



# Information Entropy Monte Carlo Simulation

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# Outline

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- **Shannon's Information Entropy**
- **Monte Carlo Simulation**

# Which attribute(s) should We Use ?

## Innumerable Seismic Attributes

Type	Seismic Attribute	Major Geological significance
Conventional property	Amplitude	Lithological contrast, bedding continuity
	Interval velocity	Lithology, Porosity, Fluid Content
	Acoustic impedance	Lithology, Porosity, Fluid Content
Volume-related attribute (Multi-trace attribute)	Reflection geometry	Reservoir Architecture, Sedimentary Structure
	Trace continuity	Fault geometry, Fault distribution, Stratigraphic continuity
	Time curvature, Dip, Azimuth	Detailed Reservoir Architecture, Fault geometry, Fault distribution, Fracture density
Pre-stack attribute	AVO	Fluid Content, Lithology, Porosity
	Impedance (elastic/S-wave)	
	Poisson's ratio	
	$\lambda, \mu$ (Lame constant)	
	AVOZ	Fracture Orientation, Fracture Density, Fluid Content
Instantaneous attribute	Instantaneous phase	Bedding continuity
	Instantaneous Frequency	Bed thickness, lithologic contrast, fluid content
Miscellaneous attribute	Frequency	Fluid content
	Attenuation	
	Anything computed from seismic traces	???

For the purposes of predicting porosity, which attributes should we use?

- $AVO$   $P_0$  &  $G$
- $AI$  &  $EI$
- $\lambda\rho$  &  $\mu\rho$
- $Vp/Vs$  etc

Shannon's Information Entropy can give us the solution quantitatively.

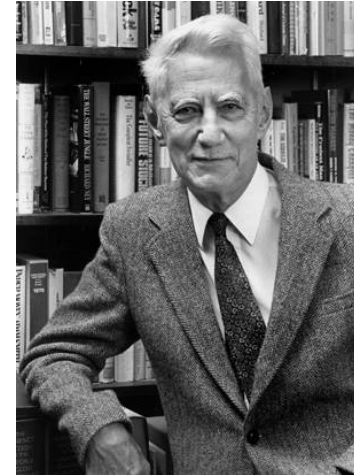
# Shannon's Information Theory

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*Shannon and Weaver (1949)*  
*"The Mathematical Theory of Communication"*

Defined **Quantity of Information**

- **Information Content**
- **Information Entropy**
- **Mutual Information**



Dr. Claude Shannon

# Information Content

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*It will snow !!*

## Houston

*Low Probability*

Very rare (1%)



Very surprised !!

(valuable information)

$$I_{\text{snow}} = 6.64$$

## Calgary

*High Probability*

Very often (70%)



Not surprised

(Not-valuable information)

$$I_{\text{snow}} = 0.51$$

**Information Content** (Quantity of information)

$$I = -\log P$$

How surprised one would be if the event happened.

\* Base = 2

# Information Entropy

Expected Value of Information Content

## Information Entropy

$$H(X) = -\sum_i^n P_i \cdot \log P_i$$

$$X = \{x_1, \dots, x_N\}$$

- **Expected Surprise**
- Quantity of **uncertainty** associated with  $P$

### Houston

$$H=1.65$$

$X_i$	$P_i$ (%)
Sunny	33
Cloudy	33
Rain	33
Snow	1

### Calgary

$$H=1.36$$

Forecast is easy  
Uncertainty is small

$X_i$	$P_i$ (%)
Sunny	10
Cloudy	10
Rain	10
Snow	70

### NY

$$H=2.00$$

Forecast is difficult  
Uncertainty is large

$X_i$	$P_i$ (%)
Sunny	25
Cloudy	25
Rain	25
Snow	25

# Conditional Entropy

## ■ Conditional Information Entropy

$$H(X | A)$$

Conditional Information Entropy at a given additional information (A)

