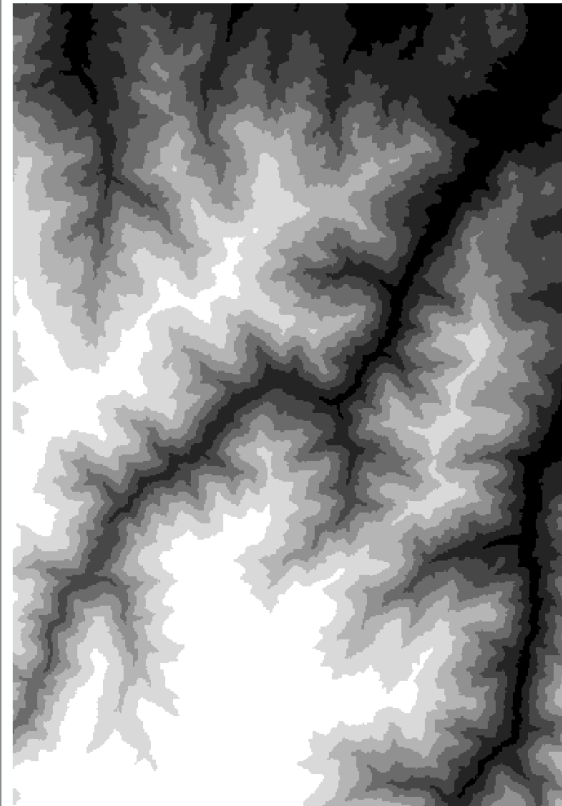
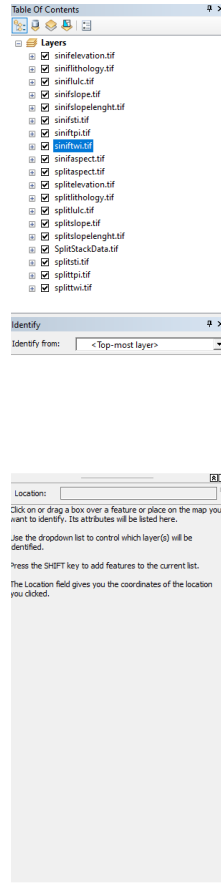
2_Feature Binning

1. Import only continuous feature in (Factors-Raster) tab.
2. Select Classifier method.
3. Enter number of class
4. If the user wants to process the feature weighting for future work, the pick as "Split Sub-Classes" must be approved
5. Import only features that categorical data type such as lithology and LULC. **However, since both lithology and lulc log are transformed by the Log Transformation process in this study, therefore both data should be selected in the first input area.**
6. Give the output location for the classified features.
7. Select a folder and click Add
8. Give the output location for split features
9. Select a folder and click Add
10. Click «OK»



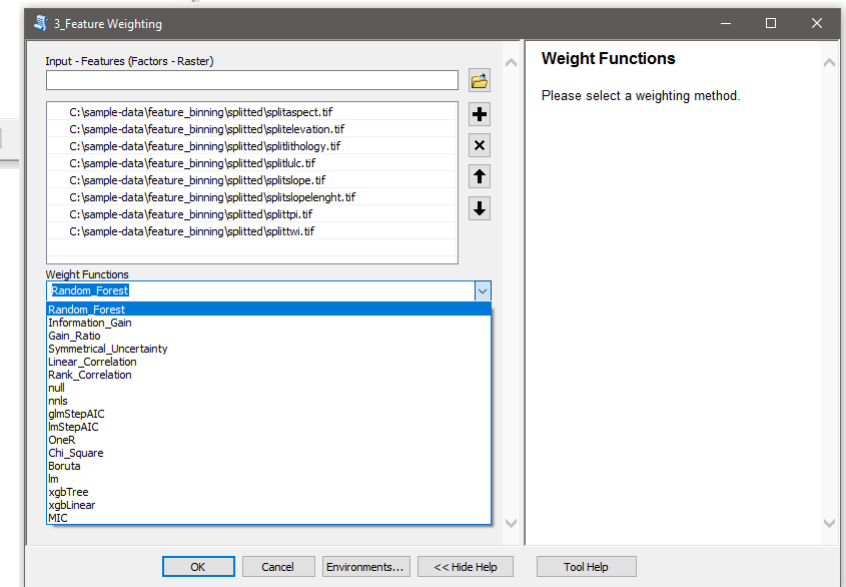
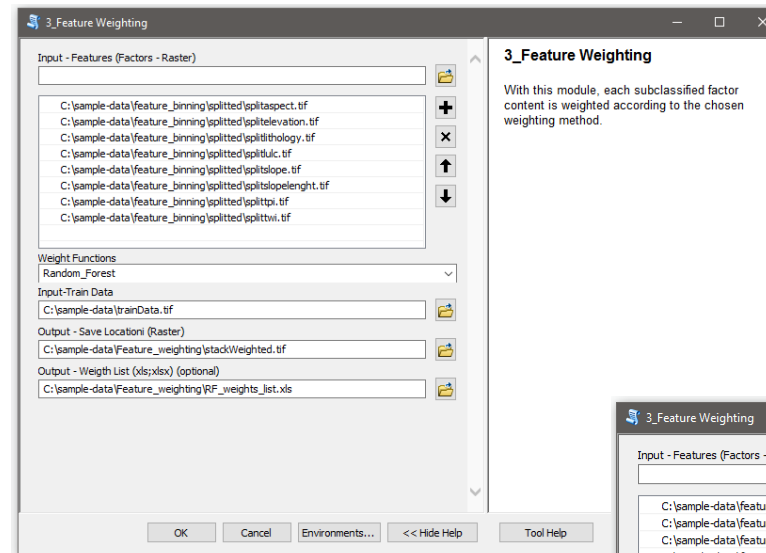
2_Feature Binning

