

# Multi-dimensional visualization for prediction models of geologic hazard

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**[Click here to start](#)**

## **Table of Contents**

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[Multi-dimensional visualization for prediction models of geologic hazard](#) **Email:** [fabbri@itc.nl](mailto:fabbri@itc.nl)

[Multi-dimensional visualization for prediction models of geologic hazard](#)

[Definition: Hazard vs. Prediction](#)

[Tsitika Creek Pilot Study British Columbia, Canada](#)

[PPT Slide](#)

[Themes derived from DEM](#)

[Year of study and validation data](#)

[Training and Target Spatial Data](#)

[Three Mathematical Frameworks for Prediction Models](#)

[Probability interpretation](#)

[Probability interpretation if we have future landslides](#)  
[F, we obtain the following exact formula](#)

[Predicted value ranges](#)

[Visualization of predicted values](#)

[PPT Slide](#)

[From 2D to 2.5D to 3D](#)

## [Prediction Draped on Shaded Relief](#)

### [3D Perspective - Inset 1](#)

### [3D Perspective - Inset 2](#)

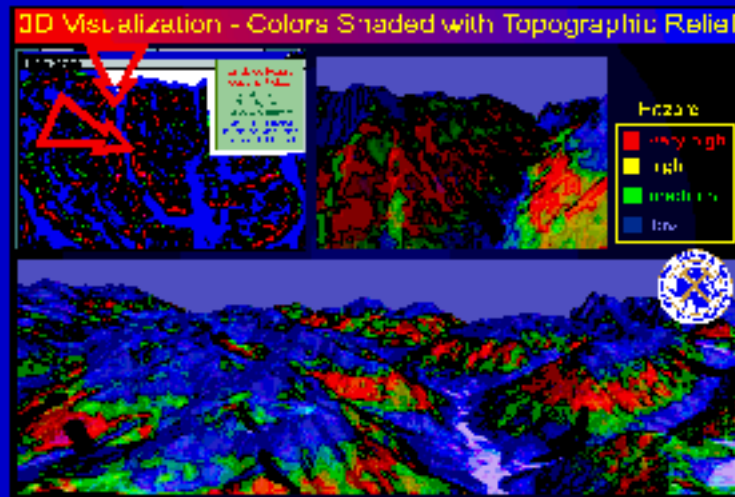
## [Importance of visualization](#)

## [Further considerations](#)



# Multi-dimensional visualization for prediction models of geologic hazard

*Chang-Jo F. Chung<sup>1</sup>, Andrea G. Fabbri<sup>2</sup>, David F. Garson<sup>1</sup>*



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<sup>2</sup>International Institute for Aerospace Surveys and Earth Sciences, ITC, Enschede, The Netherlands



## Multi-dimensional visualization for prediction models of geologic hazard

- hazard and prediction
- study area
- spatial data
- mathematical frameworks
- 2D to 2.5D to 3D
- importance of visualization



## Definition: Hazard vs. Prediction

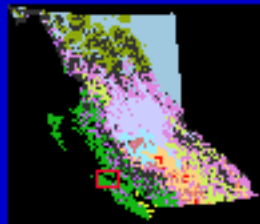
- **Natural hazard** is the probability of **FUTURE** occurrence, within a specific period of time and within a given area, of a potentially damaging phenomenon (e.g., a given type of mass movement)
- **Prediction** is finding areas, reasonably small sub-areas, where **FUTURE** occurrences of a potentially damaging phenomenon are likely to be located (e.g., the highest values of probability of occurrence of a given type of mass movement)



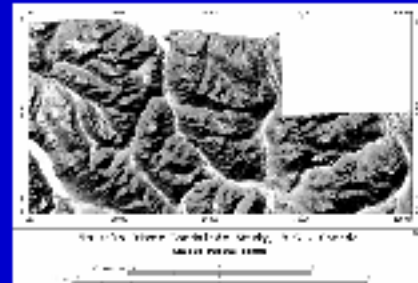
# Tsitika Creek Pilot Study British Columbia, Canada