NY

$$H(X) = 2.00$$

Uncertainty is large

Weather	Pi (%)
Sunny	25
Cloudy	25
Rain	25
Snow	25

e.g. A = atmospheric pressure change data

At given additional information A

$$H(X|A) = 1.54$$

Uncertainties reduce

Weather	Pi (%)
Sunny	50
Cloudy	25
Rain	10
Snow	5

## ■ Mutual Information

$$I(X | A) = H(X) - H(X | A)$$

Quantity of uncertainty reduced by additional information **A**

# Our Case : Porosity Prediction

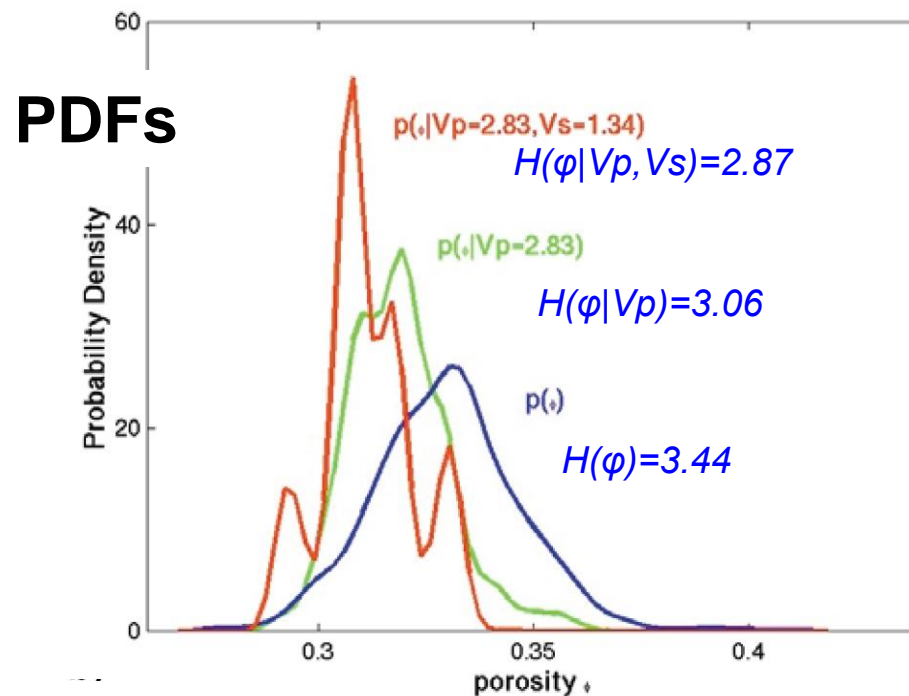
## ■ Porosity Prediction by Seismic Attribute

X : Porosity (continuous variable)

A : Seismic Attribute

$H(X)$  : Information Entropy for Porosity's PDF

$H(X|A)$  : Information Entropy for Porosity's PDF at given seismic attribute



Adding more seismic attributes

PDF shape → Narrow, steep

Uncertainty → Decrease

In. Entropy → Decrease



# Information carried by Seismic Attributes

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For example,  
X = Porosity  
A = Seismic Attribute

Mutual Information

$$I(X | A) = H(X) - H(X | A)$$

Information entropy

Conditional entropy at given A

- Mutual Information can be regarded as the reduced uncertainty by the seismic attributes.
- Thus, we should choose the one which will maximize the mutual information.

# Case Studies

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- North Sea Tertiary Turbiditic Reservoir

- Case I

- Facies Identification  $H(\textit{facies} | \textit{attributes})$

- Case II

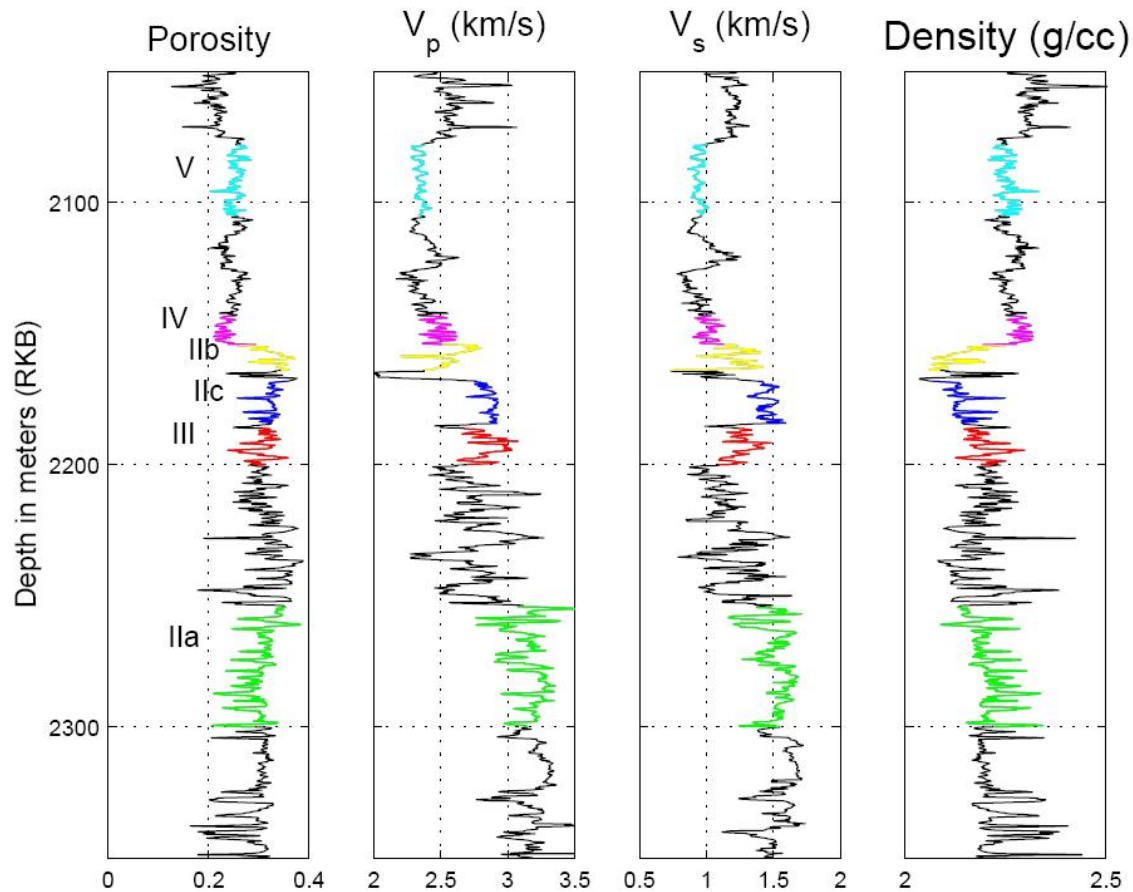
- Pore Fluid Identification  $H(\textit{fluid} | \textit{attributes})$