- Feature Binning process is completed.
- Classified and split features are stored in the selected folder after the process.



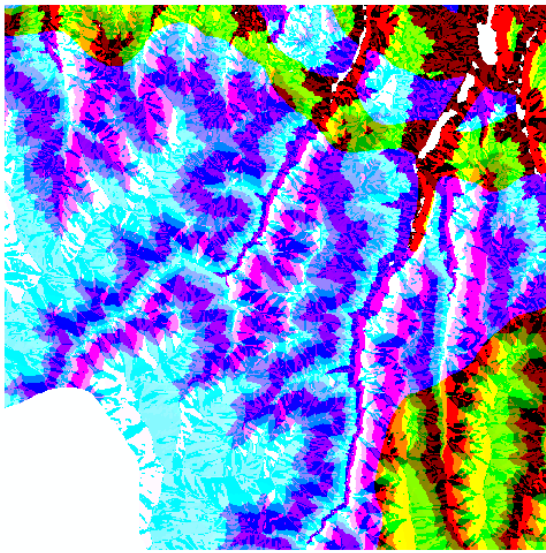
3_Feature Weighting

1. Import only split feature in (Factors-Raster) tab. Not: For this study data, the SPI factor was eliminated because of some problem about spi data type. Therefore, only 8 factors were selected for the weighing process.
2. Select Weight method. Suer selected several method for this process. However, not guaranteed to work every method including in this tab.
Recommended to user find the most suitable method by trial and error.
3. Import train data.
4. Give the output location for the weighted features as a raster stack data.
5. Select a folder and give raster data and klik OK
6. Give the output location for weight score of features
7. Select a folder and give XLS or XLSX document name
8. Click «OK»



3_Feature Weighting

- Raster stack data and lists of weight scores are located in the selected folder.
- Finally, the necessary raster stack data was obtained to generate the Landslide Susceptibility Mapping.

[illegible]

4_Frequency Ratio

1. Import only feature produced with the Feature Selection module. Not: For this study data, the SPI factor was eliminated because of some problem about spi data type. Therefore, only 8 factors were selected for the weighing process.
2. Select the Classifier method.
3. Give number of class.
4. If «msd» method was selected as a classifier method, the user should be given a standard variation score.
5. Import only Landslide data with a polygon vector.
6. Select a folder and give XLS or XLSX document name
7. Give the output location for the raster stack data.
8. Click «OK»

4_Frequency Ratio

This module is utilized to segment Single-Band raster data with the chosen number of classes and classifier method using the frequency ratio method.

Input - Features (Continuous Data)

C:\sample-data\featureSelection_output\aspect.tif
C:\sample-data\featureSelection_output\elevation.tif
C:\sample-data\featureSelection_output\lithology.tif
C:\sample-data\featureSelection_output\julc.tif
C:\sample-data\featureSelection_output\slope.tif
C:\sample-data\featureSelection_output\slopelenght.tif
C:\sample-data\featureSelection_output\spi.tif
C:\sample-data\featureSelection_output\twi.tif

Classifier

Quantile
Enter Number of Class: 8
Standard Deviation Number (msd): 1

Input - Landslide Area (Polygon)

C:\sample-data\inventoryShape\landslide.shp

Input - Classified Factors (Raster) (optional)

