



# Machine Learning:

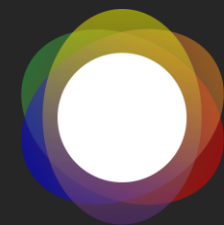
## Going Beyond the Hype and Making it Work for Earth Science



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March 2018  
Manila, Philippines

[www.irayaenergies.com](http://www.irayaenergies.com)



**iraya**  
machine learning • geoscience

# Outlines



AI – Is it Magic, or Math?

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Making AI Work for Earth Science

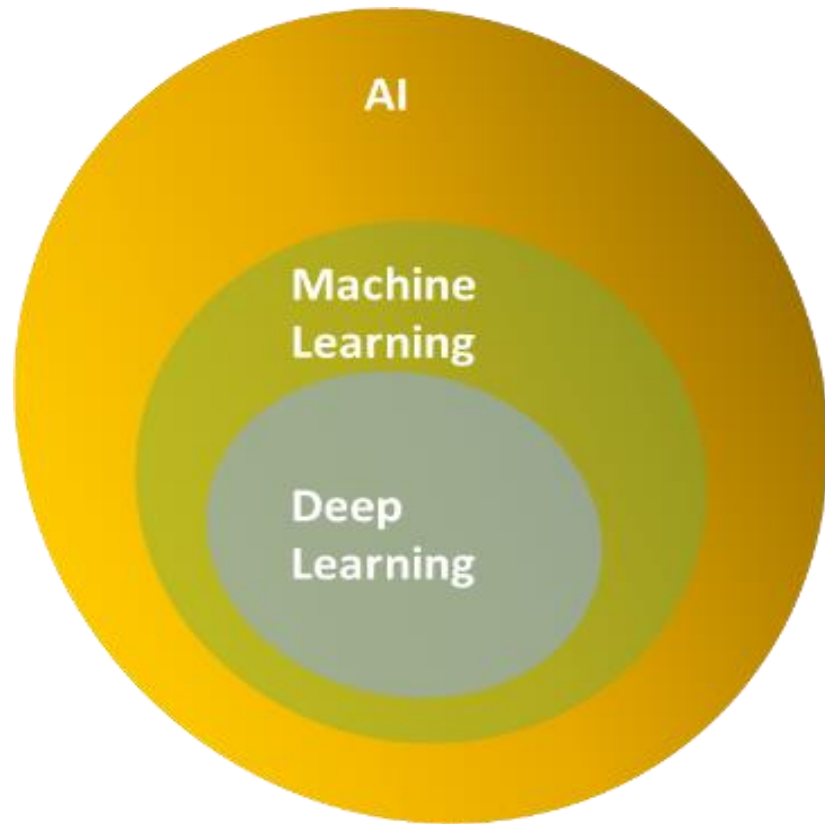
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How to Build an AI Project

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# AI & ML - What is it?

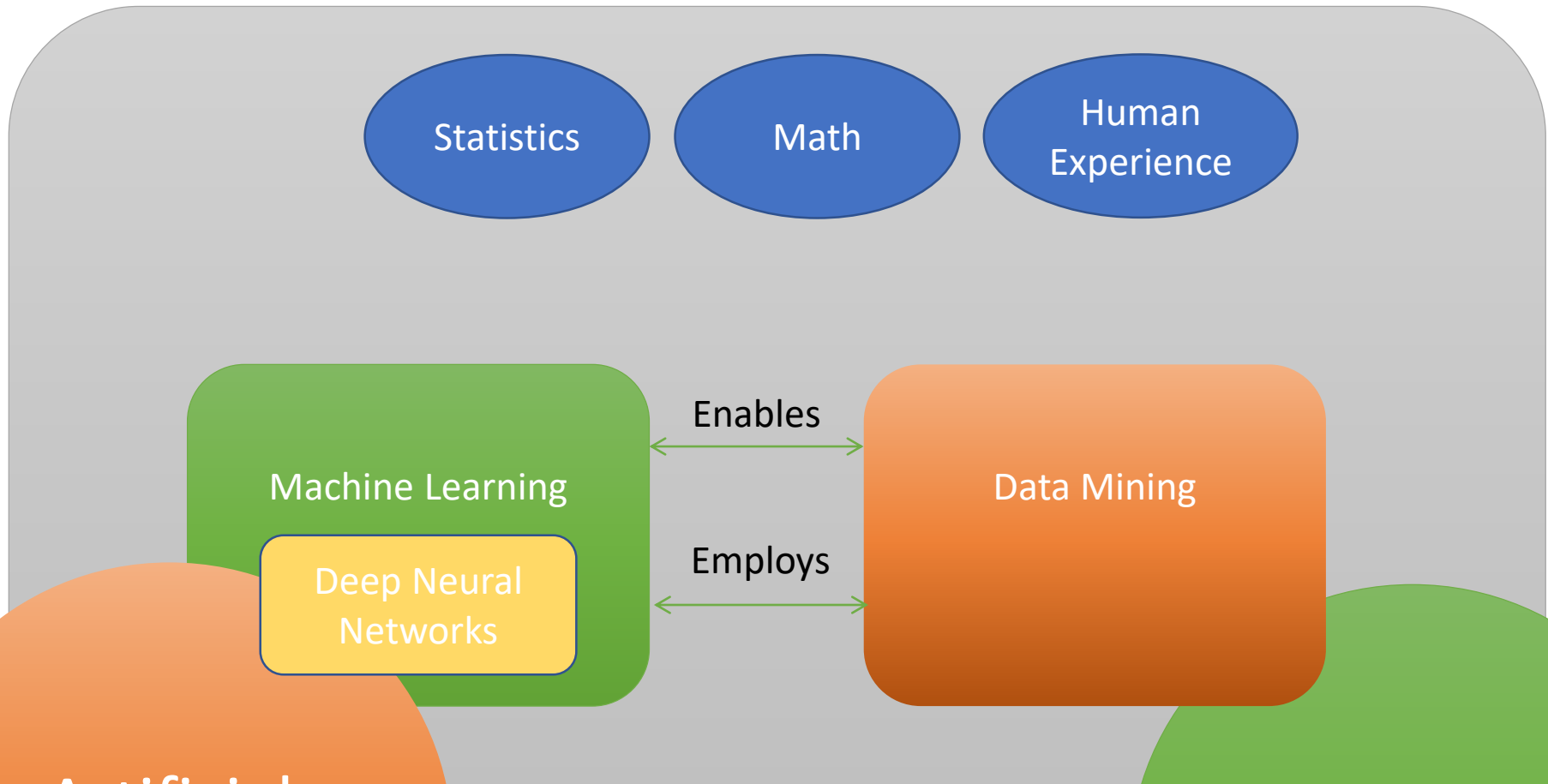


Arthur Samuel (1959) on  
**Machine Learning:**

*The field of study that gives computers the ability to learn without being explicitly programmed.*



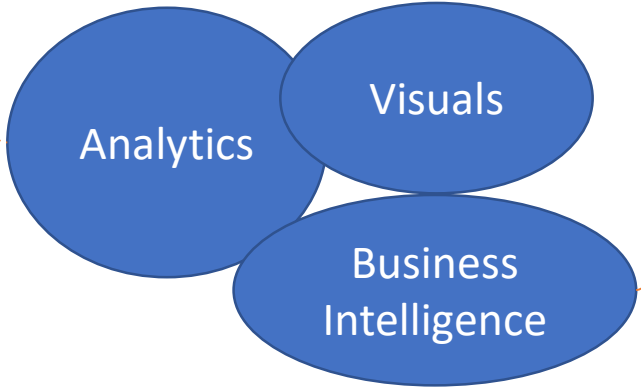
DATA



INSIGHTS

Artificial Intelligence

Big Data



# History of Artificial Intelligence

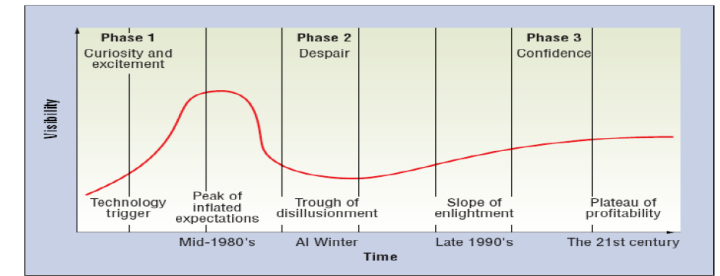


Figure 1. The Hype Cycle and AI winter [Menzi03]

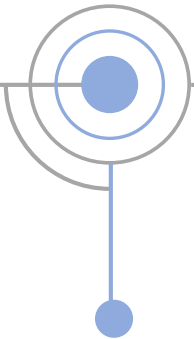
1961 FIRST COMPUTER FOR SEISMIC DATA PROCESSING

1973- COMPUTER CHESS

1980- EXPERT SYSTEMS

MACHINE LEARNING IN GEOSCIENCE

1950



1950: TURING CAN MACHINE THINK?