- References

- Tapan et al. (2001)
  - Takahashi et al (1999)

# Well Log Data

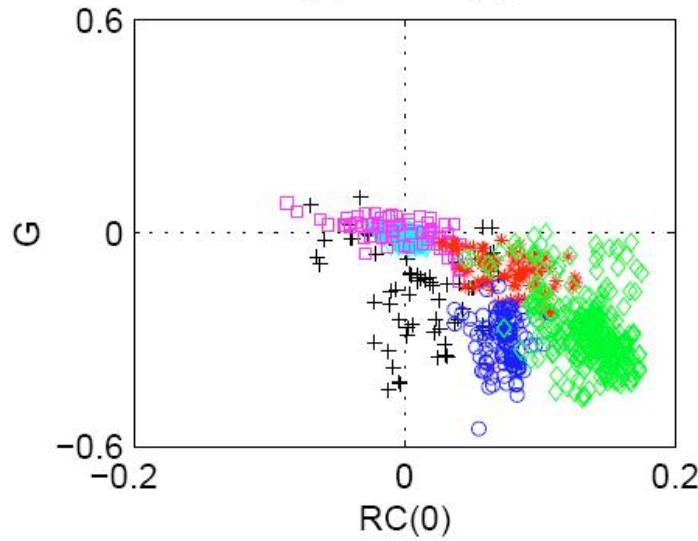
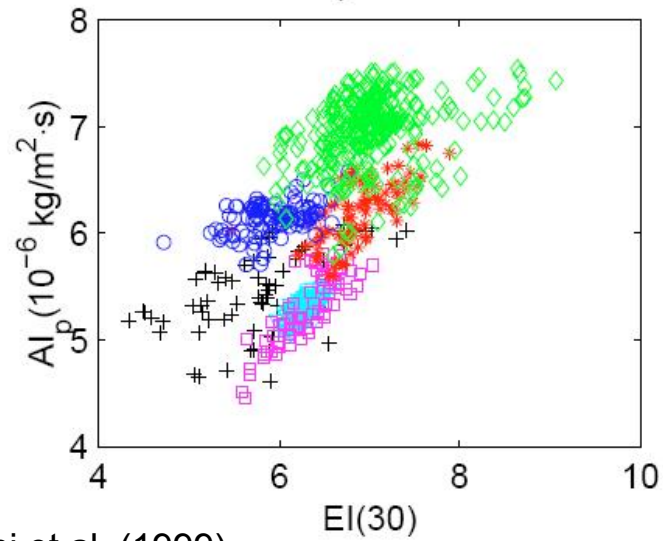
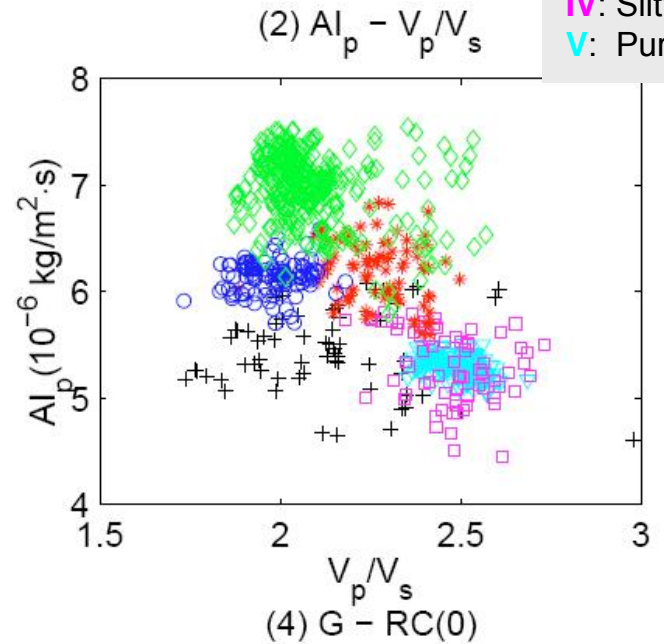
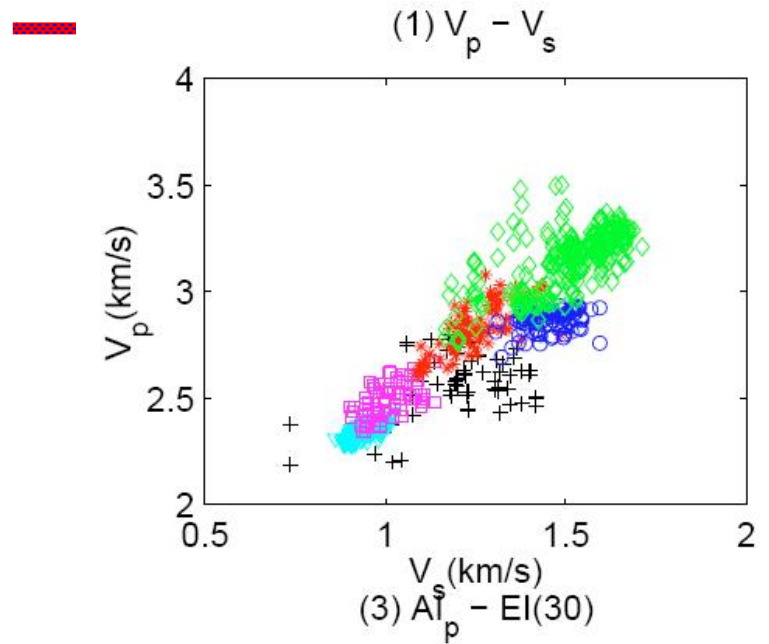
## Facies Classification in Well Log Data