Output - Frequency Ratio Table (xls; xlsx) (optional)

C:\sample-data\frequency_ratio\FR_Scores.xls

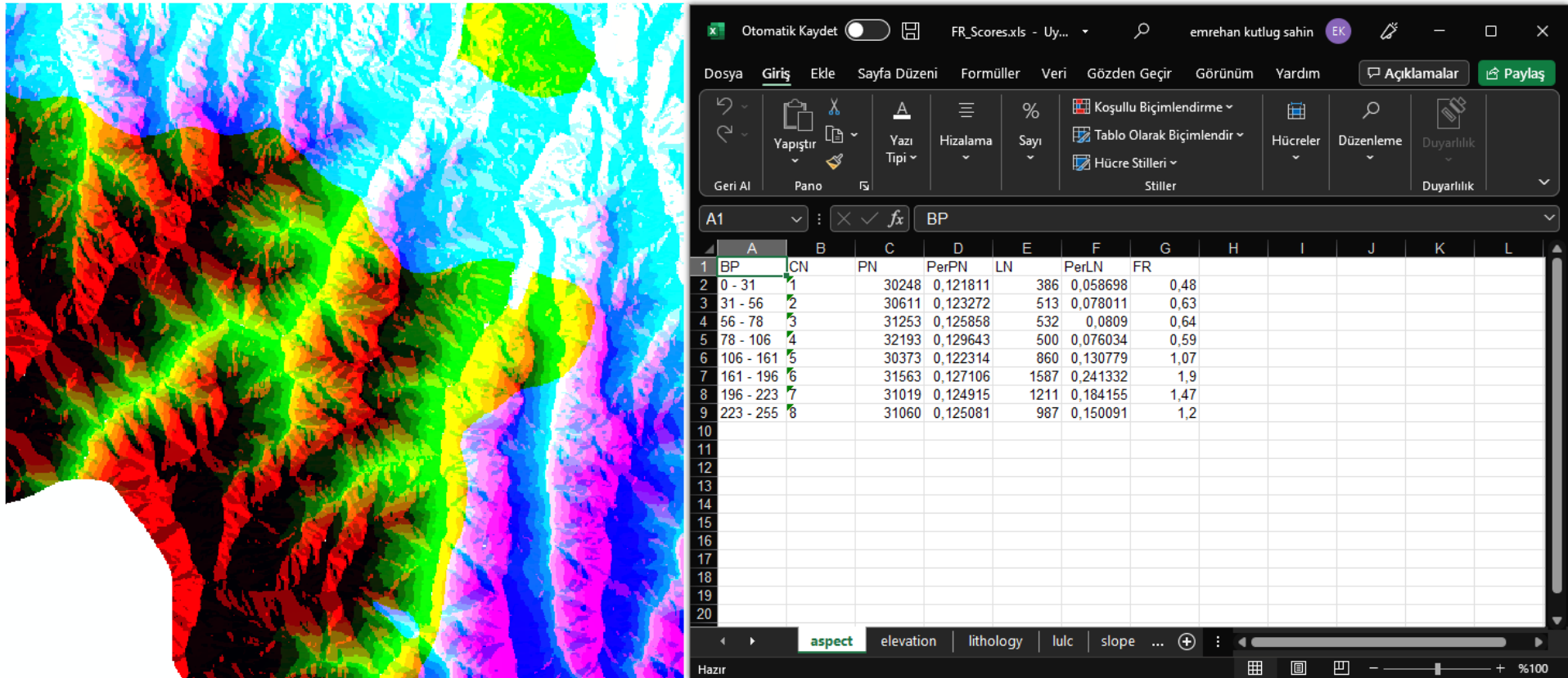
Output - Save Location (Raster)

C:\sample-data\frequency_ratio\FR_Stack.tif

OK Cancel Environments... << Hide Help Tool Help

4_Frequency Ratio

- Raster stack data and lists of frequency scores for each feature are located in the selected folder.
- Finally, the necessary raster stack data was obtained to generate the Landslide Susceptibility Mapping.