1954: RUSSIAN TRANSLATION

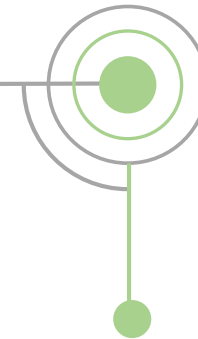
1956 FIRST USE OF WORD AI

1970



AI WINTER  
1980s-1990

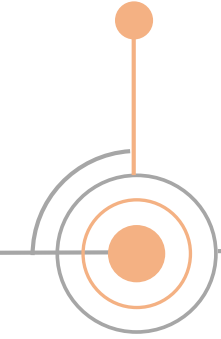
2000



1997 – IBM DEEP BLUE BEATS KASPAROV

2000 – ANN PREDICTION

2018



AUTONOMOUS CARS

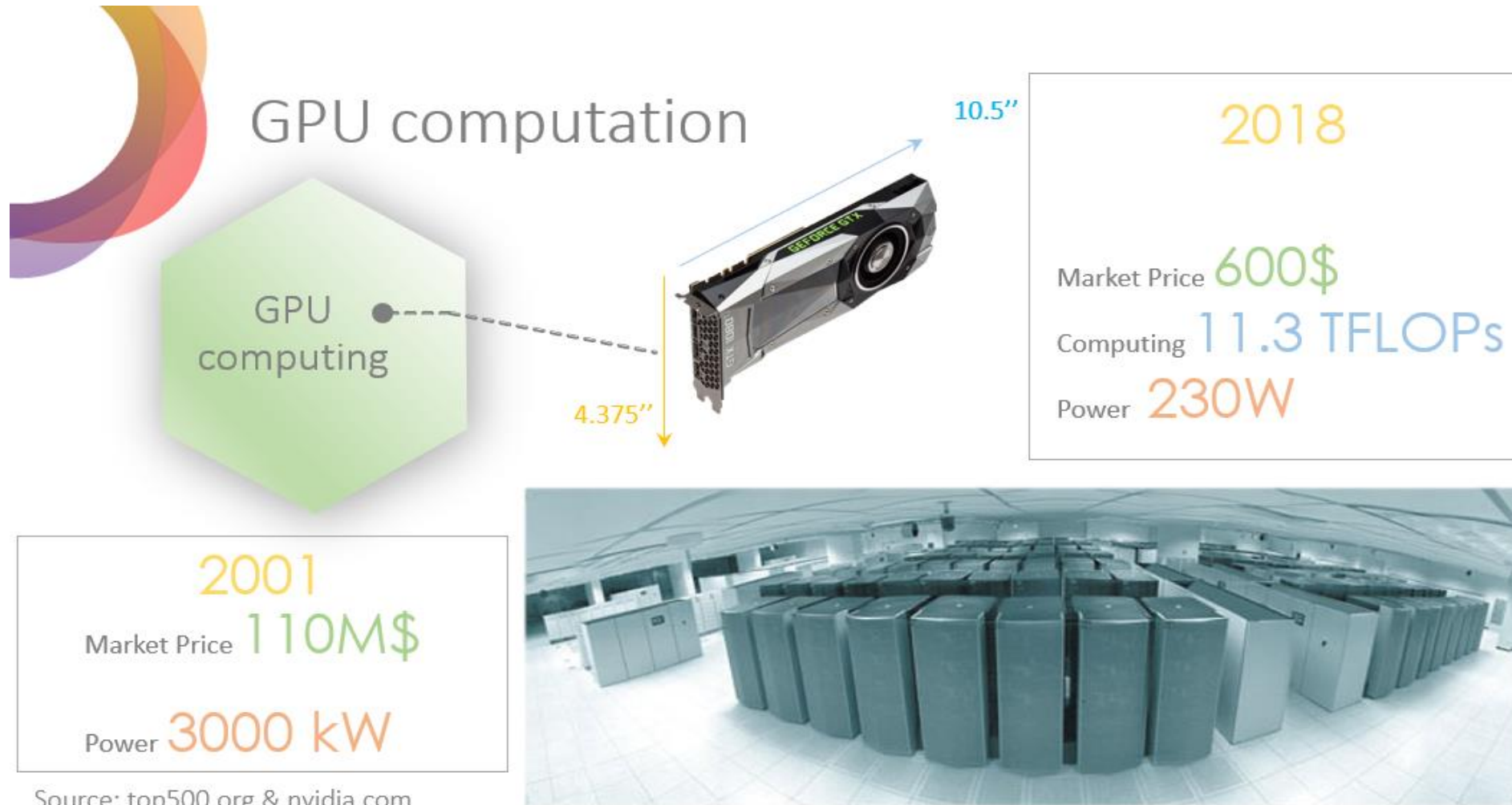
ALPHA GO

FACE RECOGNITION

EAGE PARIS 2017 ML



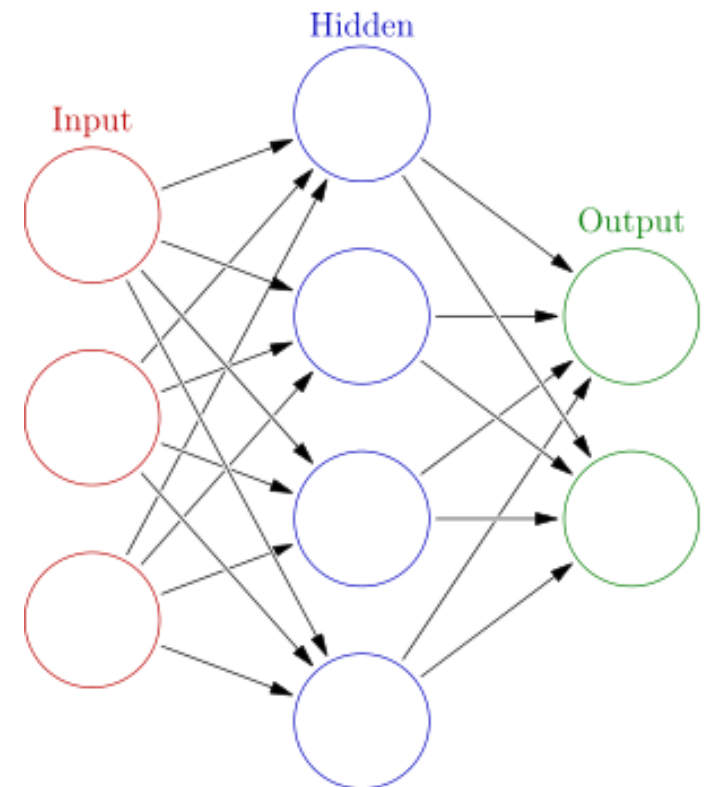
# AI explosion in 2018



# Nature as an inspiration

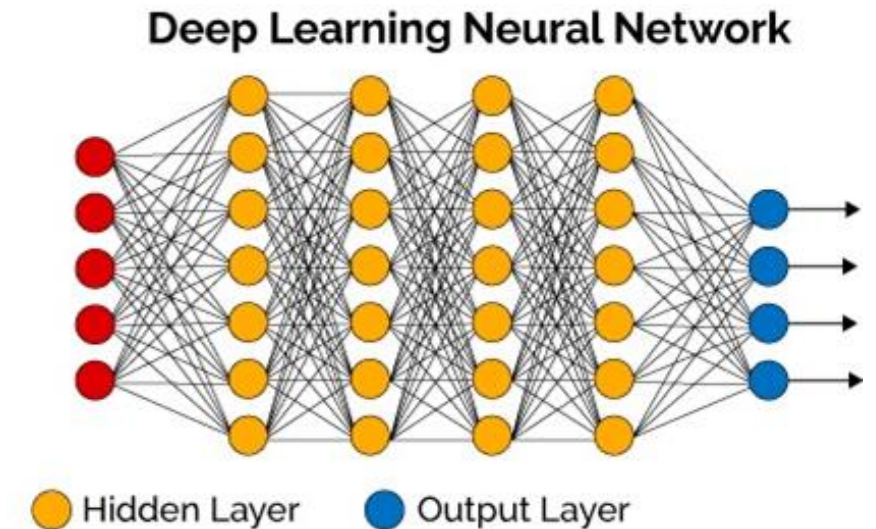
Artificial neural networks (ANN) mimic neurons in a brain

- Layers of nodes with weighted connections between layers
- Information through network changes its structure – **it learns**



# Architectural Principles of DNN

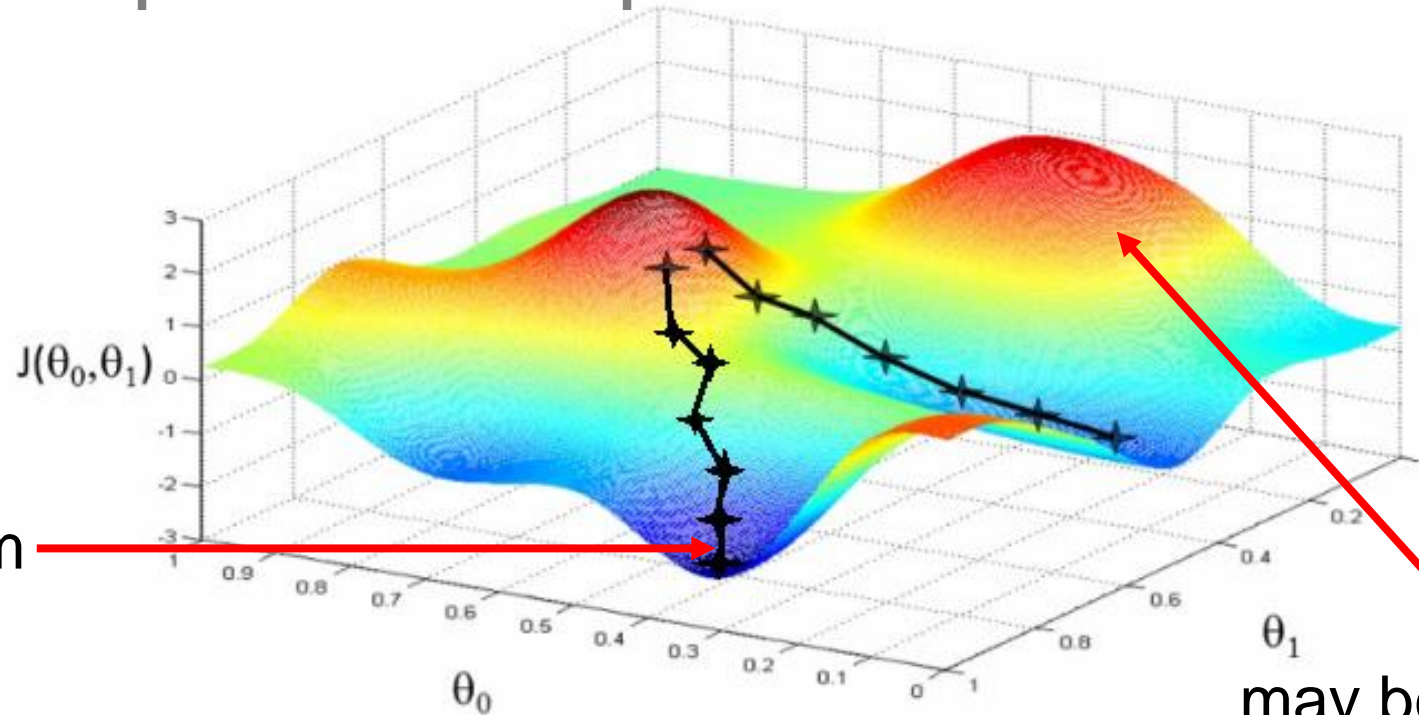
- **Features** : measurable property or characteristic of a phenomenon being observed
- **Layers**: information are passed from one layer to the other
- **Activation Function**: mathematical functions that act as filter
- **Loss Function**: define the difference between the predicted and actual features
- **Optimization Algorithms**: algorithms that minimize the loss functions or errors in the modeling
- **Hyperparameters**: adjustable parameters, such as number of layers, learning rate that can influence the effectivity a of neural network



Further Readings: Coursera  
Machine Learning, by Prof.  
Andrew Ng



Mathematically, training a neural network is an optimization problem



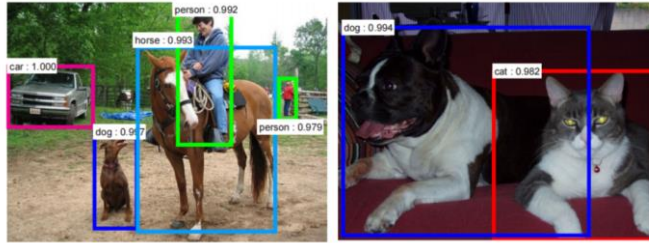
global minimum

may be N dimensional

- Show different input values and compute error
- Adjust weights in direction where error is minimized (along gradient)
- Eventually reach minimum value

# Types of Neural Network

## Convolutional Neural Nets



## Recurrent / Recursive Neural Nets



## Reinforced Neural Nets



IMAGE CLASSIFICATION  
and  
FEATURE DETECTION

TEMPORAL BEHAVIOUR MODELING  
AUDIO, TEXT  
NATURAL LANGUAGE PROCESSING

PAVLOVIAN CONDITIONING

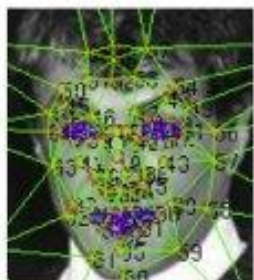
# Deep learning has found many applications in image processing



(a)



(b)



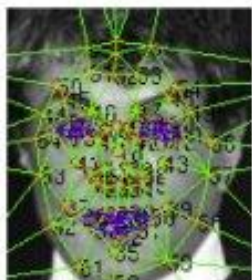
(c)



(d)



(e)



(f)



(g)



(h)