- Ia:** Cemented Sand
- Ib:** Uncololidated Sand
- Ic:** Laminated Sand
- III:** Interbedded Sand-Shale
- IV:** Silty Shale
- V:** Pure Shale

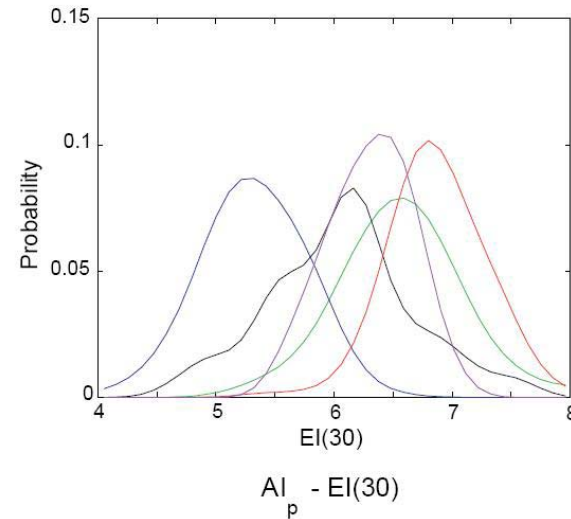
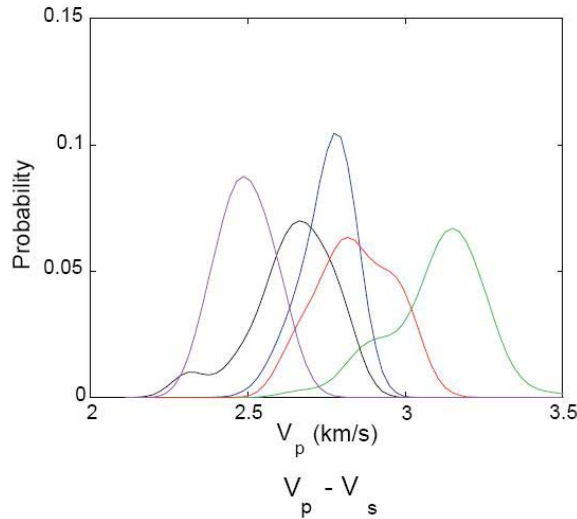
# Crossplot

- IIa: Cemented Sand
- IIb: Uncolodated Sand
- IIc: Laminated Sand
- III: Interbedded Sand-Shale
- IV: Silty Shale
- V: Pure Shale