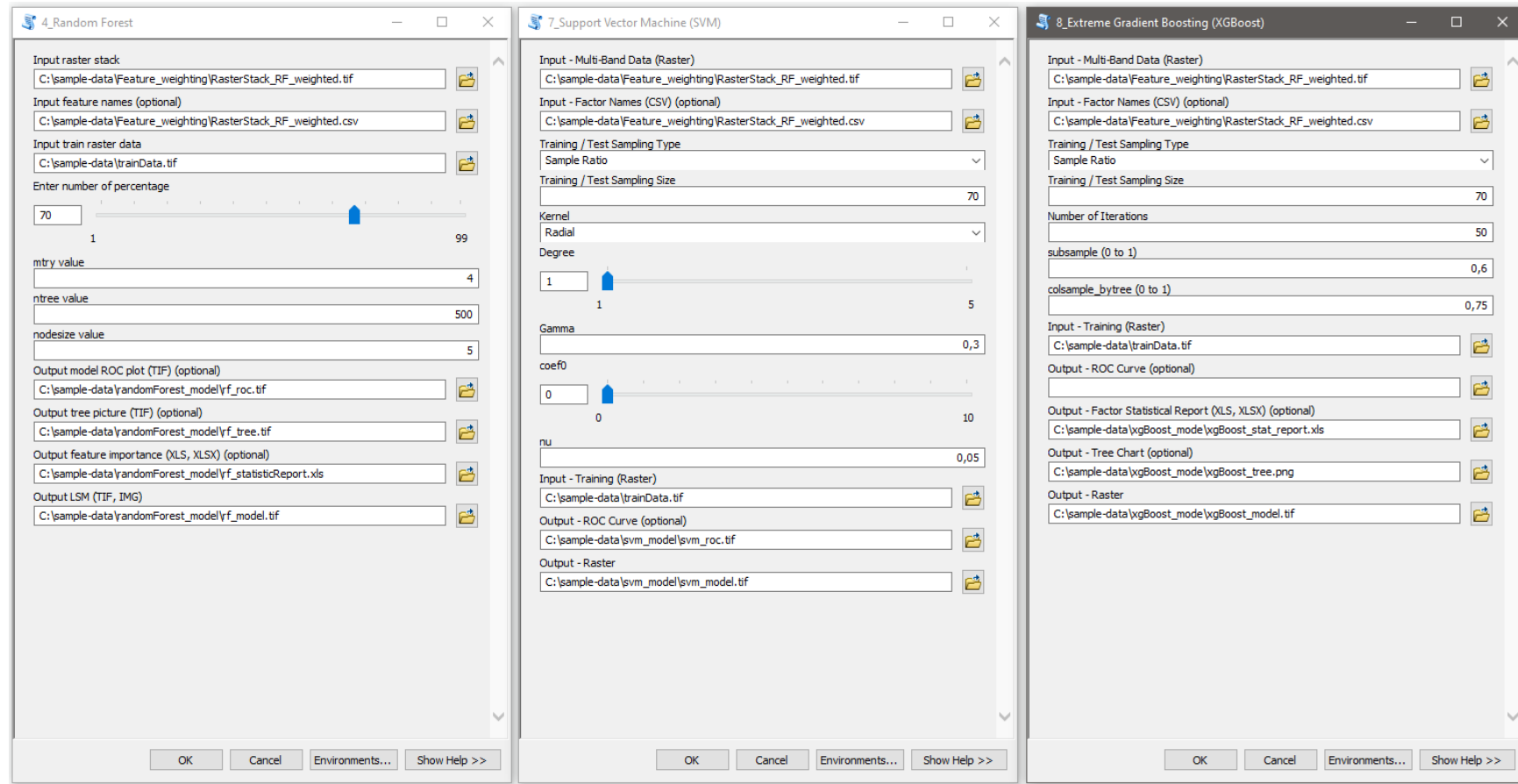


Producing LSM using LSM Tool Pack v1.0

- Three basic ML methods namely Extreme Gradient Boosting, Random Forest, and Support Vector Machine can be used for modeling.

- Users easily can be reached at LSM Tool Pack v1.0 on the given Github webpage.