*Legal requirement for forestry companies to  
conduct landslide hazard studies before  
logging permit*

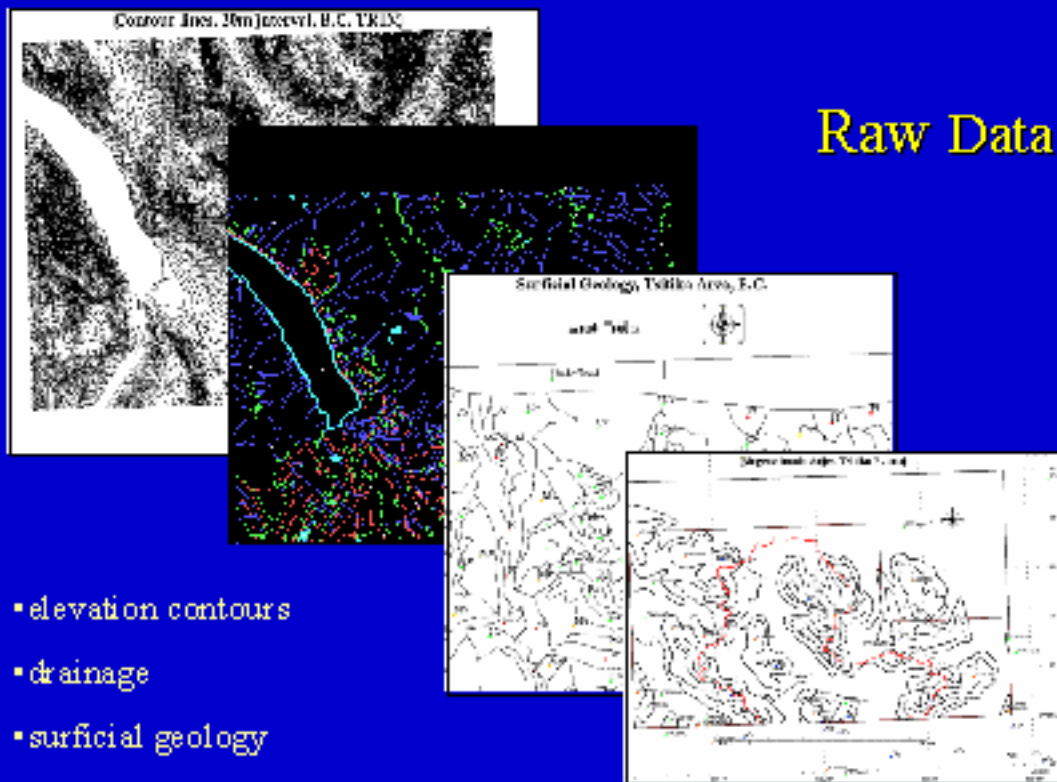


*Logistical requirement for automated  
prediction of landslide hazard potential*





