



GEE遥感训练营出品



# Google Earth Engine教学

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案例分析

Day 6

# 基本内容

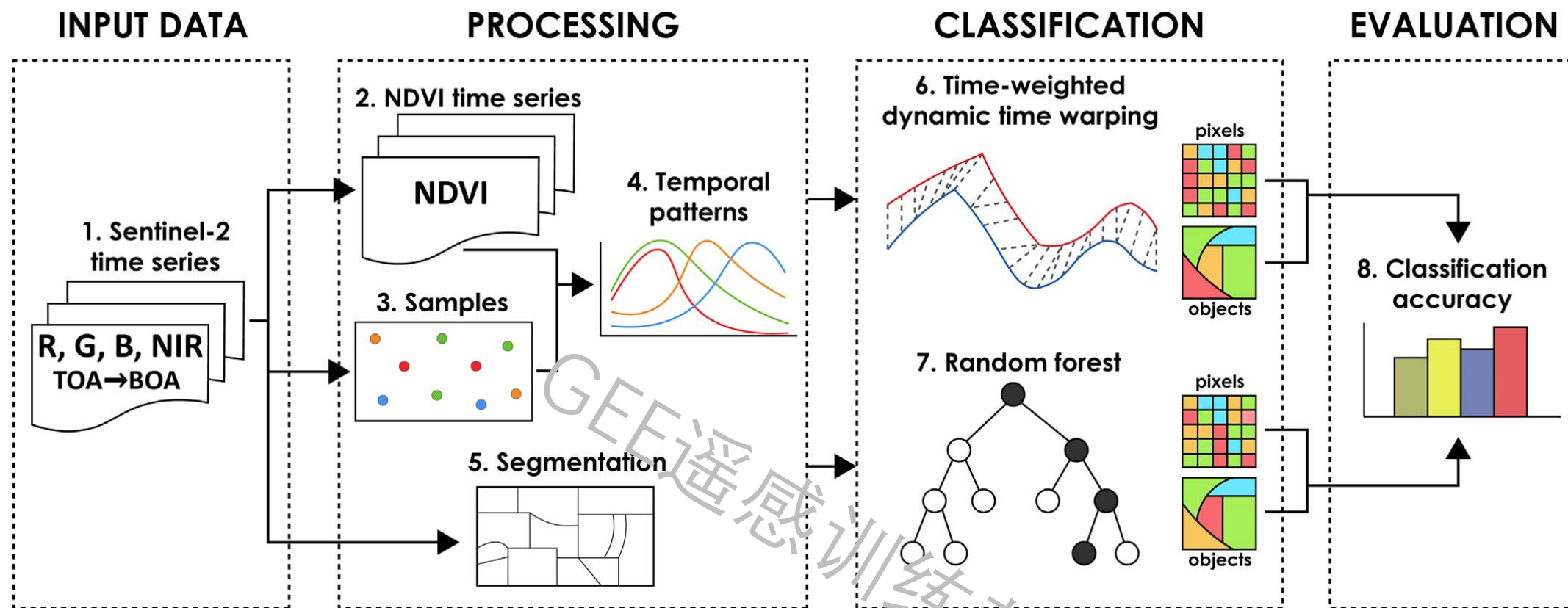
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✓ 案例分析

✓ PCA

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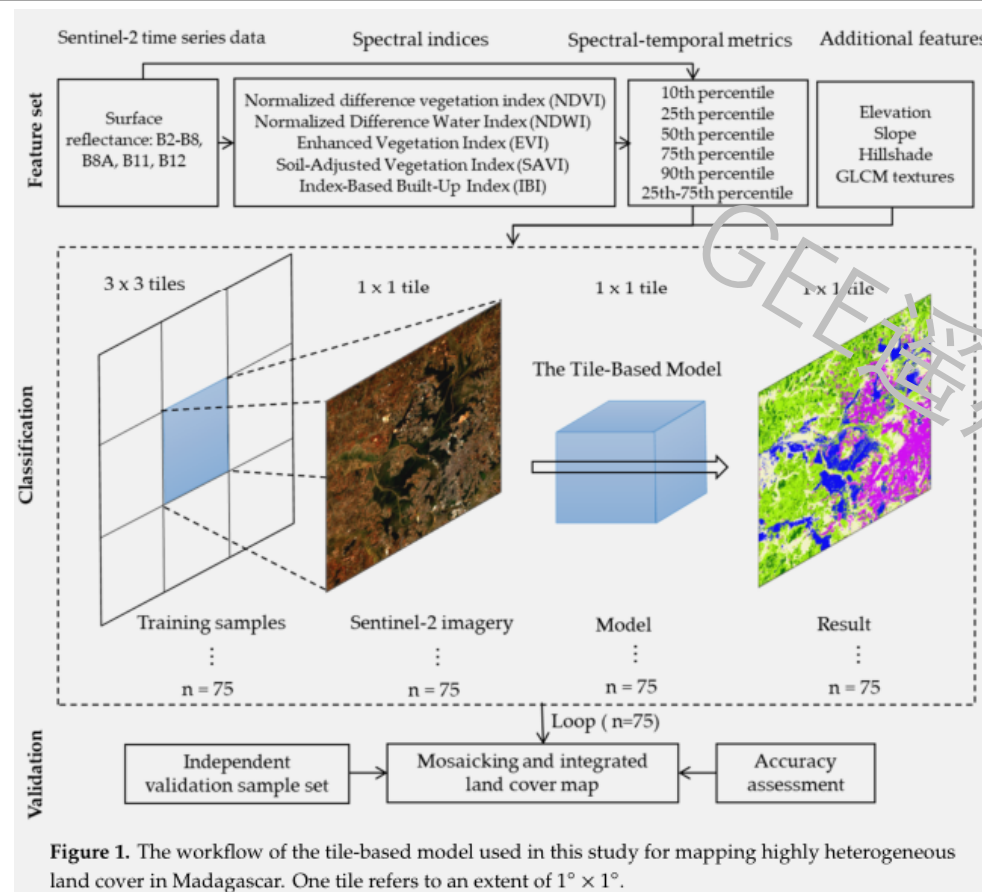
# 监督分类基本流程