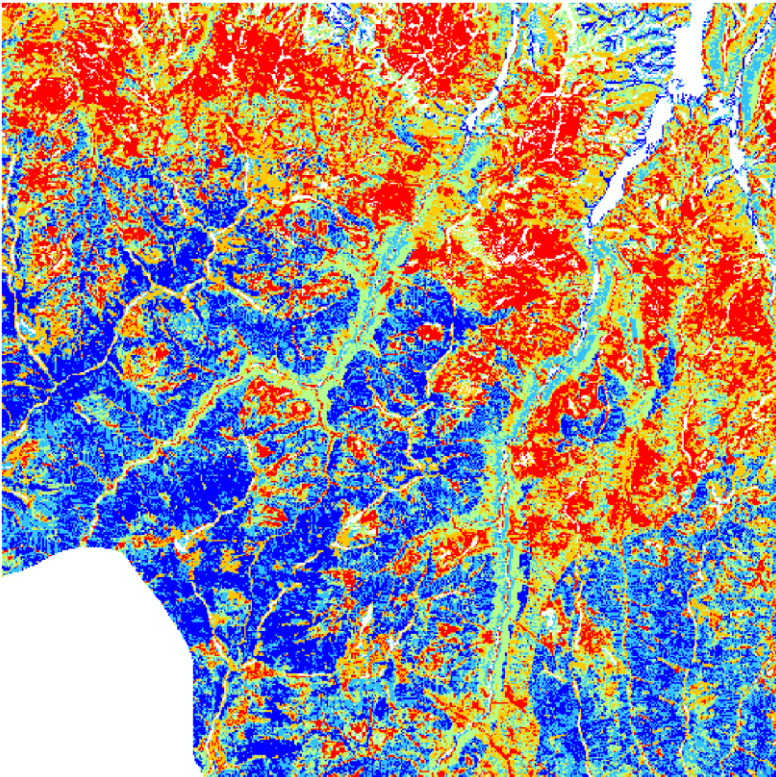
https://github.com/emrehanks/R-ArcGIS-LSM_ToolPack



Producing LSM using LSM Tool Pack v1.0

- Output of 4 Random Forest module results

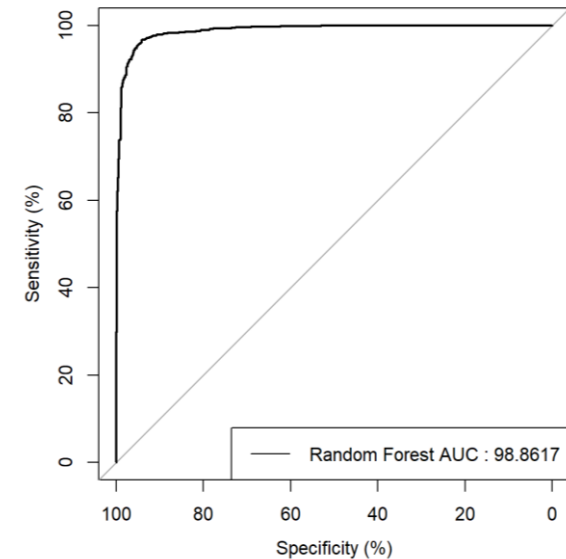
LS Map



Scores of feature importance

	IncNodePurity
splitaspect	46,69777
splitelevation	104,6985
splitlithology	158,4259
splitlulc	43,22014
splitslope	317,4535
splitslopelenght	27,36402
splittpi	36,46575
splittwi	94,80457