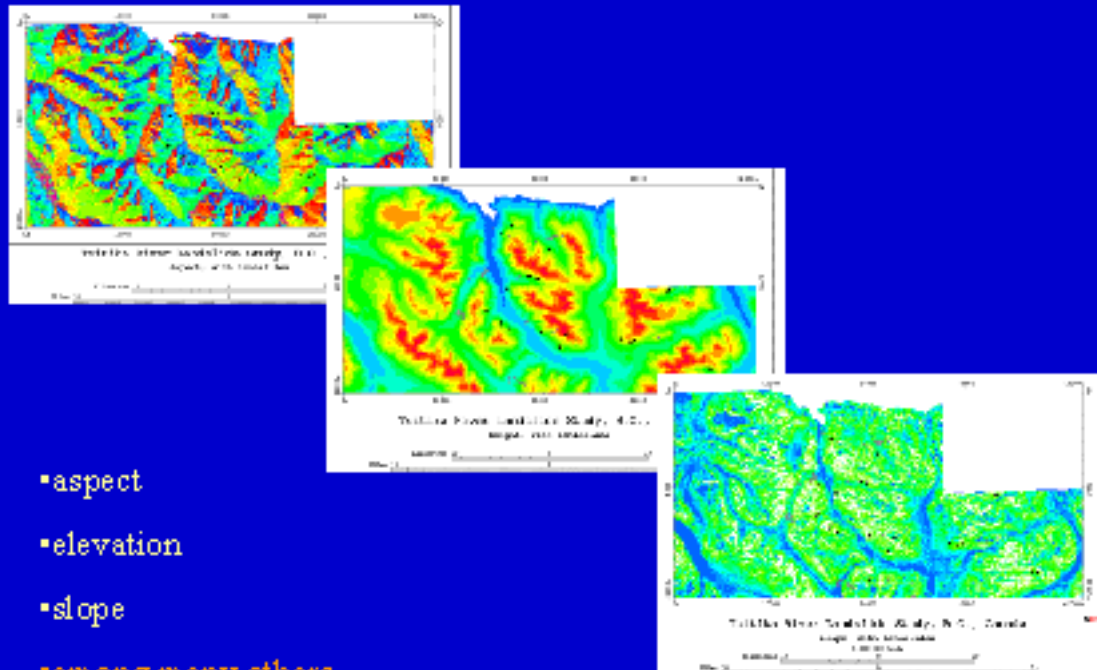
## Raw Data



- elevation contours
- drainage
- surficial geology
- biogeoclimatic zones
- digital form



## Themes derived from DEM



Slide 6 of 20

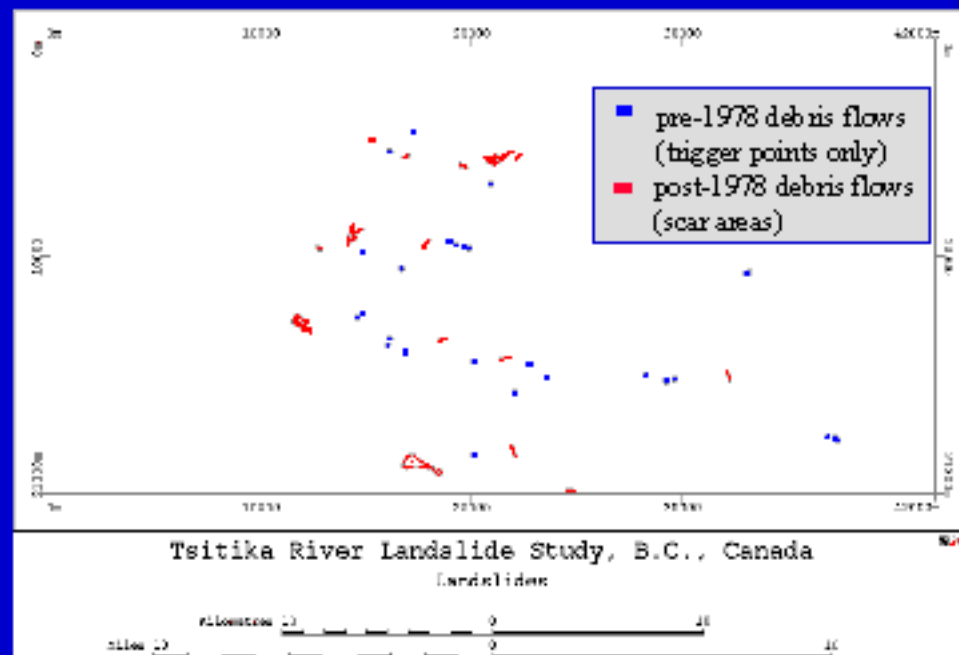


## Year of study and validation data

- Year of study is **1978**
- A geomorphologist provided the distribution of the trigger points of debris flows up to **1978**
- A **1996** study provided the distribution of debris flows scars between **1979 - 1996**
- *The distribution of debris flow scars (1979 - 1996) was used to validate the predictions*