# Improved method

## 题目：

Automatic High-Resolution Land Cover Production in Madagascar Using Sentinel-2 Time Series, Tile-Based Image Classification and Google Earth Engine



**Figure 1.** The workflow of the tile-based model used in this study for mapping highly heterogeneous land cover in Madagascar. One tile refers to an extent of  $1^\circ \times 1^\circ$ .

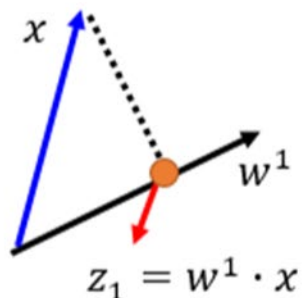
# PCA 原理

PCA

$$z = Wx$$

Reduce to 1-D:

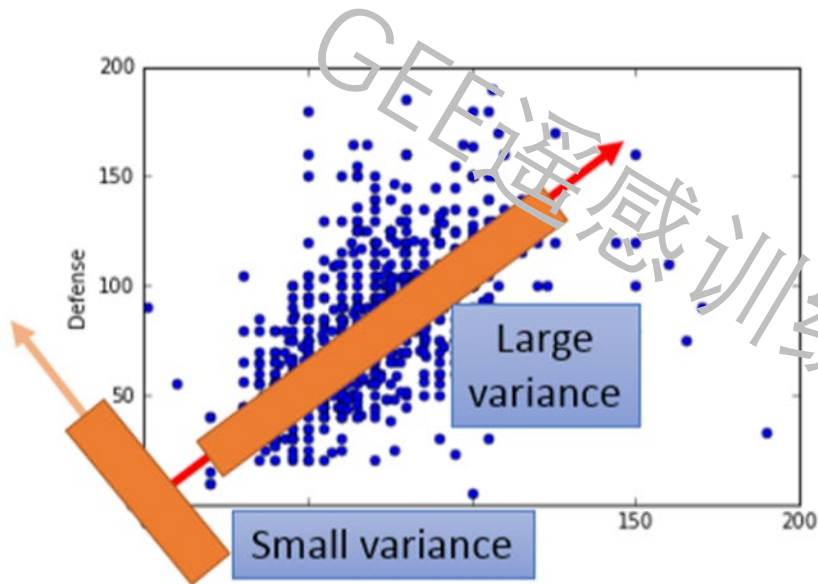
$$z_1 = w^1 \cdot x$$



Project all the data points  $x$  onto  $w^1$ , and obtain a set of  $z_1$

We want the variance of  $z_1$  as large as possible

$$\text{Var}(z_1) = \sum_{z_1} (z_1 - \bar{z}_1)^2 \quad \|w^1\|_2 = 1$$



$$\begin{cases} \max J \\ \text{s.t. } w_1^T w_1 = 1 \end{cases}$$

$$\begin{aligned} J &= \frac{1}{N} \sum_{i=1}^N [(x_i - \bar{X})^T w_1]^2 \\ &= \frac{1}{N} \sum_{i=1}^N w_1^T (x_i - \bar{X})(x_i - \bar{X})^T w_1 \\ &= w_1^T \left( \frac{1}{N} \sum_{i=1}^N (x_i - \bar{X})(x_i - \bar{X})^T \right) w_1 \\ &= w_1^T S w_1 \end{aligned}$$

$$L(w_1, \lambda) = w_1^T S w_1 + \lambda(1 - w_1^T w_1)$$

$$\begin{aligned} \frac{\partial L}{\partial w_1} &= 2S w_1 - 2\lambda w_1 = 0 \\ \Rightarrow S w_1 &= \lambda w_1 \end{aligned}$$

$$\text{Var}(a) = \frac{1}{m} \sum_{i=1}^m (a_i - \mu)^2 \quad \text{Var}(a) = \frac{1}{m} \sum_{i=1}^m a_i^2$$



# Thanks for your attention

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