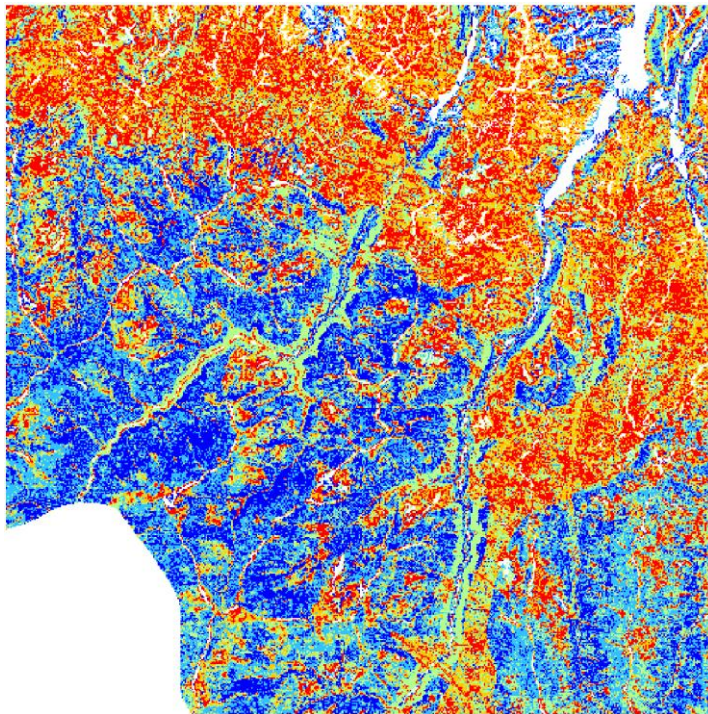
AUC-ROC



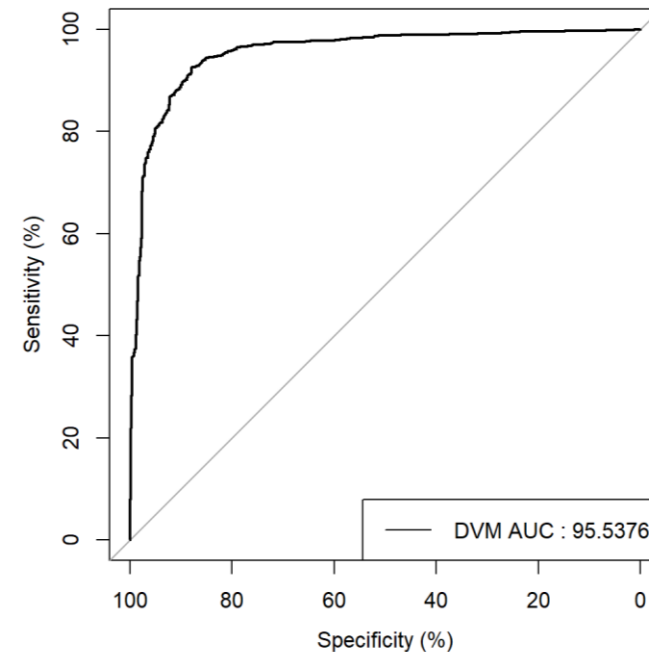
Producing LSM using LSM Tool Pack v1.0

- Output of 7 Support Vector Machine (SVM) module results

LS Map



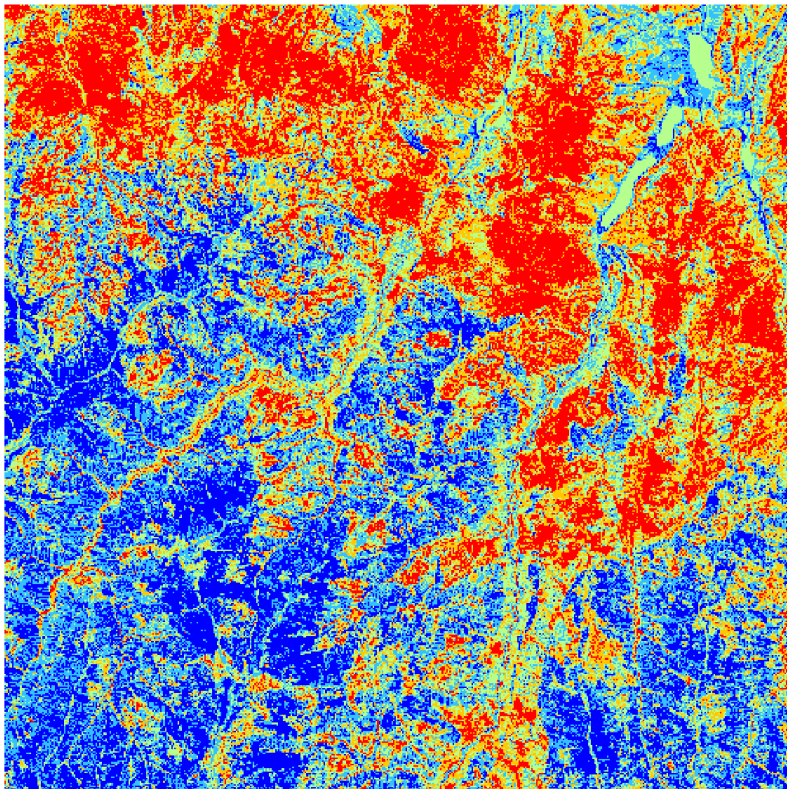
AUC-ROC



Producing LSM using LSM Tool Pack v1.0

- Output of 8 Extreme Gradient Boosting (XGBoost) module results

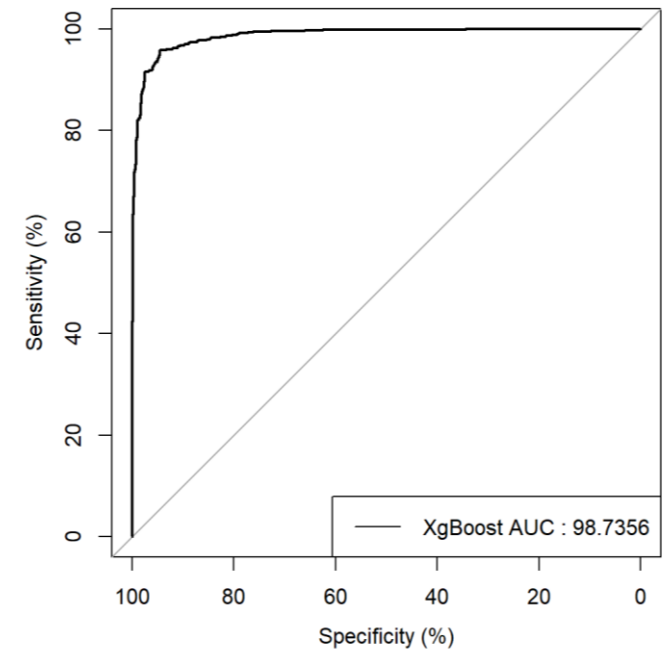
LS Map



Scores of feature importance

	Feature	Gain	Cover	Frequency
1	splitslope	0,314797967	0,179756685	0,125998771
2	splitlithology	0,175911772	0,114584788	0,077443147
3	splitelevation	0,162048064	0,170299626	0,175169023
4	splittwi	0,127587305	0,152884603	0,118623233
5	splitaspect	0,06697848	0,121317078	0,170866626
6	splitlulc	0,059982692	0,105898053	0,094652735
7	splittpi	0,049694773	0,091197581	0,145666872
8	splitslopelenght	0,042998946	0,064061585	0,091579594