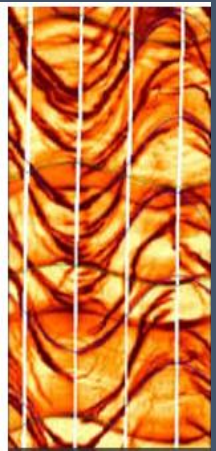
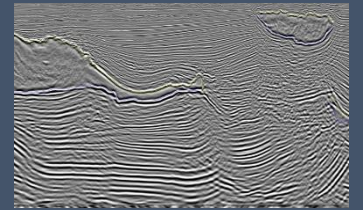
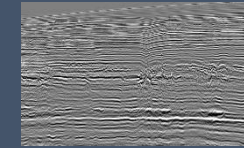
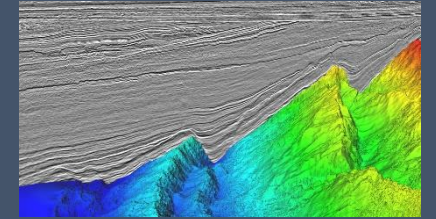
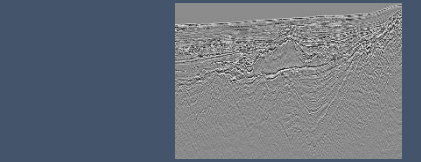
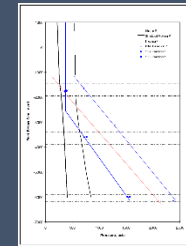
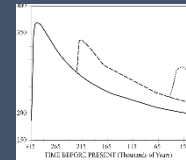
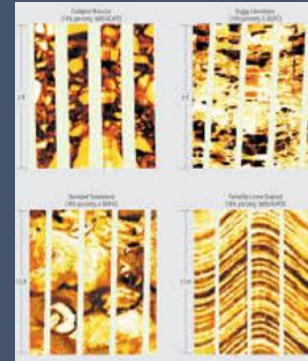
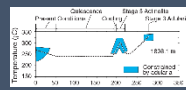
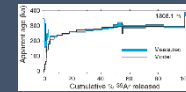
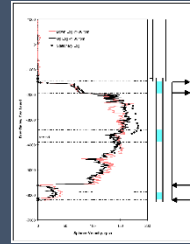
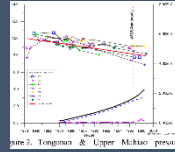
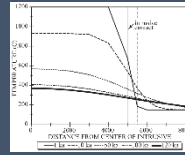
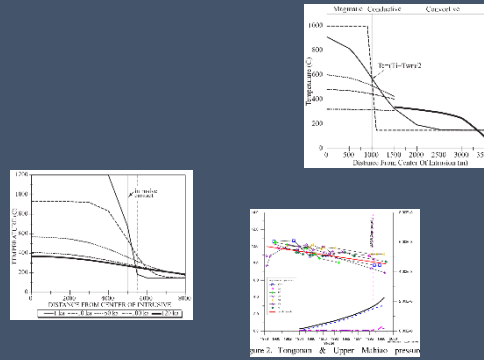
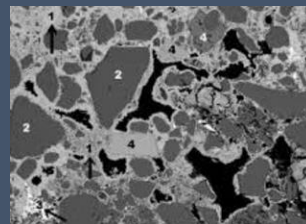
Facebook

- Facebook's DeepFace for facial verification
- DNN with 9 layers
- Trained using millions of images uploaded by users
- Accuracy reaching 97.35%



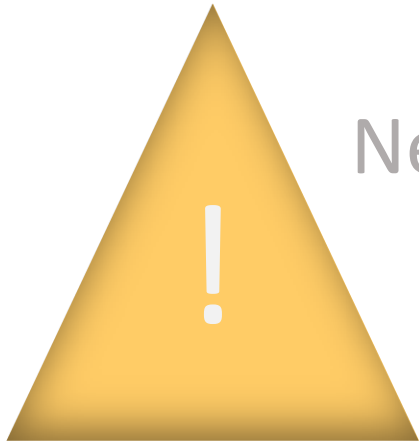
What do geoscientists do on a daily basis?  
We make (image) files

# EARTH PROBLEMS



# Making AI work for Earth Science

Use AI to (classify, predict, learn from)  
**archived, historical megadata**



New Data Acquisition  
is very **costly**

Learn Effectively



10 vs 1,000 wells

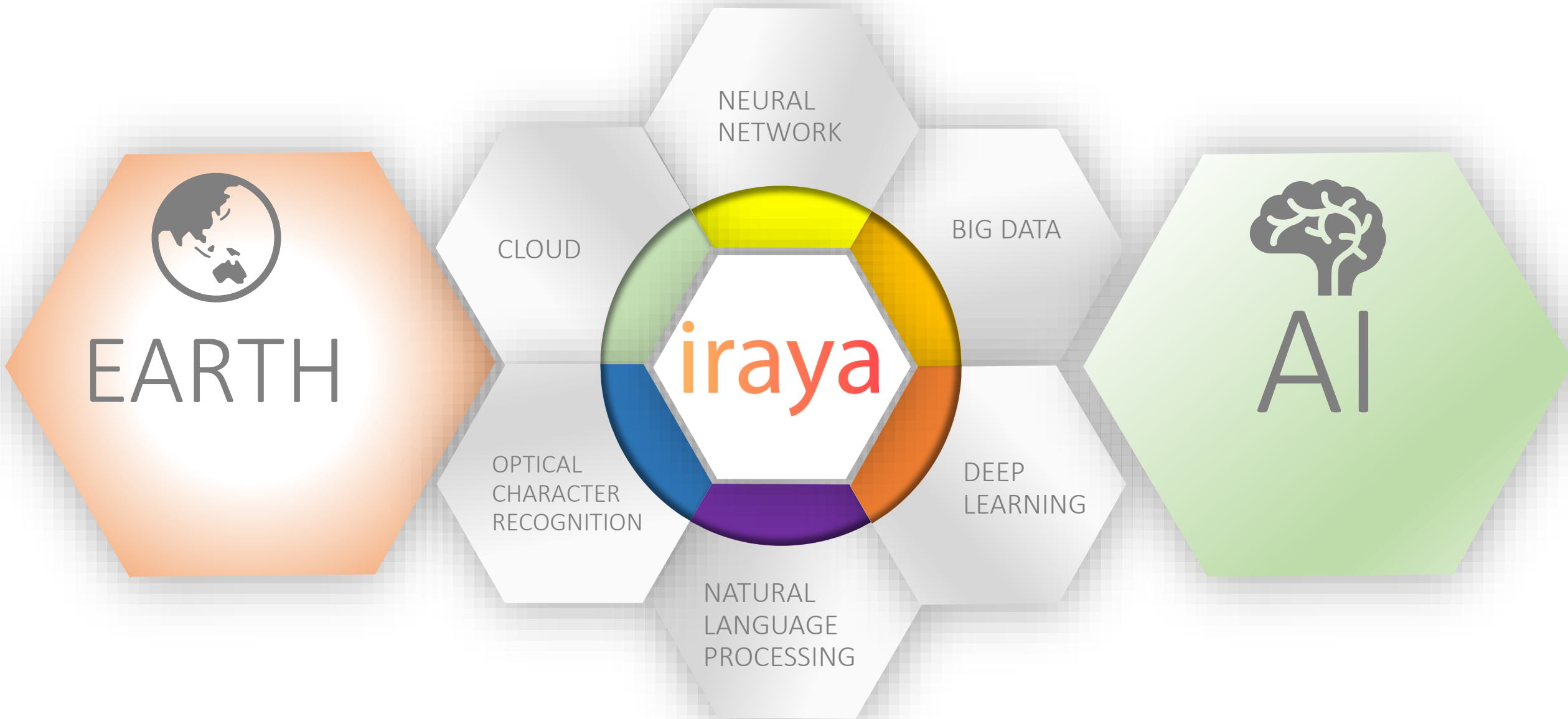
20 vs 2,000 seismic  
lines

Yet-to-Find becomes Easy-to-Find

Optimize efficiency during exploration and early  
development phase



# Making AI work for Earth Science



# Iraya Use Cases of AI

- Use Case # 1 : Data Mining
- Use Case # 2: Well Twinning
- Use Case # 3 : Clustering
- Use Case # 4: Deep Resolution

# Use Case #1: Data Mining

- **Problem Definition:**

Extract information from a unstructured dataset

- **Standard Solution:**

Download data, manually read metadata and load in a spreadsheet

- **Machine Learning Solution:**

Apply mining robots, elastic search, natural language processing, optical character recognition to reduce timeframe by a factor of 100.

# Data Mining Analogy



Harvesting



Transform

Sort



Enhance

# Data Harvesting Analogy



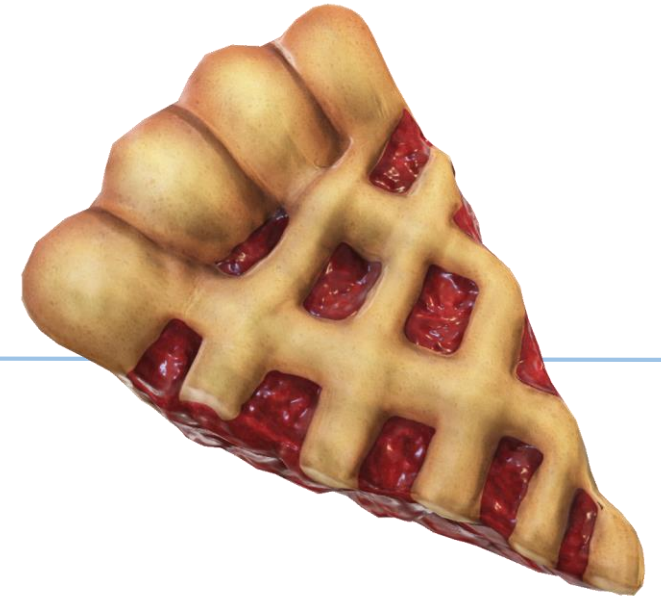
Harvesting



Fully automatic - AI driven



Transform  
Sort



Enhance



# MINE-BOT

# CLASSIFY

# VISUALIZE

# BIG DATA ANALYTICS

Digital or Physical Storage  
Web, USB, HDD,  
Vintage



All types of file type  
\*las, segy, pdf,xls..

Catalogue  
by file type or  
application

## Wells

- Metadata
- Curve class
- Frequency count

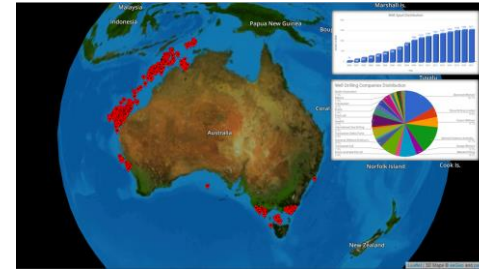
## Reports

## Seismic

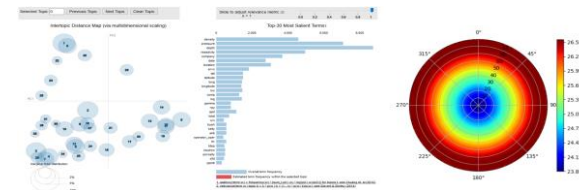
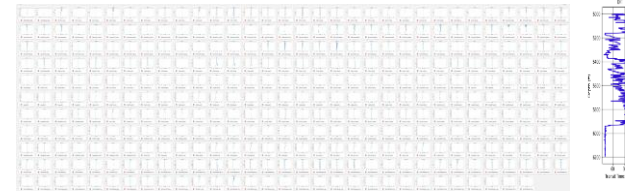
- Metadata extraction
- OCR Digitization
- Text vs Image

