

# Chapter 1 Introduction

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## What is Pattern?

A pattern is the opposite of a chaos ; it is an entity vaguely defined, that could be given a name.

### Various kinds of patterns

- Visual patterns ( 视觉模式 ) such as eyes, nose, mouth, face, fingerprint, etc.
- Temporal patterns ( 时序模式 ) such as speech, audios, videos, data streams, etc.
- Logical patterns ( 逻辑模式 ) such as characters, strings, images, etc.

## What is Recognition?

Identification of a pattern as a member of a category we already know, or we are familiar with.

识别是将模式鉴定为我们已知或者熟悉的类别的成员

### Two Types

Classification: Categories are **known** and the task is to assign a proper class label for each pattern

Clustering: Categories are **unknown** and the task is to learn categories and group the patterns accordingly

## What is Pattern Recognition?

Pattern recognition is the procedure of **processing and analyzing diverse information** (numerical, literal, logical) characterizing the objects or phenomenon, so as to **provide descriptions, identifications, classifications and interpretations** for them.

对表征事物或现象的各种形式的（数值的，文字的和逻辑关系的）信息进行处理和分析，从而对事物或现象进行描述、辨认、分类和解释的过程。

### Three procedures

- Perceive: : Observe the environment (e.g. interact with the real world)
- Process: Learn to distinguish patterns of interest from their background.
- Prediction: make sound and reasonable decisions about the categories of the patterns

### Applications

- Character Recognition
- Speech Recognition
- Fingerprint Recognition
- Signature Recognition
- Face Detection
- Text categorization

# Basic Concepts

## Model

Descriptions which are typically mathematical in form  
e.g. image to matrix; sound waves to frequency vector

## Sample

Representatives of the patterns we want to classify

## Training Set

A set of samples used to train classifiers

## Test Set

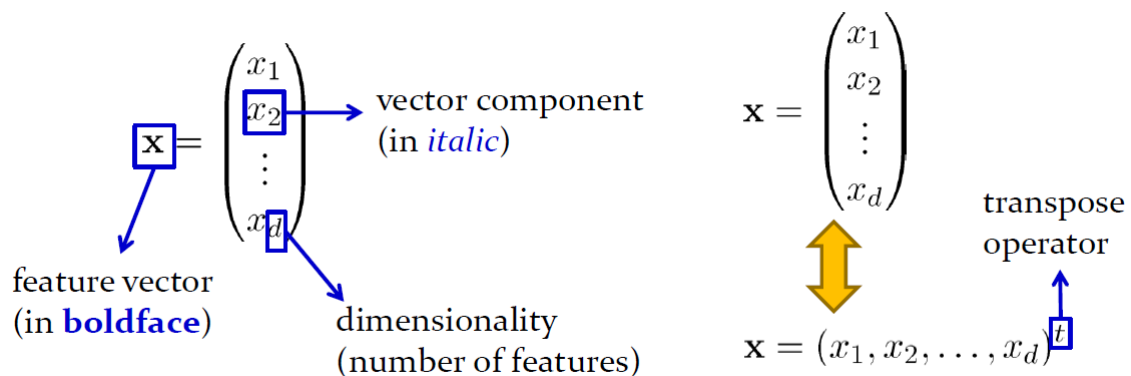
A set of samples to be classified, **usually being mutually exclusive to training set**

## Feature

Attributes which characterize properties of the samples

## Feature Vector

Vector formed by a group of features, **usually in column form**



## Feature Space

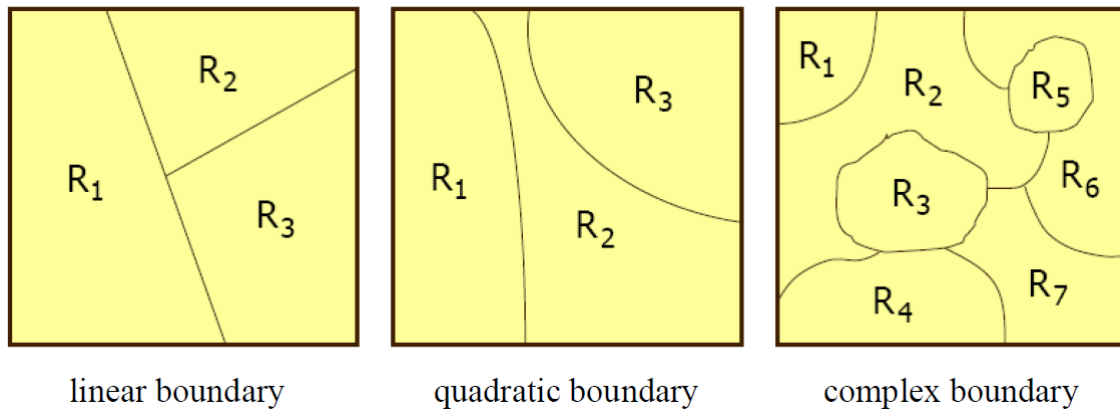
Space containing all the possible feature vectors