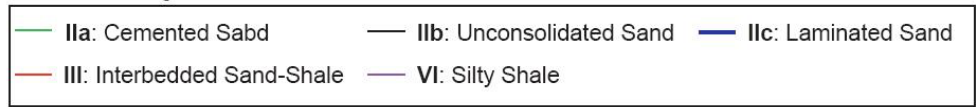
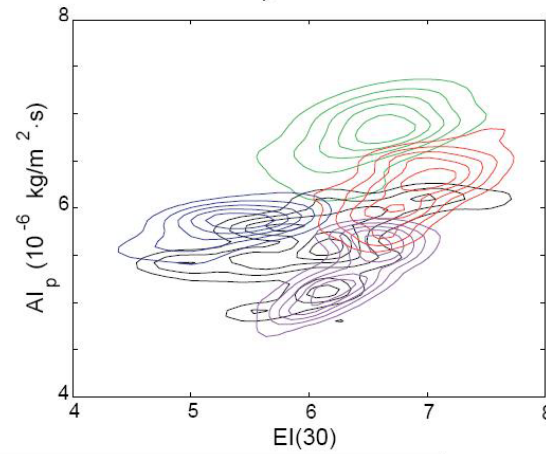
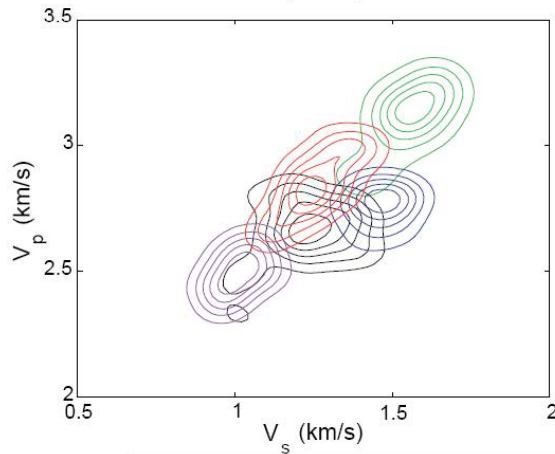