## Cloud

## WebGIS



## Big Data Visuals



## Dimension Reduction

T-SNE

PCA

Clustering

K-means

Classification

Supervised

Unsupervised

Convolutional  
Neural  
Network  
Applications

# Data Mining

LAS DATA IN DIFFERENT FORMAT

1,595 files

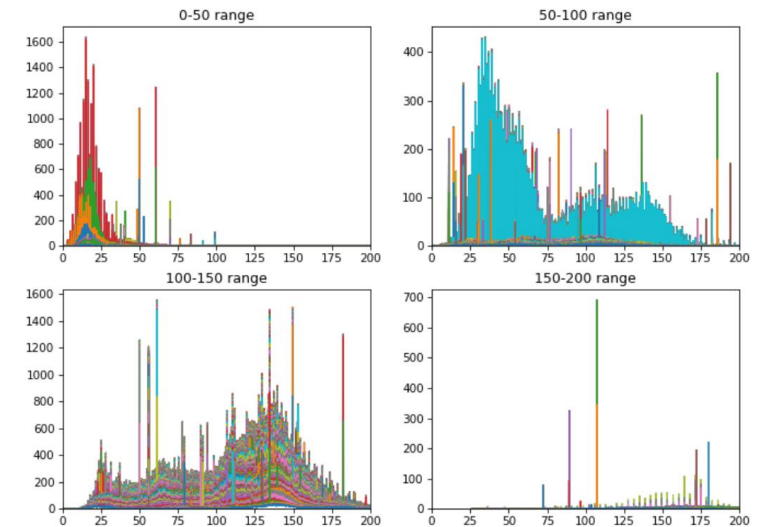
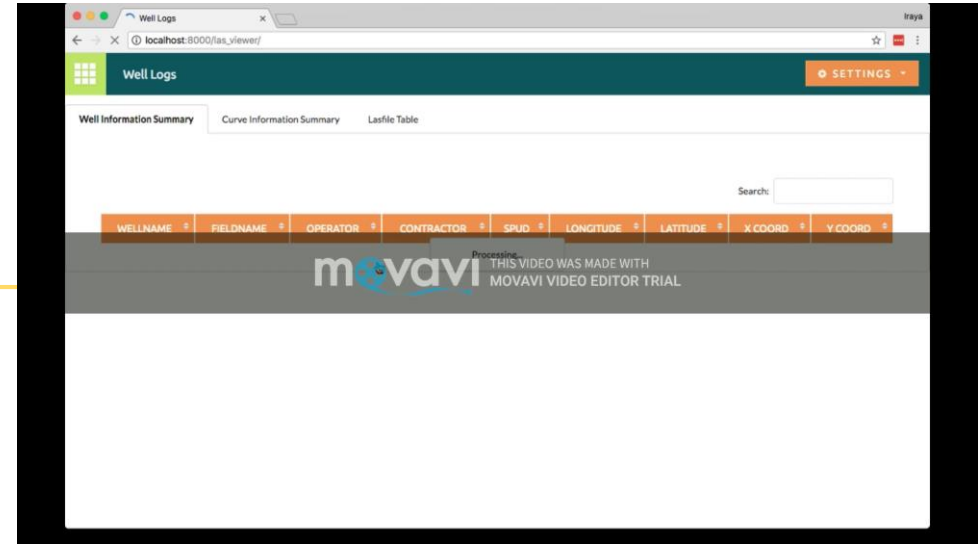
2 hrs 33.66 mins of Data Mining

Identified:

66,515 curves

5,681 most used (10% of data)

90% of DATA REMAINS TO BE TAPPED



# Use Case #2: Well Twinning

- **Problem Definition:**

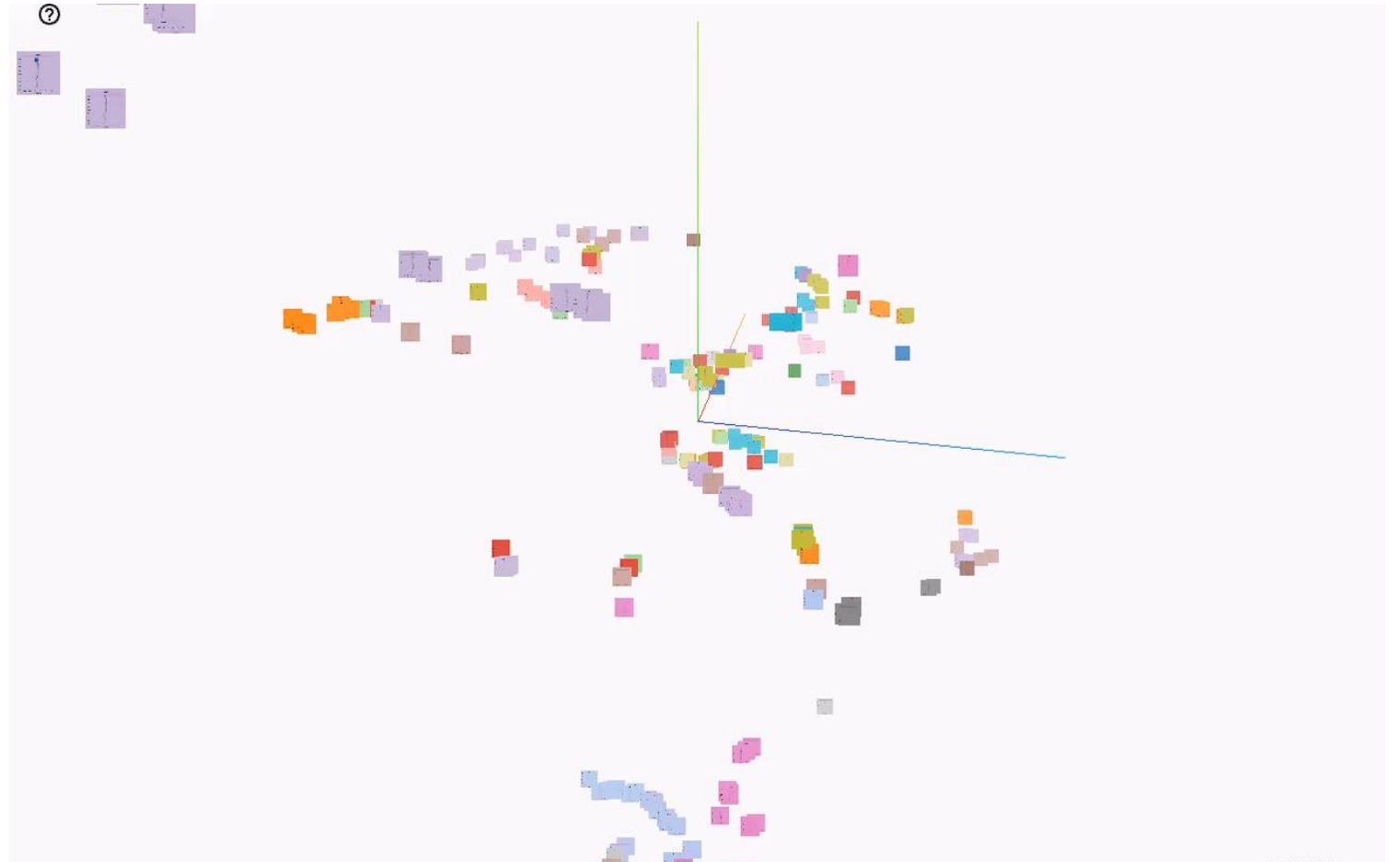
Find analog wells of a wildcat exploration area

- **Standard Solution:**

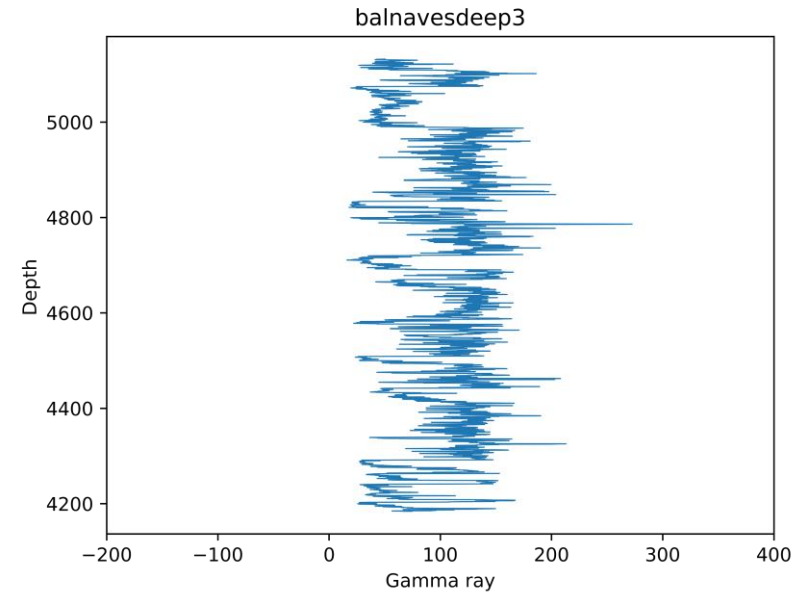
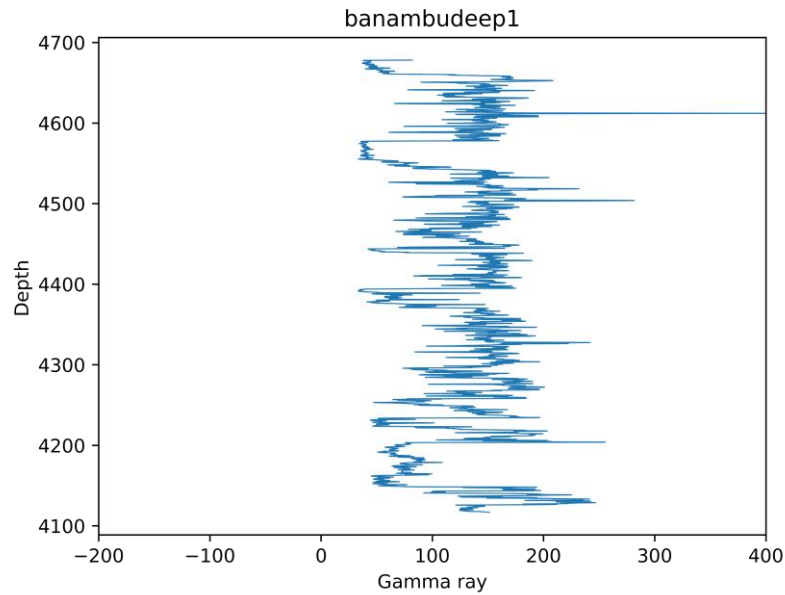
Find the nearest 1 or 2 wells in the nearest field (highly risky, does not capture all variabilities)

- **Machine Learning Solution:**

Leverage on big volume dataset to find geological analogs and de-risk potential prospect



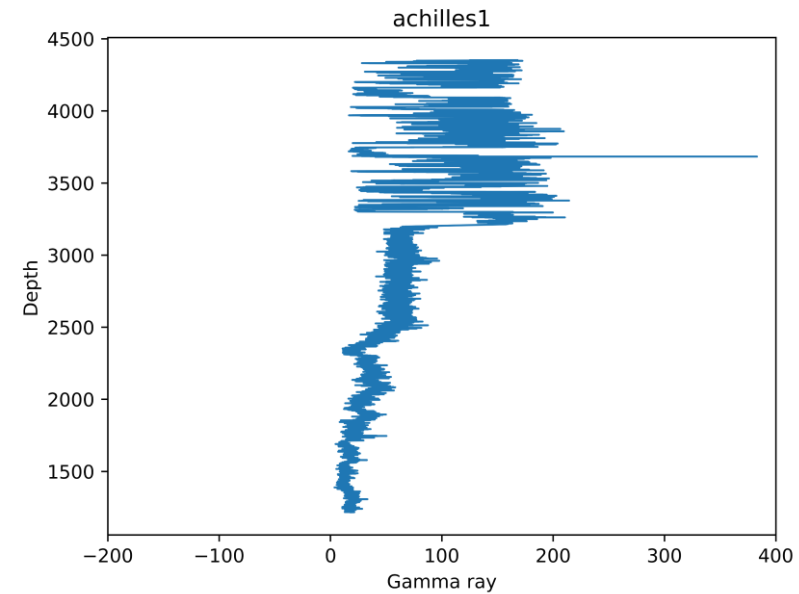
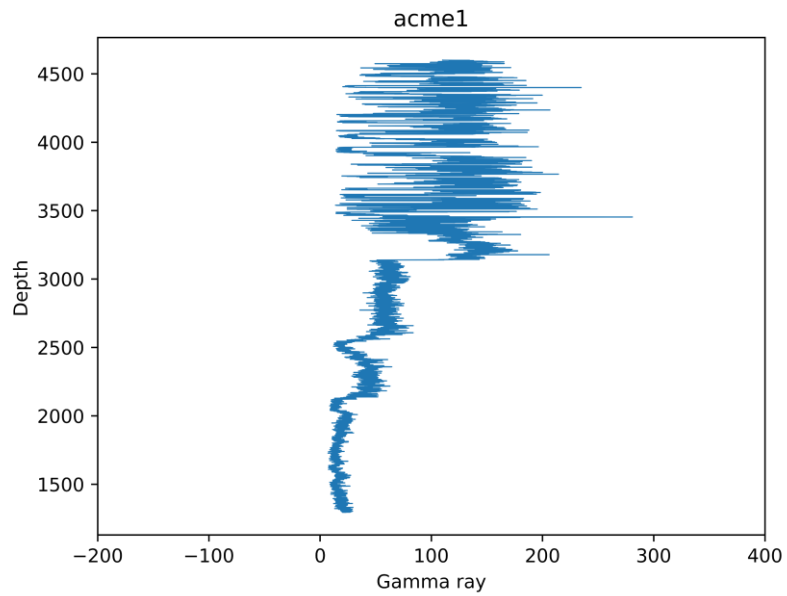
Automated Clustering of well data using t-SNE