## Training and Target Spatial Data



Slide 8 of 20



## Three Mathematical Frameworks for Prediction Models

1. Probability theory
2. Dempster-Shafer's Evidential Theory
3. Zadeh's Fuzzy Set Theory



# Probability interpretation

$f(T_p \text{ Given } \mathbf{m} \text{ evidences } v_k(p), k=1, \dots, \mathbf{m})$

$T_p$ : “ $p$  will be affected by a future landslide of type  $\mathbf{D}$ ”

Measurement of the “sureness” that the proposition  $T_p$   
 ( $p$  will be affected by a **future landslide** of type  $\mathbf{D}$ ) is  
 (likely) true, given the  $m$  evidences ( $v_k(p)$ ,  $k=1, \dots, \mathbf{m}$ ) at  $p$ .

“**sureness**”, probability, certainty, belief, **plausibility**, **possibility**...

$$f(T_p | v_1(p), v_2(p), \dots, v_k(p)) = Prob(T_p | v_1(p), v_2(p), \dots, v_k(p))$$



# Probability interpretation

if we have **future** landslides **F**,  
we obtain the following exact formula

$$\begin{aligned}
 & \textit{Prob} \{ T_p \mid v_1(p)=c_1, \dots, v_m(p)=c_m \} \\
 &= \textit{Prob} \{ p \in F \mid p \in \bigcap_{k=1}^m A_{kc_k} \} \\
 &= \textit{Prob} \{ p \in \bigcap_{k=1}^m (F \cap A_{kc_k}) \mid p \in \bigcap_{k=1}^m A_{kc_k} \} \\
 &= \text{size of } \left( \bigcap_{k=1}^m (F \cap A_{kc_k}) \right) / \text{size of } \left( \bigcap_{k=1}^m A_{kc_k} \right)
 \end{aligned}$$

There are many different ways to estimate this probability formula

(Chung and Fabbri, 1995, 1996a and 1996b)



## Predicted value ranges

- discrete values between a min. and a max.  
(most additive or multiplicative indices)
- continuous ranges between 0.0 and 1.0  
(Probability, belief, Fuzzy membership)
- continuous ranges between -1.0 and +1.0  
(Certainty factor)