# Conditional probability distributions

## Univariate



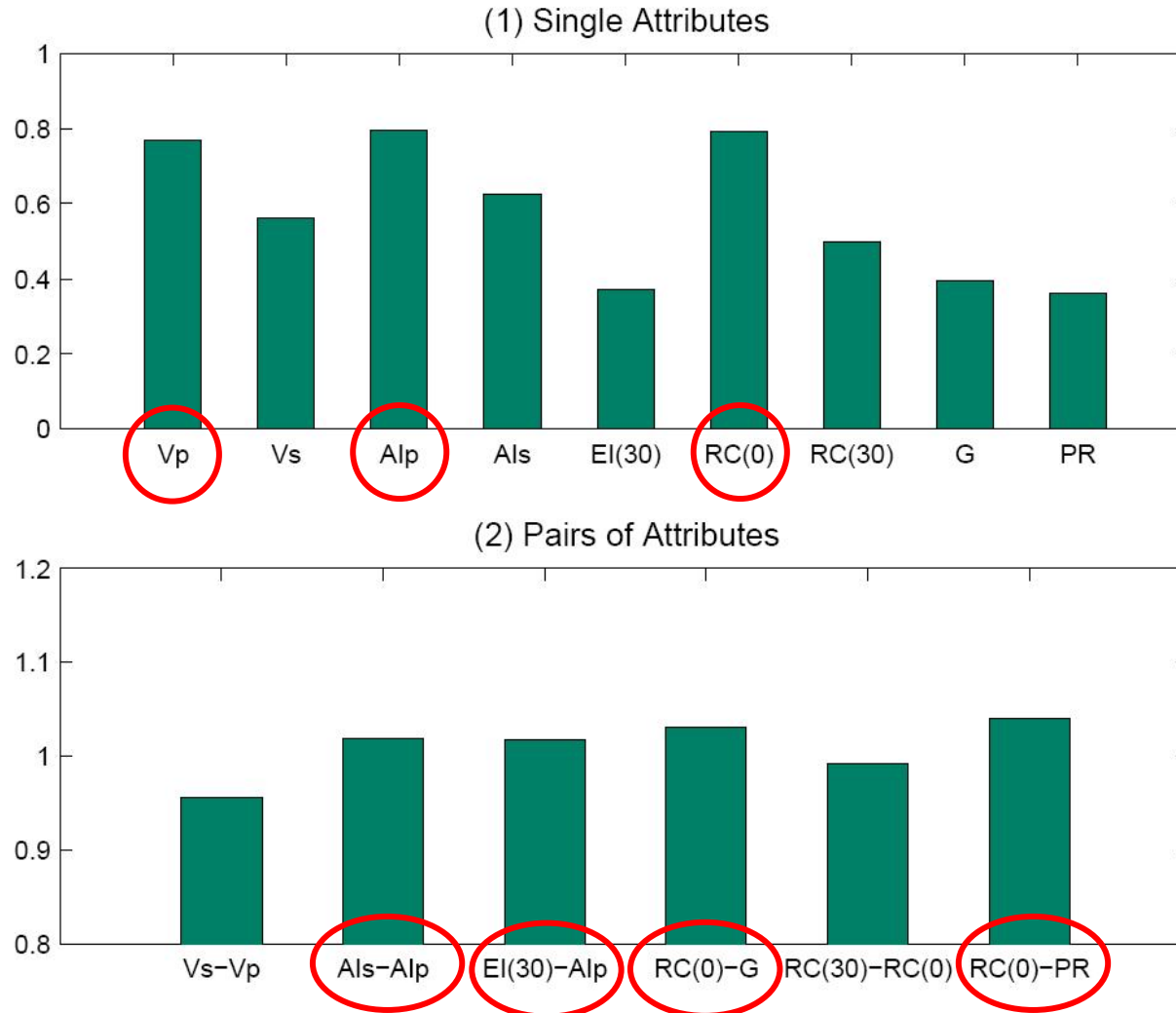
## Bivariate



# Mutual Information

Information about lithofacies carried by Seismic attributes

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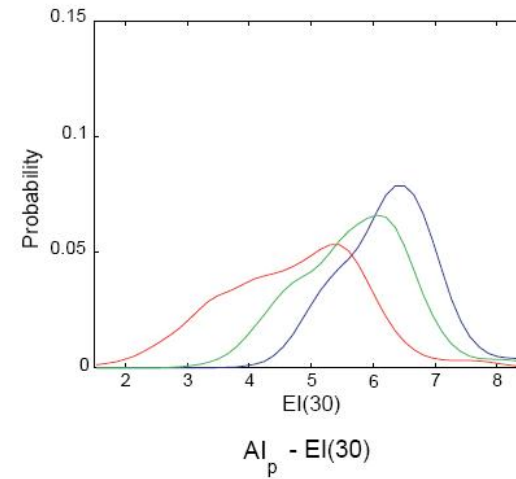
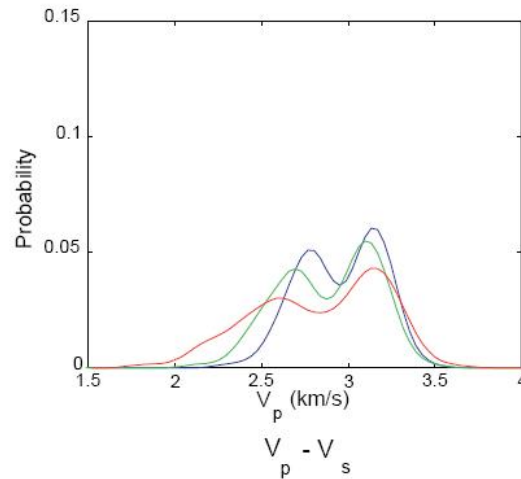


# Case Study 2 (Pore Fluid)

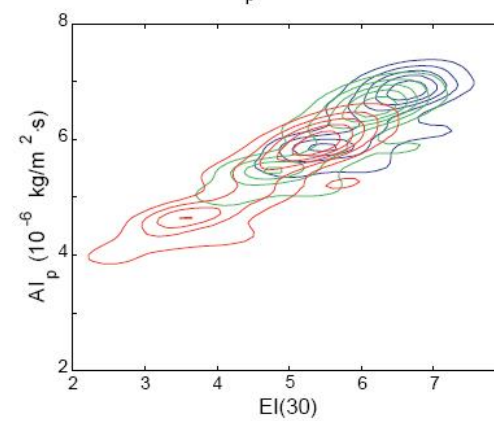
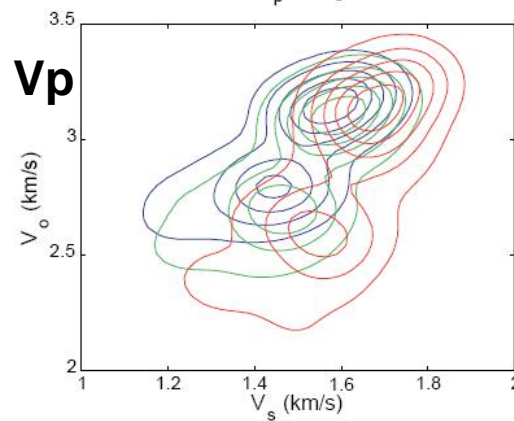
## Extended PDFs