Ⓢ

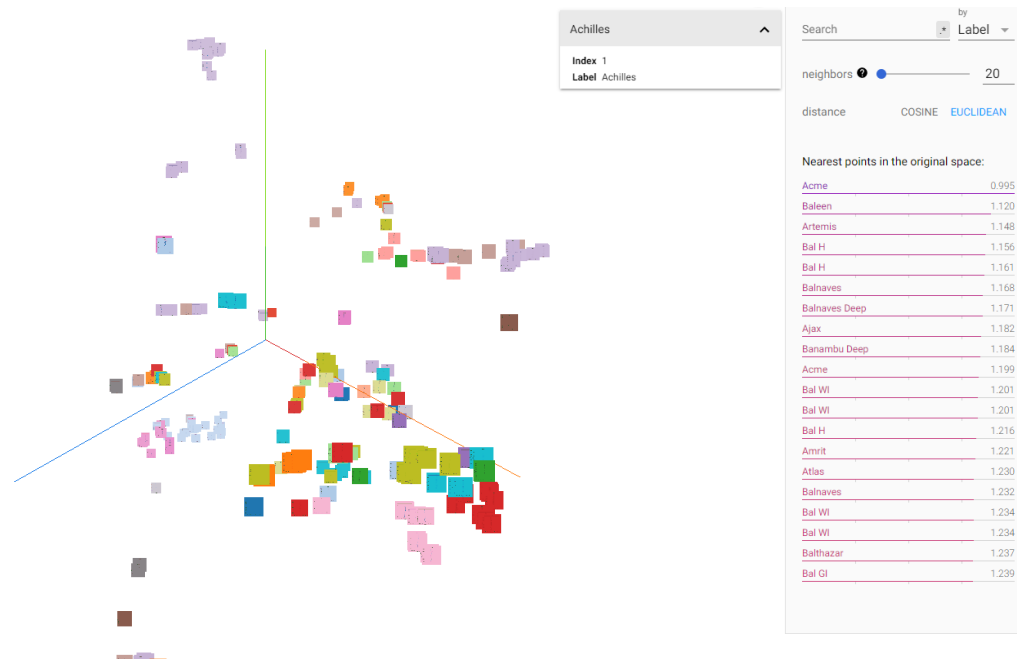


- Automated identification of the closest well “twin”, without prior geological knowledge
- Applicable in ultra-wildcat area or cross-country analog search

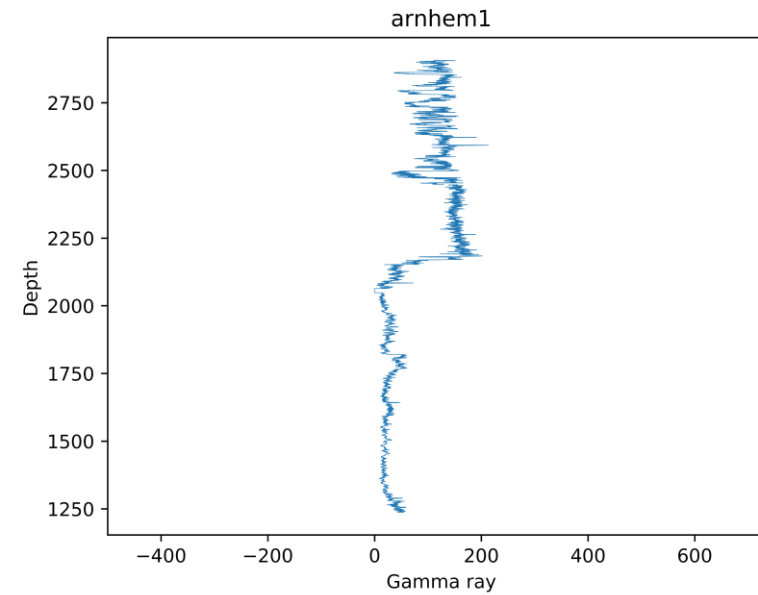
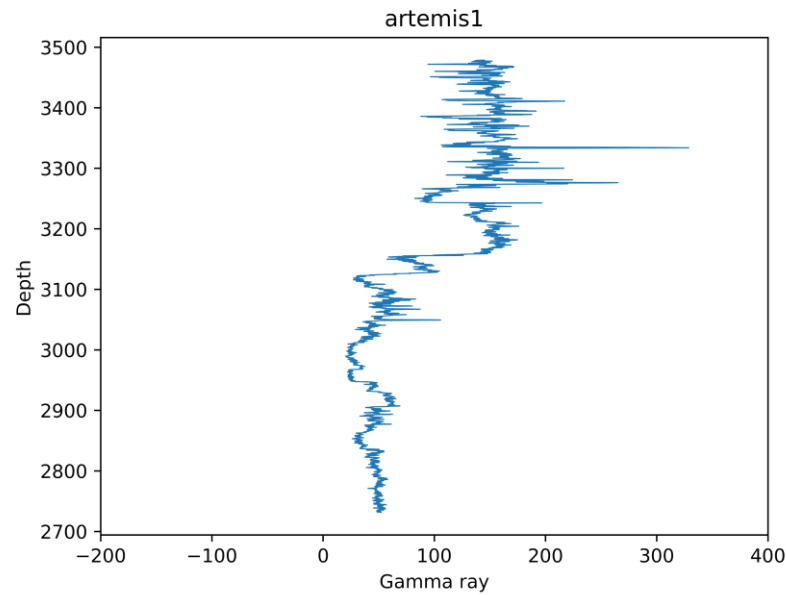


- Effective in automated identification of the closest genetic “twin” of the well
- Twin can provide valuable information on lithology, production history, drilling risks, etc.

②

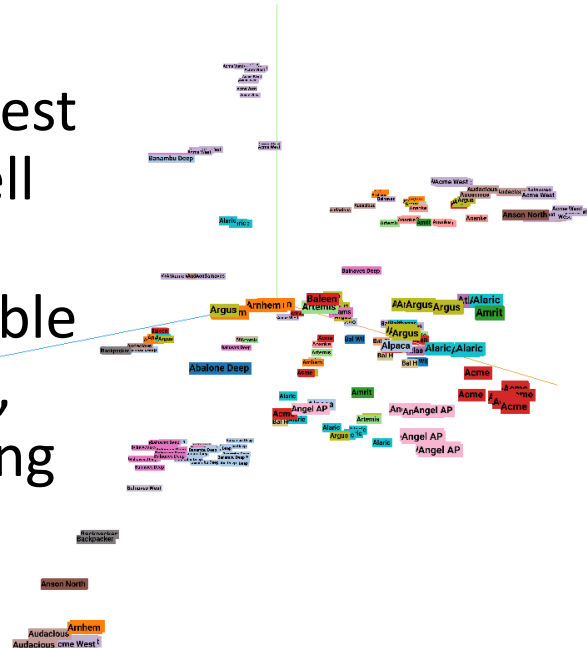






②

- Effective in automated identification of the closest genetic “twin” of the well
- Twin can provide valuable information on lithology, production history, drilling risks, etc.



Artemis

by Label

Index 15  
Label Artemis

Search

neighbors  20

distance COSINE EUCLIDEAN

Nearest points in the original space:

Arnhem	0.859
Acme	0.864
Arnhem	0.961
Arnhem	0.982
Artemis	1.069
Ananke	1.113
Arnhem	1.116
Ananke	1.117
Arnhem	1.119
Ananke	1.142
Audacious	1.143
Banambu Deep	1.149
Argus	1.163
Ajax	1.168
Arnhem	1.170
Audacious	1.171
Ajax	1.178
Artemis	1.195
Banambu Deep	1.206
Artemis	1.213

# Use Case #3: Clustering

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- **Problem Definition:**

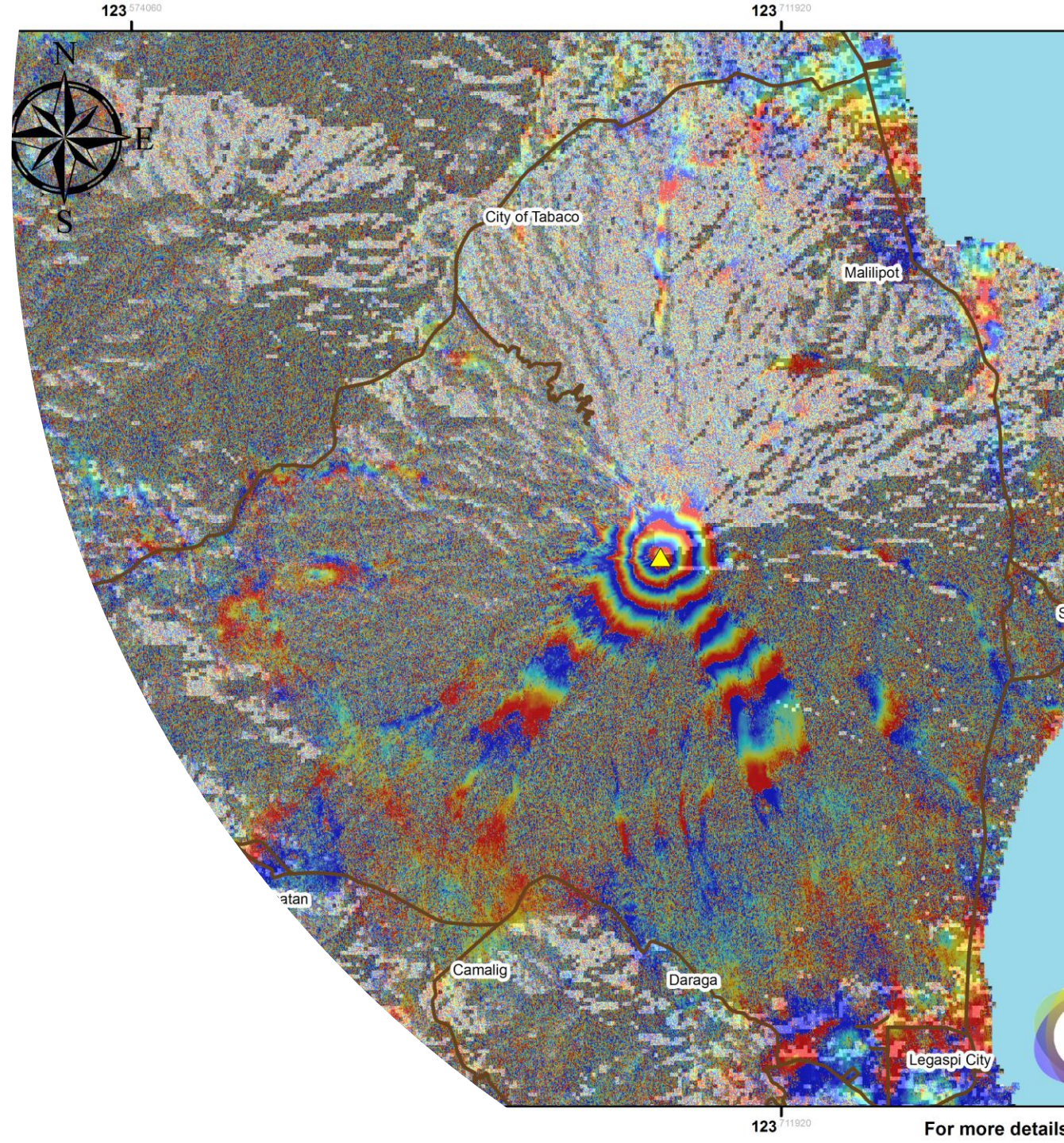
Identify surface features from satellite data