## Scatter Plot

Each sample is plotted as a point in the feature space

## Decision Boundary

Boundaries in feature space which separate different categories



## Example: Sea Bass vs Salmon

### Steps

- Preprocessing
- Feature Extraction
- Classification

### Overfitting

Trade of between: Performance on training set / Simplicity of the classifier

### Generalization (泛化能力)

The central aim of designing a classifier is to make correct decisions when presented with novel (unseen/test ) patterns

The ULTIMATE goal

## Related Fields to PR

- Pattern Recognition: Pattern to Category
- Hypothesis Testing: Null hypothesis to Rejection or Not
- Image Processing: Image to Image
- Regression: Pattern to Real Value
- Interpolation: Pattern (unexplored input range) to Interpolated Value
- Density Estimation: Patterns to Probability density function pdf for different categories

## Pattern Recognition System

- Sensing: **converts** physical inputs into digital signal data
- Segmentation: **isolate** sensed objects from the background or from other objects
- Feature Extraction: **measures** object properties that are useful for classification
- Classification: **extracted** features to assign the sensed object to a category
- Post-processing: A post processor **decide on the appropriate action** based on the classification

## Design Cycle of PR System

- Collect Data: A large part of the cost
- Choose Feature: highly domain-dependent / prior knowledge
- Choose Model
- Train Classifier
- Evaluate classifier

# Important Issues

## Noise

### Definition

Any property of the sensed pattern which is not due to the true underlying model but instead to intrinsic randomness of the world or the sensors

- Various types of noise exist
- Noise can reduce the reliability of the feature values measured
- Knowledge of the noise process can help improve performance

## Segmentation

Individual patterns have to be segmented for subsequent pattern recognition operations

One of the **deepest** and **hardest** problems

Different segmentations: e.g. BEATS → BE, BEAT, EAT, AT, EATS...

## Data Collection

- A small set of “typical” examples → Preliminary study of system feasibility
- Much more data → Assure good performance in the fielded system

The Data collected: Is **adequately large**? Is **Representative**?

The efforts of data collection could be rather **demanding**

## Domain Knowledge (prior knowledge)

Type I: Incorporate domain knowledge on the patterns themselves - HARD

To recognize all types of chairs — hard to find **commonness** for chairs

Type II: Incorporate domain knowledge on the pattern generation procedure

Optical character recognition (OCR) → Assume handwritten characters are written as a sequence of strokes

First try to recover stroke representations → deduce the character from the identified strokes

## Feature Extraction

A domain dependent problem which influences the classifier's performance

Good extracted features → Make classification easier

### Distinguishing Capability

Whose values are very **similar for objects in the same category**, while very **different for objects in different categories**

### How to choose features?

- simple to extract
- robust to noise
- lead to simpler decision boundaries

## Pattern Representation

Various ways for pattern representation:

- Statistical: feature vector (the most popular)
- Template Matching: prototype(原型) templates
- Syntactic(句法): rules or grammars

### Desired Properties

- Patterns from the **same classes / different classes** should have **similar/different** representations
- Pattern representations should be **invariant to transformations** such as translations, rotations, resizes, reflections, non rigid deformations
- Intra-class/Inter-class variation should be small/large

## Missing Features

In practical problems, values for certain features may be missing

### Solutions

Naïve method: choose **zero / average** value

Sophisticated method: **regression** techniques

## Model Selection

Each pattern recognition method employs certain **model hypothesis**

Every pattern recognition problem has its own **underlying true model** (not known)

### Fundamental questions on model selection

- How do we know whether the hypothesized model **is (relatively) consistent with** the underlying true model?
- How are we to know to **reject a class** of models and try another one?
- Can we **automate the process of model selection**, instead of trial and error (试错) which is random and tedious?

## Overfitting

We can get **perfect classification performance** on the training data by **choosing complex models**

Complex models are tuned to the **particular training samples**, rather than the **characteristics of the true model**

## Context (上下文)

**Input-dependent** information, other than from the pattern itself

The same pattern within different context might have different meanings

## Classifier Ensemble (分类器集成)

Classifier ensemble aims to **improve generalization performance** by **employing a number of classifiers** for the same task

Also known as Multi classifier System, Mixture of Experts, Classifier Fusion

Diverse ensemble techniques: Bagging, Boosting, Random subspace

## Methods

- Majority voting: vote for the category where most classifiers agree
- Weighted voting: weight each vote by classifier's confidence
- Stacking: learn the rule of combination (more complicated)

## Costs & Risks

**Cost is the loss after making incorrect decisions**

- Equal cost: In OCR, the cost of mistaking "6" as "9" might be equal to that of mistaking "9" as "6"
- Unequal cost: False Negative & False Positive

**Risk is total expected cost which we want to optimize**

- Error rate: percentages of test patterns being wrongly classified

## Questions

- incorporate knowledge of costs
- estimate the lowest possible risk of any classifier

## Computational Complexity

An algorithm scale with

- The number of features (dimensionality)
- The number of training patterns
- The number of possible categories

Brute force (蛮力) approaches might lead to perfect classification, but with **impractical time and storage requirements**