**MAPS OF PREDICTED VALUES HAVE  
TO BE CLASSIFIED: ART FORM?**

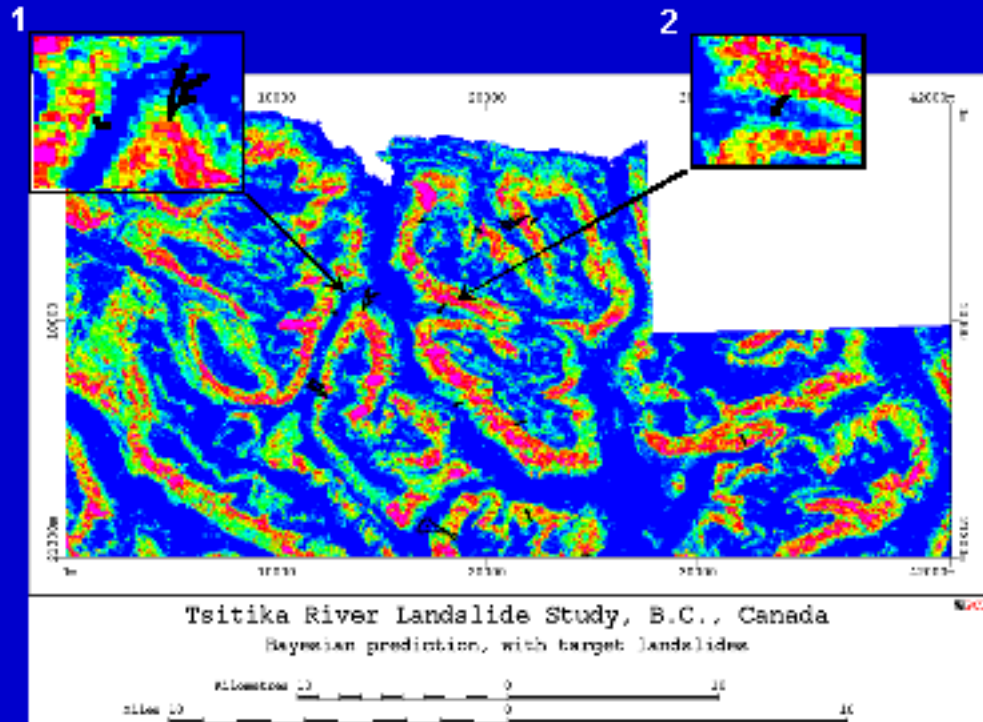


## Visualization of predicted values

- sort all predicted values in decreasing order
- perform percentile analysis of the values according to the % of study area occupied
- use pseudo-color LUT for subsequent intervals (e.g., 1% or 5%)
- validate prediction by comparing the % classes of prediction with the corresponding % of “future landslides”



## 2D Representation of Prediction

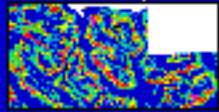


Slide 14 of 20



## From 2D to 2.5D to 3D

Color table model (PCT)  
of prediction map



PCT to RGB

RGB  
model



RGB to IHS



Intensity based on  
shaded relief



Hue from IHS  
model of RGB

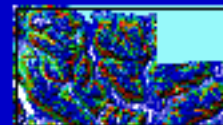


Saturation grey-level  
user-selected (i.e. 175)