- **Standard Solution:**

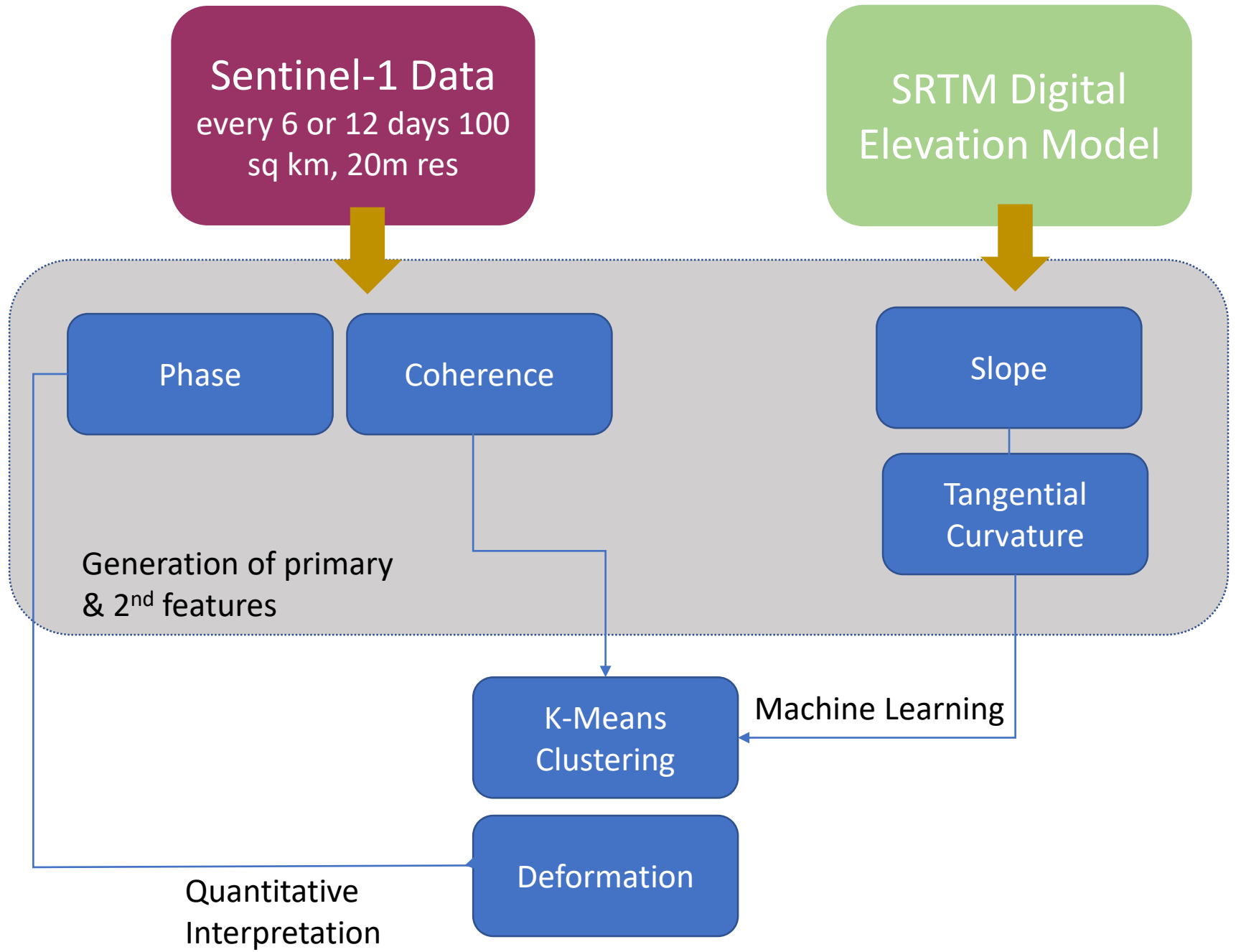
Manual Interpretation

- **Machine Learning Solution:**

Unsupervised classification of multiple extracted features

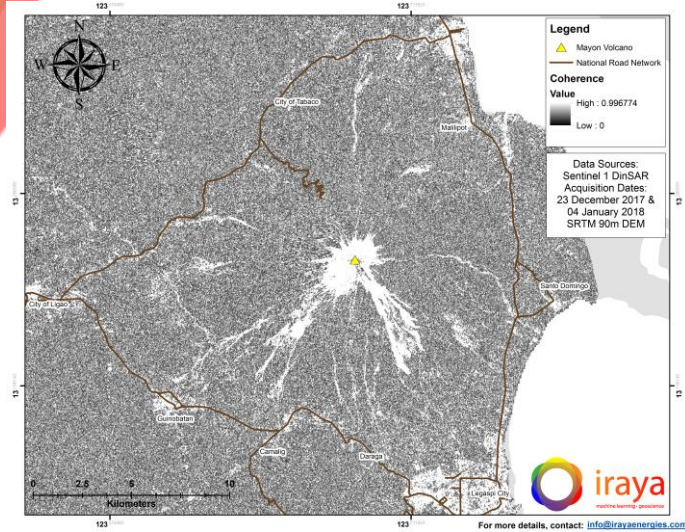




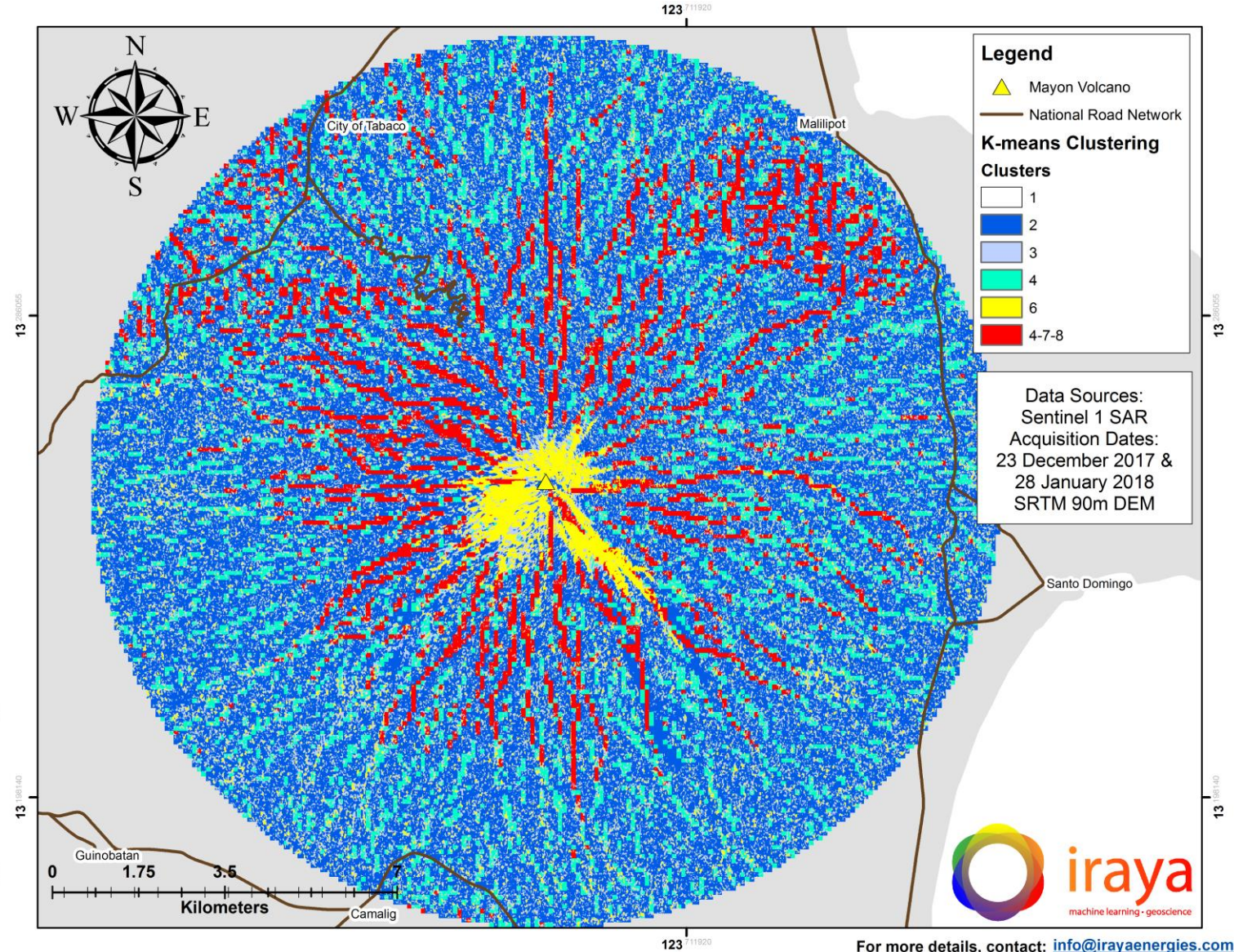




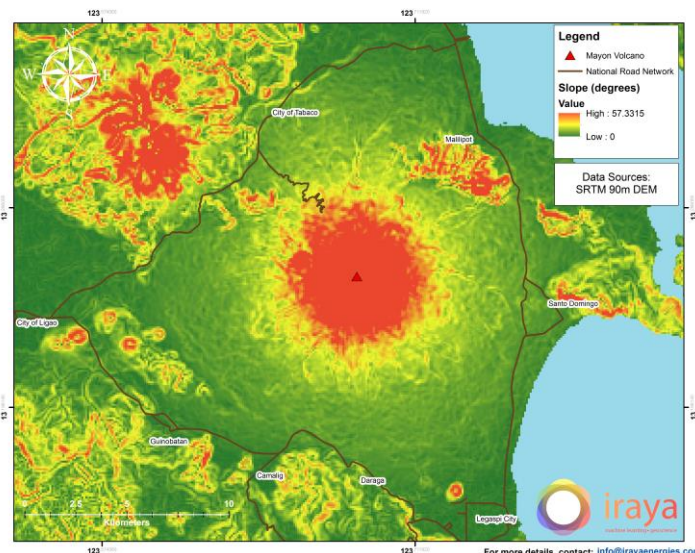
# Coherence



# K-means Cluster



# Slope





# Use Case #4: Resolution Enhancement

- **Problem Definition:**

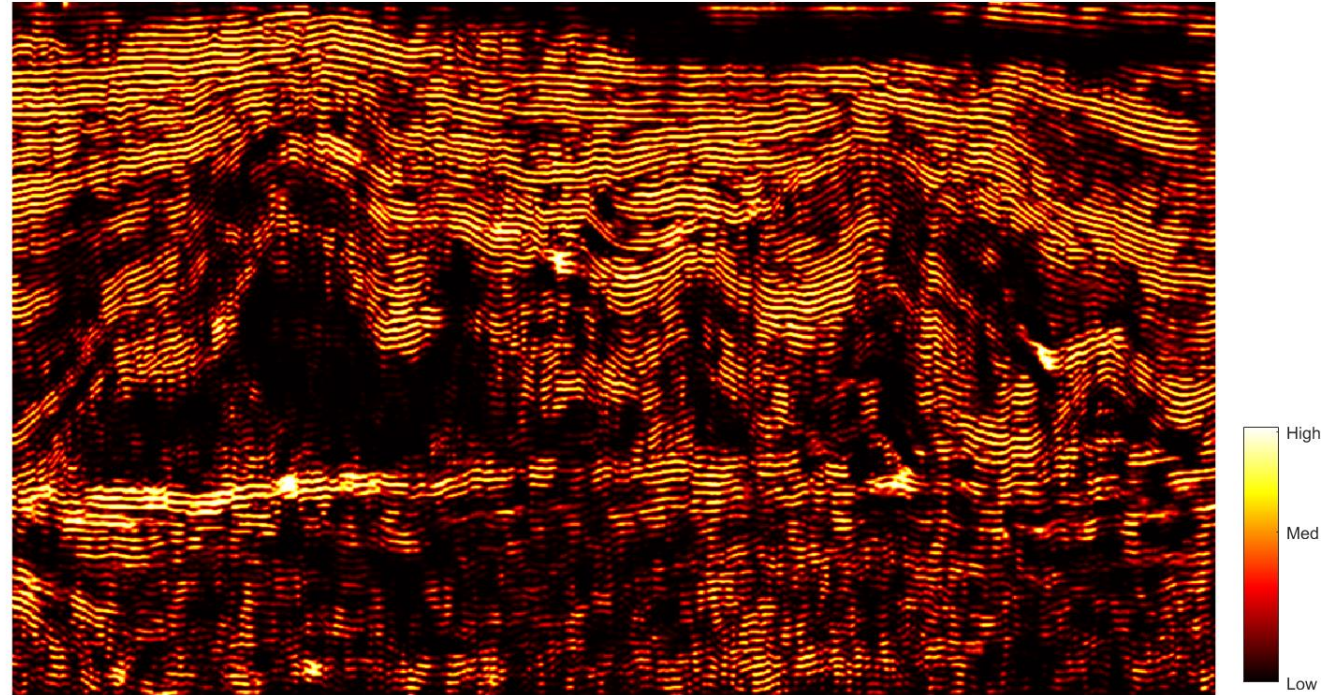
Increase seismic image Quality in Vintage Seismic acquisitions for better interpretation

- **Standard Solution:**

Traditional Seismic Processing + Stochastic Static Modeling

- **Machine Learning Solution:**