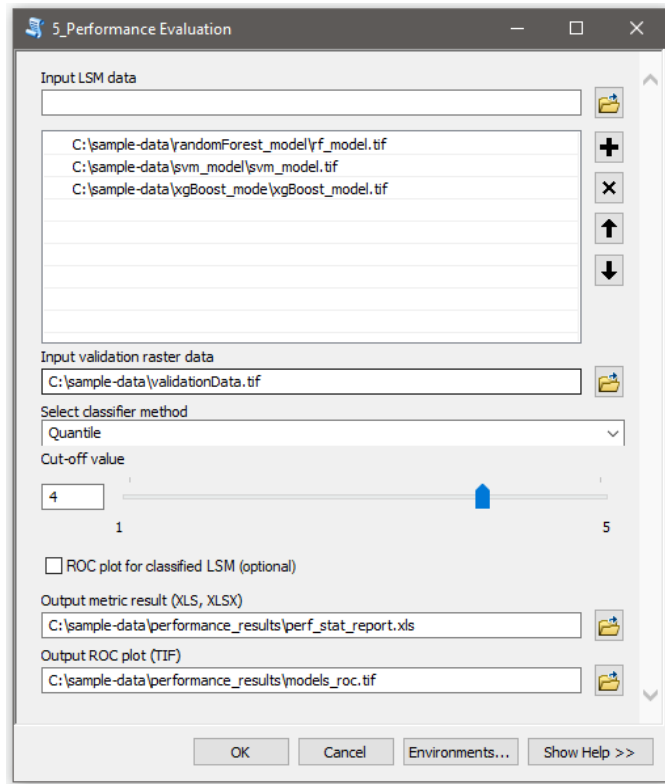
AUC-ROC



Producing LSM using LSM Tool Pack v1.0

- Output of 5_Performance Evaluation module results

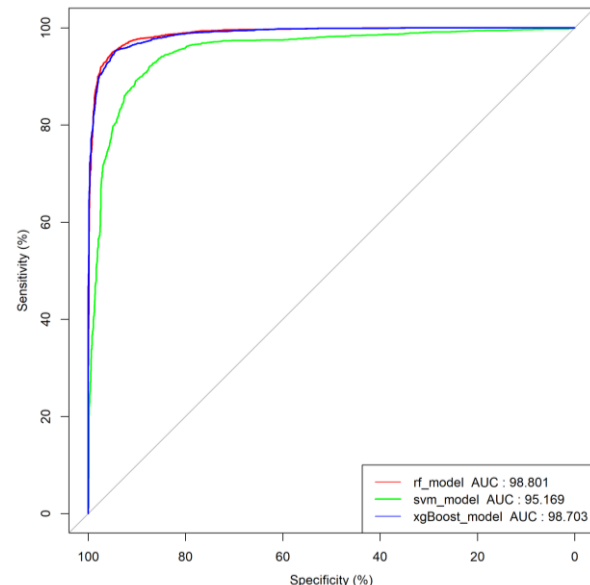
5_Performance Evaluation Module GUI



Performance results of three models

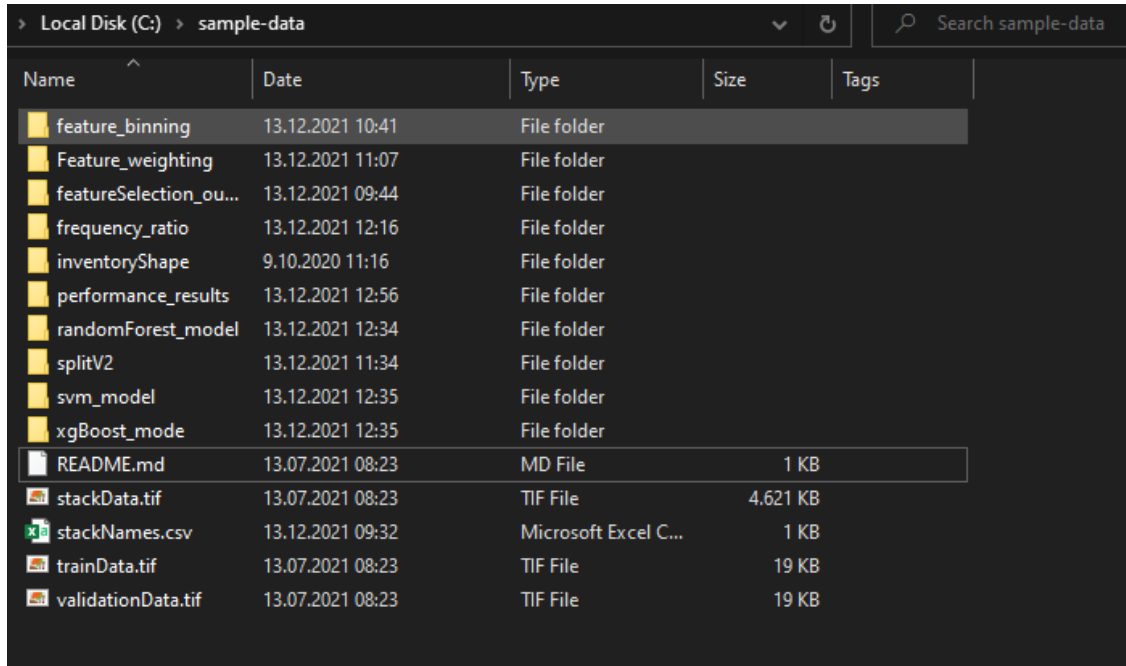
	Accuracy	AUC.Class	AUC.NonC	MAE	RMSE	Kappa	Precision	Recall	F1
xgBoost_model	0,909594	0,925583	0,987035	0,090406	0,300677	0,811683	0,987911	0,870542	0,92552
rf_model	0,890765	0,91232	0,988009	0,109235	0,330507	0,775588	0,991235	0,838119	0,90827
svm_model	0,8552	0,873667	0,951688	0,1448	0,380526	0,702629	0,959145	0,810097	0,878343

Comparison ROC graph of models



Results

- For this sample study, XGBoost model constructed with RF-selected subset was superior to other models.
- All output was published on article supplementary data folder.



Name	Date	Type	Size	Tags
feature_binning	13.12.2021 10:41	File folder		
Feature_weighting	13.12.2021 11:07	File folder		
featureSelection_ou...	13.12.2021 09:44	File folder		
frequency_ratio	13.12.2021 12:16	File folder		