Gassmann eq. was applied for fluid substitution in only sand reservoir

*Univariate*



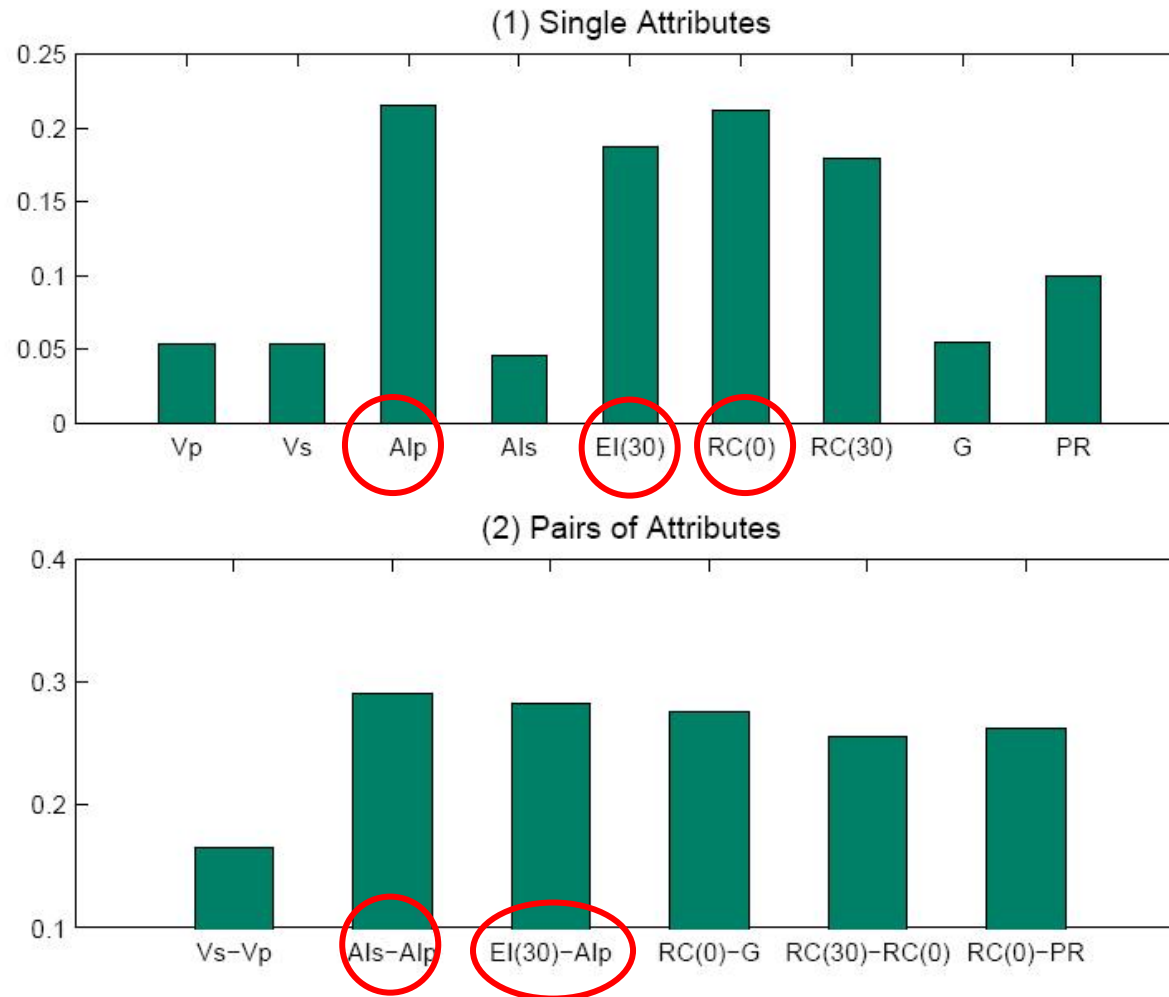
*Bivariate*



# Mutual Information

Information about **pore fluid** carried by Seismic attributes

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# Discussions

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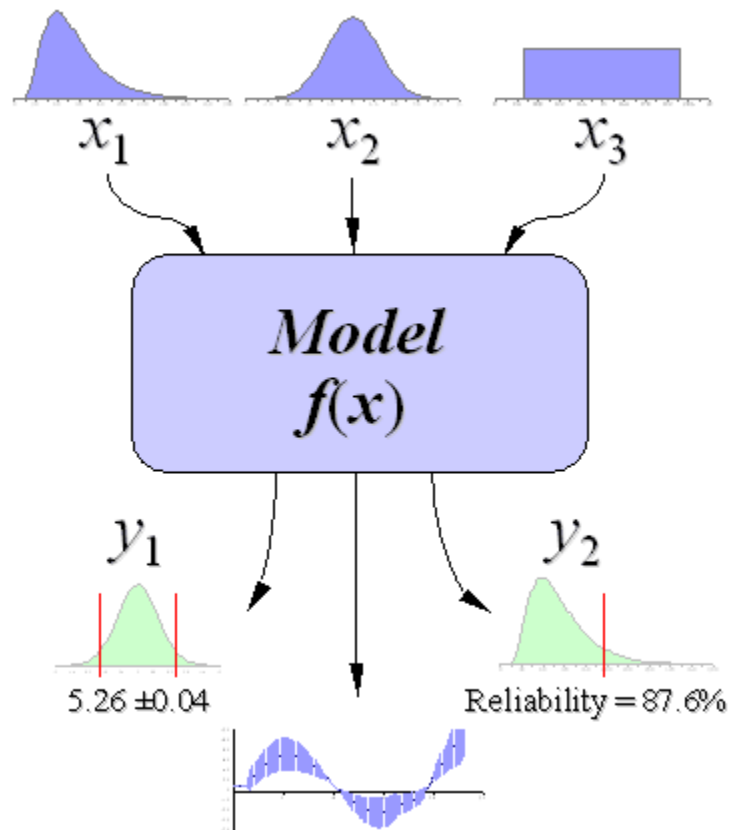
Linear measures of uncertainty, such as variance ( $\sigma^2$ ) and covariance ( $\sigma_{12}$ ), can be used instead of the entropy (H) ?