Model-based residual processing using deep convolutional neural network





# Iraya Machine Learned (ML) broadband enhancement – Methodology

## Architecture

Cascaded architecture using Adam optimizer, dropout 75% and transfer learning

## Training

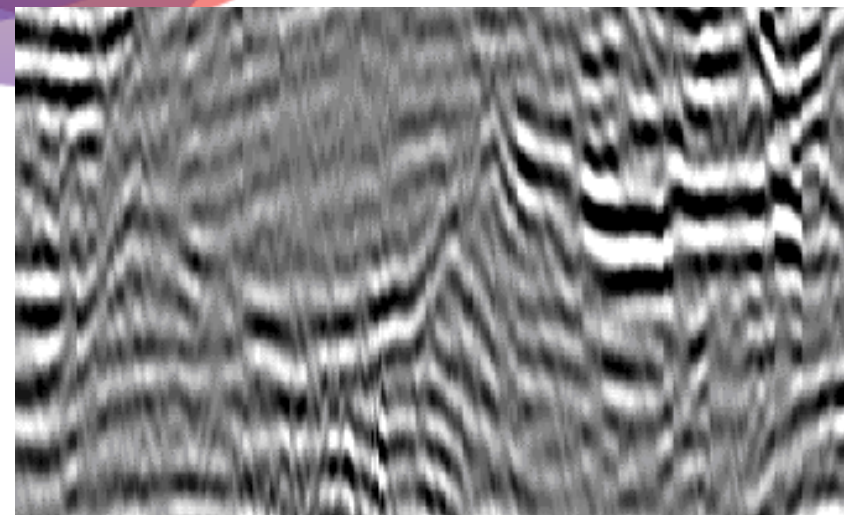
Epochs of over 120,000+, global 2D seismic examples, 1 month per training

## Inference

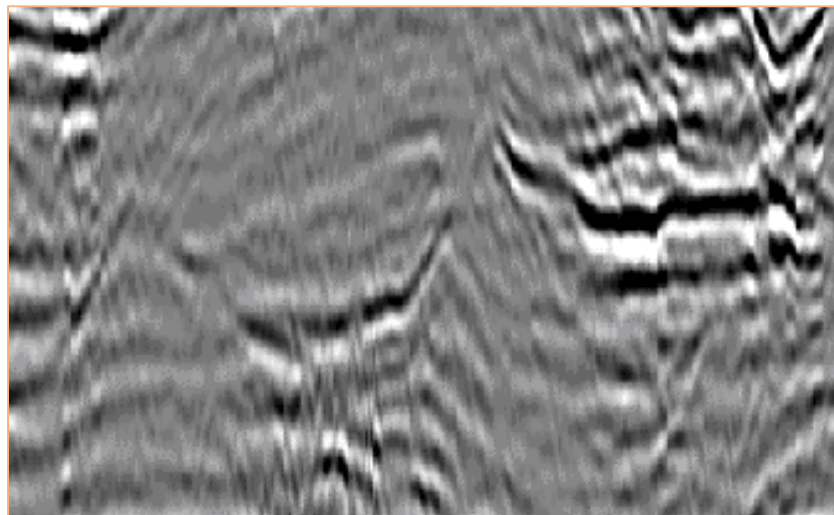
10minutes



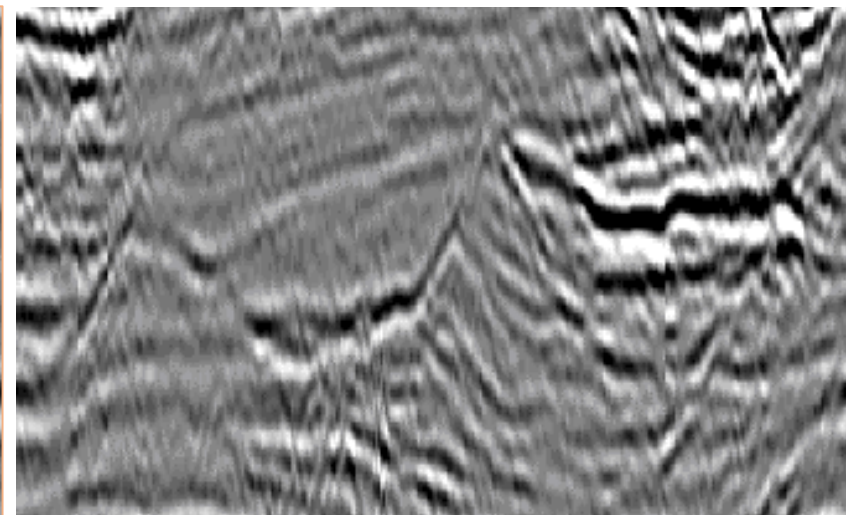
# Iraya Machine Learned (ML) broadband enhancement – Training



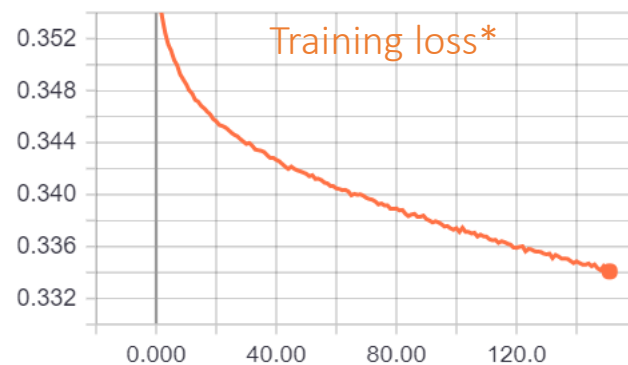
Input



ML enhanced

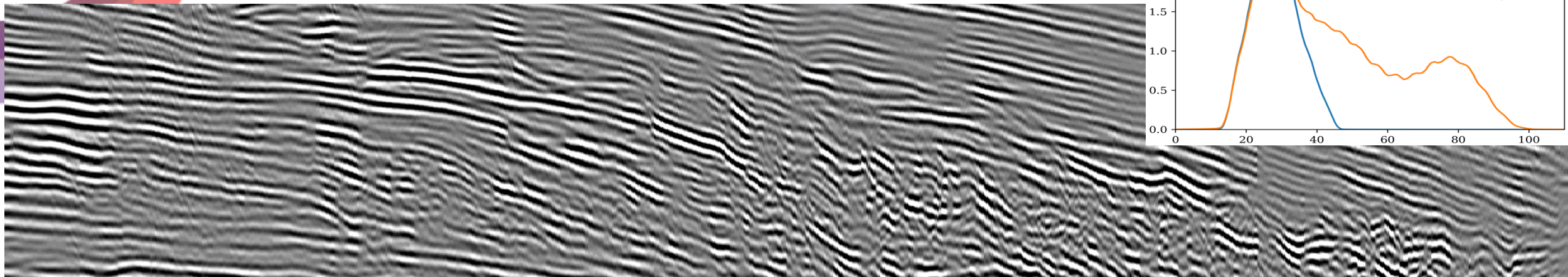
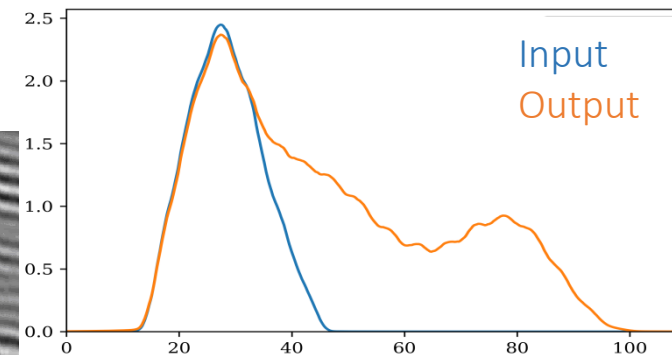


