

Set Theory summary

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Set theory

- **Sets**
- Fuzzy sets
- Applications

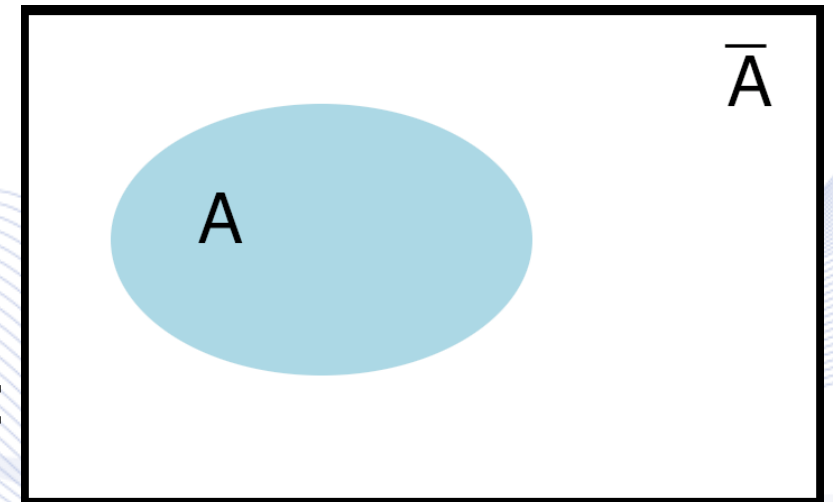
Sets

- A set of n set members $z_i = 1, \dots, n$ is denoted by:

$$\mathcal{A} = \{z_1, \dots, z_n\}.$$

- $|\mathcal{A}| = n$: **set cardinality**.
- If \mathcal{B} is a **subset** of \mathcal{A} , it is denoted as :

$$\mathcal{B} \subset \mathcal{A} \equiv z \in \mathcal{B} \Rightarrow z \in \mathcal{A}.$$



Sets

- Subsets of \mathcal{A} are the null set \emptyset , as well as sets containing any number of set \mathcal{A} members z_i .
- Always:

$$\{\emptyset\} \subset \mathcal{A}, \quad \mathcal{A} \subseteq \mathcal{A}.$$

- If \mathcal{E} is the **universal set** of all sets :

$$\mathcal{A} \subset \mathcal{E}, \quad \forall \mathcal{A}.$$

- **Power set** of a set \mathcal{A} has members all possible subsets of \mathcal{A} .

Sets

- **Set union** $\mathcal{A} \cup \mathcal{B}$ is defined by:

$$z \in \mathcal{A} \text{ or } z \in \mathcal{B} \Rightarrow z \in (\mathcal{A} \cup \mathcal{B}).$$

- **Set intersection** $\mathcal{A} \cap \mathcal{B}$ is defined by:

$$z \in \mathcal{A} \text{ and } z \in \mathcal{B} \Rightarrow z \in (\mathcal{A} \cap \mathcal{B}).$$

Sets

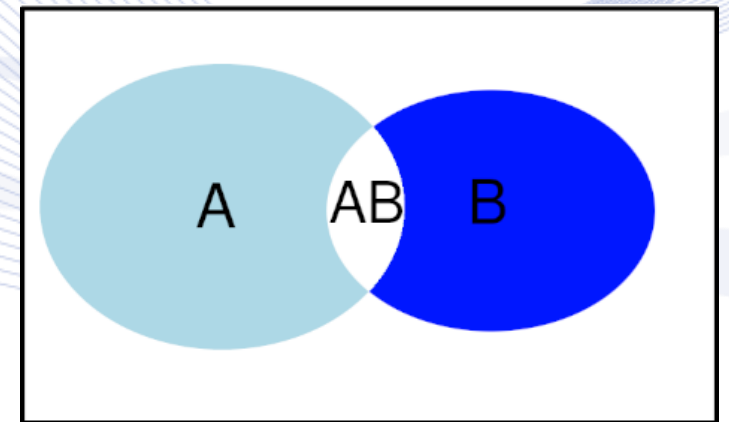
- Set operation properties:

$$A \cup B = B \cup A,$$

$$(A \cup B) \cup C = A \cup (B \cup C) = A \cup B \cup C,$$

$$A \cap B = B \cap A,$$

$$(A \cap B) \cap C = A \cap (B \cap C) = A \cap B \cap C.$$



Sets

- **Borel Field** (σ -field) is a field having the following additional property:

$$\mathcal{A}_i \in \mathcal{F} \Rightarrow (\cup_i \mathcal{A}_i) \in \mathcal{F} \Rightarrow (\cap_i \mathcal{A}_i) \in \mathcal{F}$$

for any possible countable infinite sequence \mathcal{A}_i .

- It can be proven that the collection of all \mathcal{F} subsets of the set \mathcal{E} is a Borel Field.

Set theory

- Sets
- **Fuzzy sets**
- Applications

Fuzzy sets

- Let \mathcal{E} is a universal set of objects $z \in \mathcal{E}$.
- Crisp set \mathcal{A} ($\mathcal{A} \subset \mathcal{E}$) membership function $A_i(z): \mathcal{E} \rightarrow \{0,1\}$:

$$A_i(z) = \begin{cases} 1, & \text{if } z \in \mathcal{A} \\ 0, & \text{if } z \notin \mathcal{A}. \end{cases}$$

- **Fuzzy set \mathcal{A} ($\mathcal{A} \subset \mathcal{E}$) membership function:**
 $A_i(z): \mathcal{E} \rightarrow [0,1]$.

Fuzzy sets

- Fuzzy set \mathcal{A} is fully characterized as follows :

$$\mathcal{A} = \{(z, A_i(z)), z \in \mathcal{E}\}.$$

- When $\mathcal{E} = \{z_1, \dots, z_n\}$ is a finite element set, a fuzzy set \mathcal{A} is expressed by:

$$\mathcal{A} = A_i(z_1)/z_1 + \dots + A_i(z_n)/z_n.$$

Set theory

- Sets
- Fuzzy sets
- **Applications**

Images as sets

- An image domain is set:
- Continuous images: $\mathcal{X} \in \mathbb{R}^2$.
- digital images (set of *pixels*): $\mathcal{X} \in \mathbb{Z}^2$.
- Typically is $\mathcal{X} \in \mathbb{Z}^2$ is a $N \times M$ rectangle of pixels.

Volumetric images:

- Their domain (set of *voxels*): $\mathcal{X} \in \mathbb{Z}^3$.



Image domain \mathcal{X} .

Mathematical morphology



Erosion and dilation are special cases of ***Minkowski set addition*** and ***Minkowski set subtraction***:

$$\mathcal{X} \oplus \mathcal{B}^s = \bigcup_{b \in \mathcal{B}} \mathcal{X}_b, \quad \mathcal{X} \ominus \mathcal{B}^s = \bigcap_{b \in \mathcal{B}} \mathcal{X}_b.$$



(a)



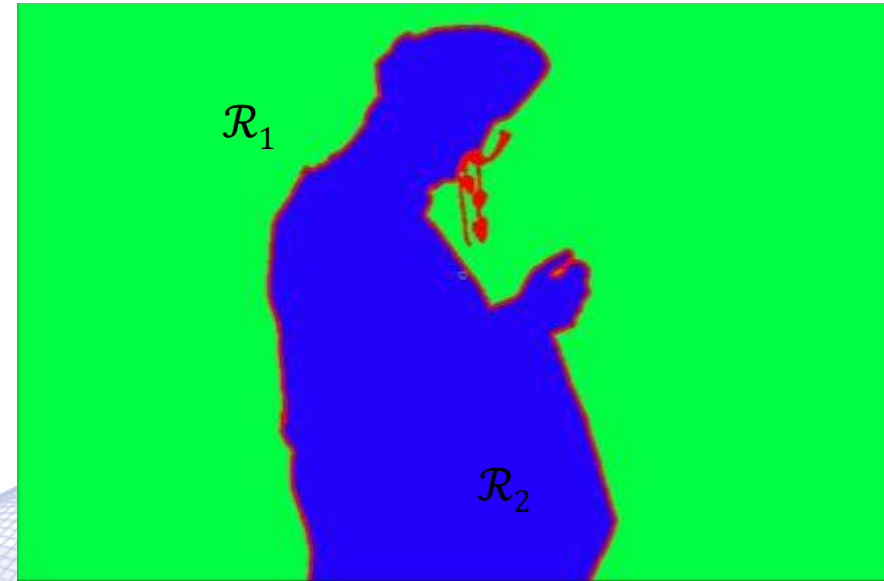
(b)