IHS to RGB

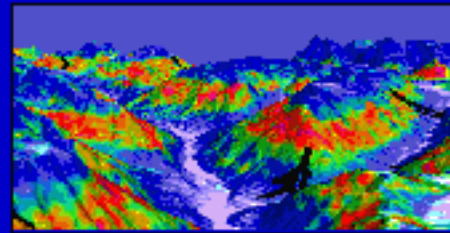
RGB model,  
perspective view

2.5D



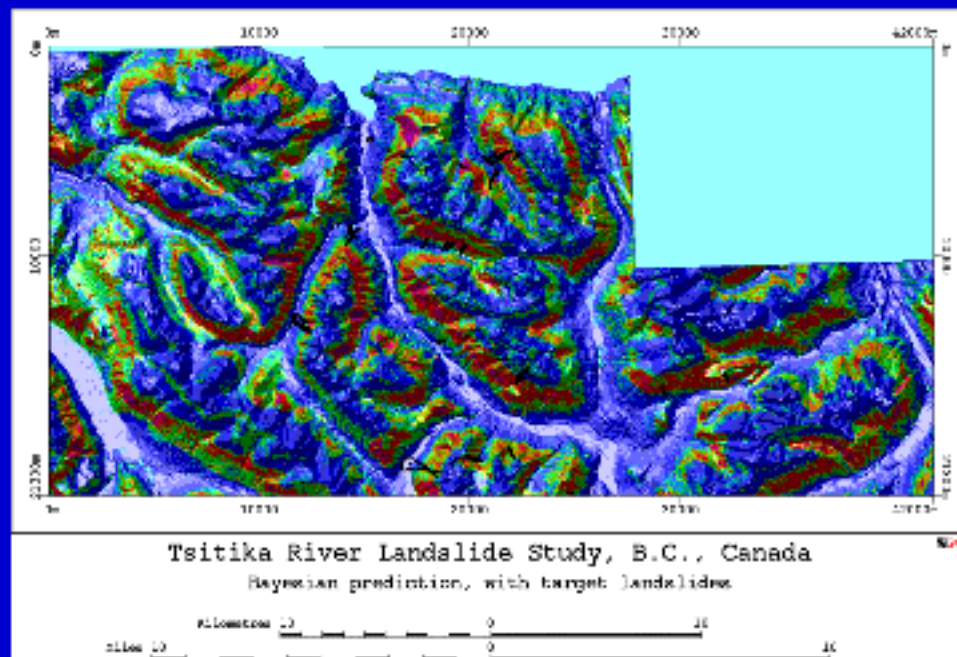
RGB model,  
overhead view

3D



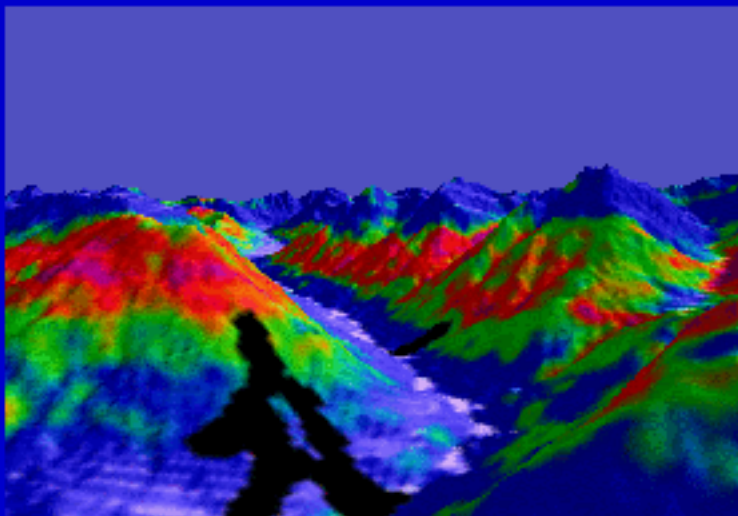


## Prediction Draped on Shaded Relief





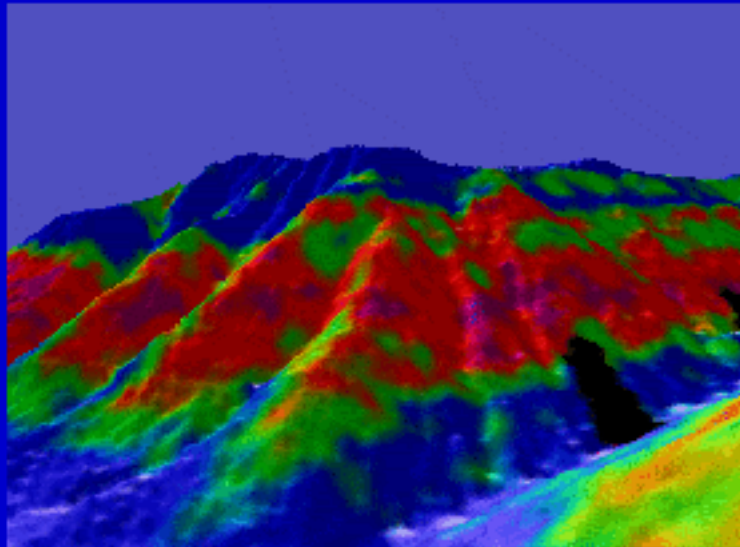
## 3D Perspective - Inset 1



Slide 17 of 20



## 3D Perspective - Inset 2



Slide 18 of 20



## Importance of visualization

- Predictions generate adimensional values
- Experts tend to prefer arbitrary slicing of the index values, possibly associated with conventional color schemes:  
THIS MAKES COMPARISONS DIFFICULT.
- Areal ranking of predictions and of predicted values allows to “see” before classifying.
- Even if percentile analysis of predicted value ranges versus the area percentage of future or later events is made,  
THE FLAT REPRESENTATION IS INSUFFICIENT!
- The enhanced 3D visualization does justice to the implicit dynamism of the mass movement (FLY THROUGH).
- The critical features to evaluate a prediction are  
THE TRIGGER POINTS



## Further considerations

- When slicing a sorted set of predicted hazard values, we must remember that the assumptions implicit in the prediction models (e.g., probability, belief) may not permit “**direct interpretation**”.
- “**Relative interpretations**” and comparisons between predictions, however, can still be made due to the *maintained relative importance* of the predicted values.
- The 3D displays can be considered as **VALUE-ADDED MAPS**: they provide additional information on hazard predictions.