Original

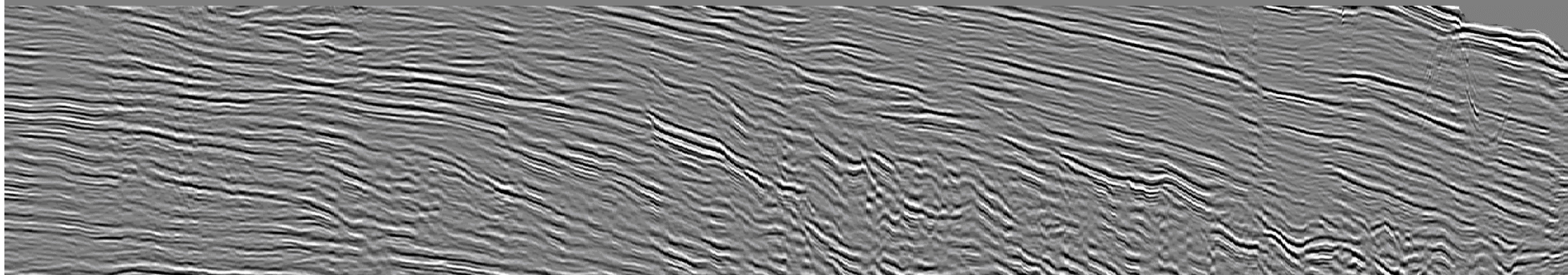


\*Loss computed over 120,000+ 2D Seismic

# Iraya Machine Learned (ML) broadband enhancement – Inference



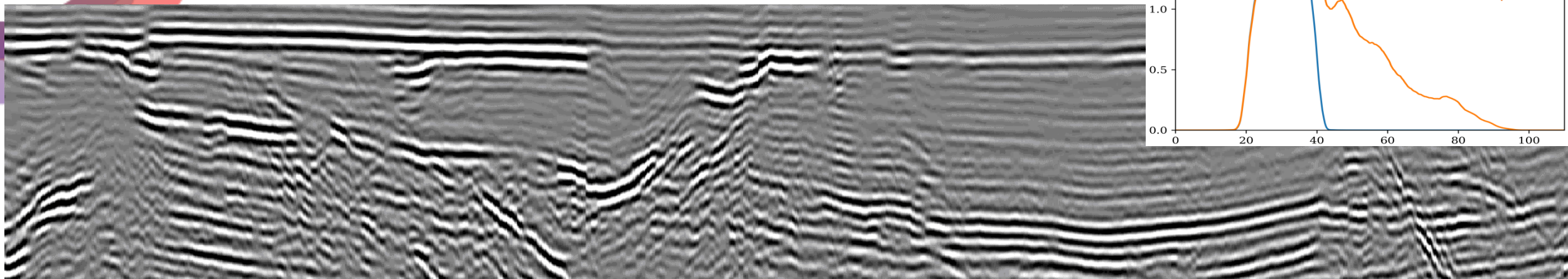
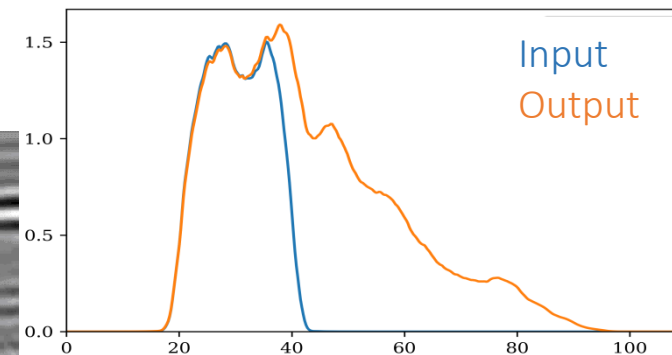
Original input



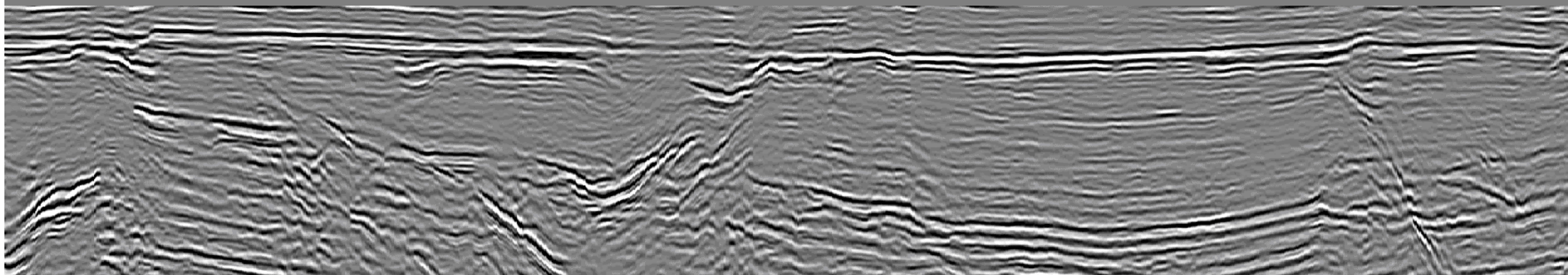
ML enhanced output



# Iraya Machine Learned (ML) broadband enhancement – Inference



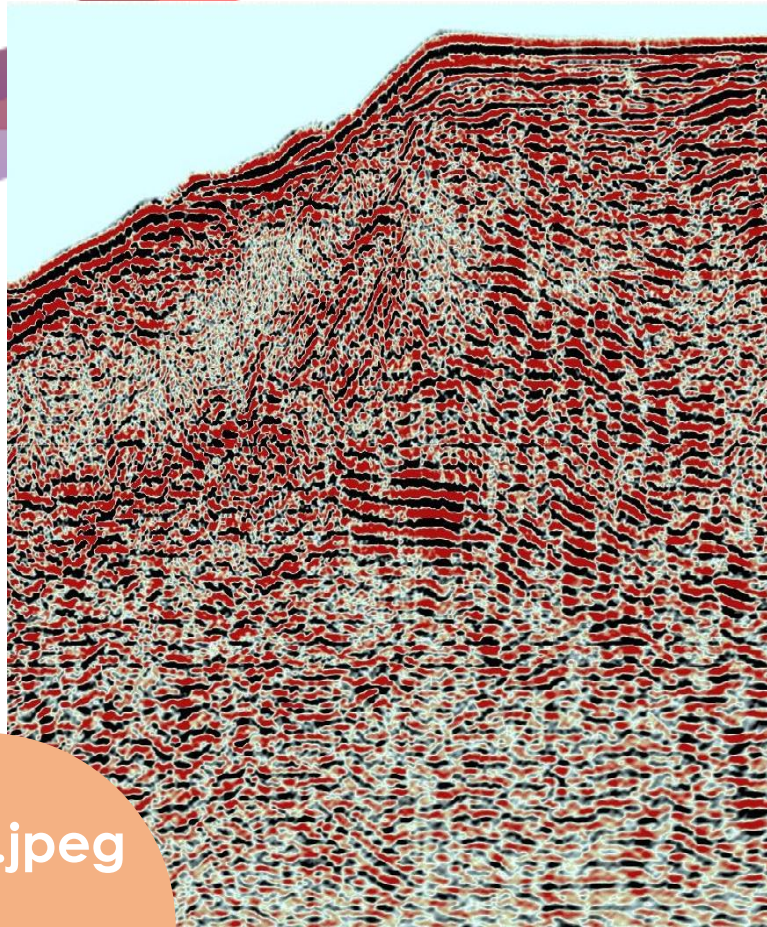
Original input



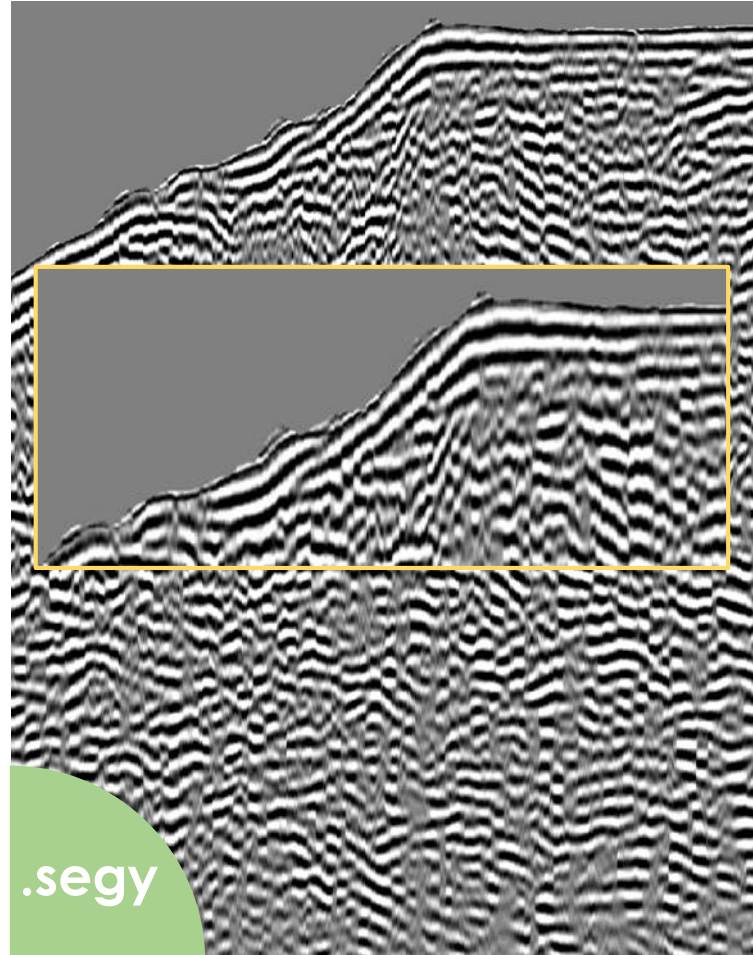
ML enhanced output



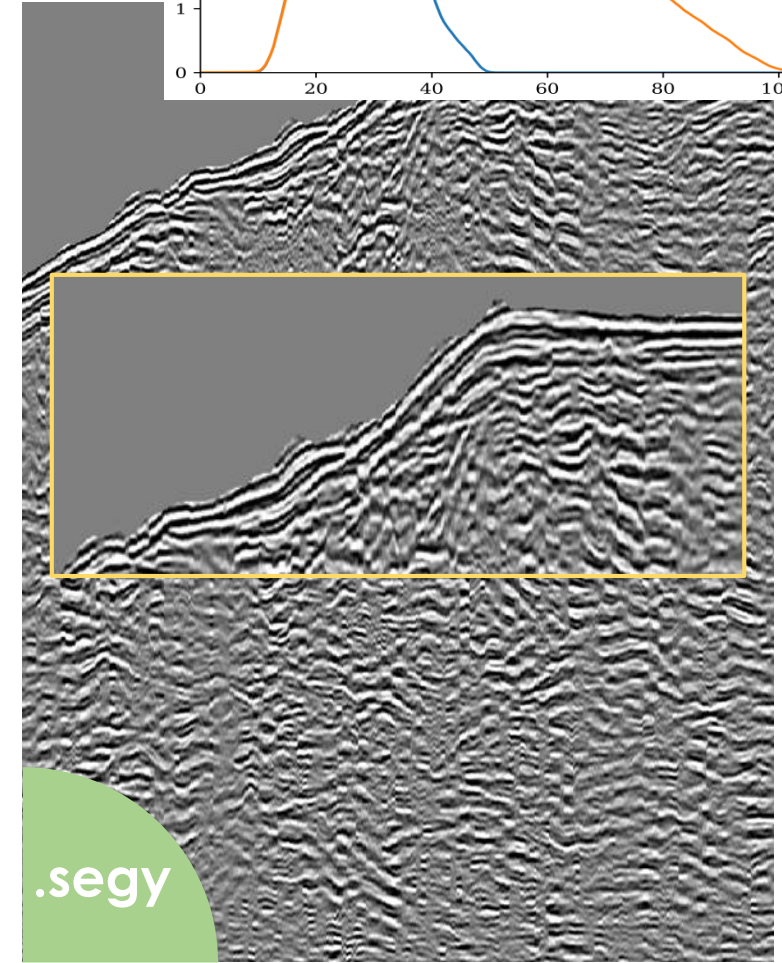
# Let's assume...



We have a powerpoint with somewhere a seismic image



We would like to find it, clean it and convert it into .segy



And why not enhance it at the same time

## Fully automatic - AI driven



Conclusion:

How to build an AI project

# Conclusion

## D2V Approach

- ✓ Define Problem
- ✓ Define Data
- ✓ Train
- ✓ Test
- ✓ Validate

Classification: Supervised vs Unsupervised

Prediction: Regression

Available : Check quality, quantity

Not Available: Find other sources? Public data, etc.





# Thank you!

To build AI experiments or participate in beta test, you may contact:

[info@irayaenergies.com](mailto:info@irayaenergies.com) or [nmh@irayaenergies.com](mailto:nmh@irayaenergies.com)