- Variance (covariance) can work only at limited situation
  - Parametric PDFs, such as Gaussian distribution
  - Continuous variable
- Information Entropy can work more flexibly
  - Nonparametric PDFs
  - Categorical variables (Shale, Sand)

The Entropy offers a more flexible representation of the state of information about the rock.

# Monte Carlo Simulation

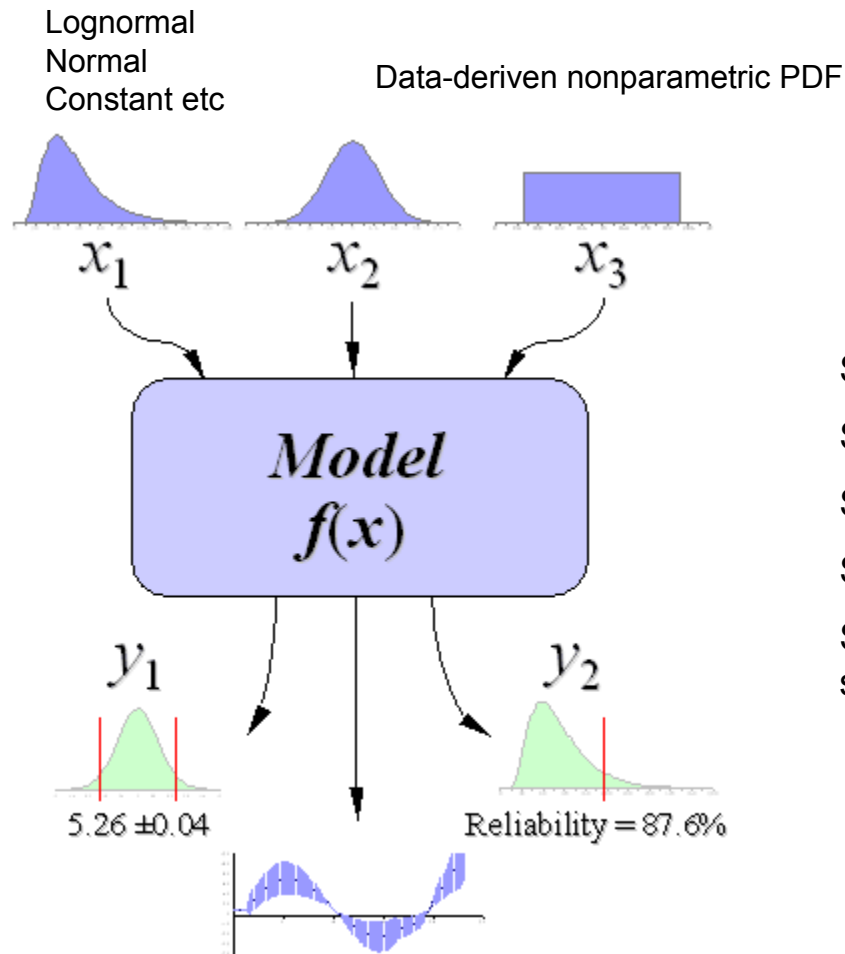
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A technique using random numbers for probabilistic solution of a model

- Model is nonlinear system
- Input parameters with uncertainty
- Uncertainty analysis instead of deterministic method

# Monte Carlo Simulation



Step 1: Create a model,  $y = f(x_1, x_2, \dots, x_q)$ .

Step 2: Generate a set of random inputs,  $x_{i1}, x_{i2}, \dots, x_{iq}$ .

Step 3: Use the model to obtain outputs.

Step 4: Repeat steps 2 and 3 for  $i = 1$  to  $n$ .

Step 5: Analyze the results using histograms, summary statistics, confidence intervals, etc.

<http://www.vertex42.com/ExcelArticles/mc/MonteCarloSimulation.html>

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**Thank you for attentions**