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[Previous slide](#)

[Back to first slide](#)

[View graphic version](#)

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## THE TRIGGER POINTS

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# 3D Perspective - Inset 2

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# 3D Perspective - Inset 1

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# Prediction Draped on Shaded Relief

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# From 2D to 2.5D to 3D

RGB

model

3D

2.5D

Intensity based on

shaded relief

Hue from IHS

model of RGB

Saturation grey-level

user-selected (i.e. 175)

IHS to RGB

PCT to RGB

RGB to IHS

RGB model,

overhead view

RGB model,

perspective view

Color table model (PCT)

of prediction map

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# PPT Slide

2D Representation of Prediction

1

2

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# Visualization of predicted values

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[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

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[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# Probability interpretation if we have future landslides F, we obtain the following exact formula

$$\begin{aligned} & \text{Prob}\{ T_p \mid v_1(p)=c_1, \dots, v_m(p)=c_m \} \\ &= \text{Prob}\{ p \in F \mid p \in A_k \} \\ &= \text{Prob}\{ p \in (F \cap A_k) \mid p \in A_k \} \\ &= \text{size of } \{ (F \cap A_k) \} / \text{size of } (A_k) \end{aligned}$$

There are many different ways to estimate this probability formula

(Chung and Fabbri, 1995, 1996a and 1996b)

U

m

k=1

U

m

k=1

k=1

U

U

U

U

m

U

m

k=1

m

$k=1$

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# Probability interpretation

Measurement of the “sureness” that the proposition  $T_p$

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“sureness”, probability, certainty, belief, plausibility, possibility...

$f(T_p \mid v_1(p), v_2(p), \dots, v_k(p)) = \text{Prob}\{T_p \mid v_1(p), v_2(p), \dots, v_k(p)\}$

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# Three Mathematical Frameworks for Prediction Models

1. Probability theory
2. Dempster-Shafer's Evidential Theory
3. Zadeh's Fuzzy Set Theory

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# Training and Target Spatial Data

pre-1978 debris flows

(trigger points only)

post-1978 debris flows

(scar areas)

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# Year of study and validation data

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- . The distribution of debris flow scars (1979 - 1996) was used to validate the predictions**

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# Themes derived from DEM

- aspect
- elevation
- slope
- among many others

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# PPT Slide

## Raw Data

- elevation contours
- drainage
- surficial geology
- biogeoclimatic zones
- digital form

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# Tsitika Creek Pilot Study British Columbia, Canada

Legal requirement for forestry companies to conduct landslide hazard studies before logging permit

Logistical requirement for automated prediction of landslide hazard potential

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# Definition: Hazard vs. Prediction

- . Prediction is finding areas, reasonably small sub-areas, where FUTURE occurrences of a potentially damaging phenomenon are likely to be located (e.g., the highest values of probability of occurrence of a given type of mass movement)**
- . Natural hazard is the probability of FUTURE occurrence, within a specific period of time and within a given area, of a potentially damaging phenomenon (e.g., a given type of mass movement)**

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)



# **Multi-dimensional visualization for prediction models of geologic hazard**

- . hazard and prediction**
- . study area**
- . spatial data**
- . mathematical frameworks**
- . 2D to 2.5D to 3D**
- . importance of visualization**

[Previous slide](#)

[Next slide](#)

[Back to first slide](#)

[View graphic version](#)

# Multi-dimensional visualization for prediction models of geologic hazard

**Chang-Jo F. Chung<sup>1</sup>, Andrea G. Fabbri<sup>2</sup>, David F. Garson<sup>1</sup>**

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[Next slide](#)

[Back to first slide](#)

[View graphic version](#)