inventoryShape	9.10.2020 11:16	File folder		
performance_results	13.12.2021 12:56	File folder		
randomForest_model	13.12.2021 12:34	File folder		
splitV2	13.12.2021 11:34	File folder		
svm_model	13.12.2021 12:35	File folder		
xgBoost_mode	13.12.2021 12:35	File folder		
README.md	13.07.2021 08:23	MD File	1 KB	
stackData.tif	13.07.2021 08:23	TIF File	4.621 KB	
stackNames.csv	13.12.2021 09:32	Microsoft Excel C...	1 KB	
trainData.tif	13.07.2021 08:23	TIF File	19 KB	
validationData.tif	13.07.2021 08:23	TIF File	19 KB	

Link:

LSM Tool Pack v1.0

https://github.com/emrehanks/R-ArcGIS-LSM_ToolPack

Semi-Automatic Feature Engineering Toolbox: FE.LSM Tool Pack

https://github.com/emrehanks/R-ArcGIS_Feature-Engineering_ToolPack

Emrehan Kutlug Sahin, Ismail Colkesen, Suheda Semih Ac mali, Aykut Akgun, Arif Cagdas Aydinoglu, Developing comprehensive geocomputation tools for landslide susceptibility mapping: LSM tool pack, Computers & Geosciences, 2020, 104592, ISSN 0098-3004, <https://doi.org/10.1016/j.cageo.2020.104592>. (<http://www.sciencedirect.com/science/article/pii/S009830042030577X>)