(c)

a) Thresholded image; b) eroded and c) dilated image by the structuring elements SQUARE.

Image segmentation



Foreground and background segmentation.

Object Detection Performance Metrics



Intersection over Union (IoU):

$$J(\mathcal{A}, \mathcal{B}) = |\mathcal{A} \cap \mathcal{B}| / |\mathcal{A} \cup \mathcal{B}|.$$

- \mathcal{A}, \mathcal{B} : estimated, ground truth ROIs (sets, bounding boxes).
- $|\mathcal{A}|$: set cardinality (area counted in pixels)
- Also called **Jaccard Similarity Coefficient** or **Overlap Score**.

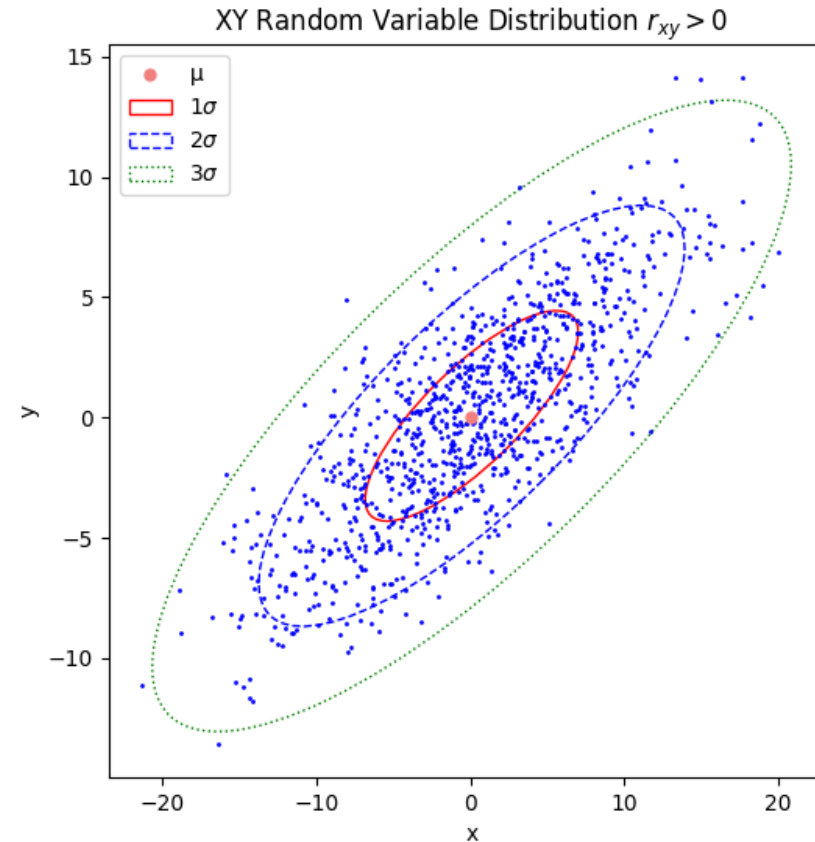
Object Detection Performance Metrics



Object detection: a) $J(\mathcal{A}, \mathcal{B}) = 0.67$; b) $J(\mathcal{A}, \mathcal{B}) = 0.27$.

Data sets

- An object can be represented by a real-valued feature vectors: $\mathbf{x} \in \mathbb{R}^n$.
- A data set is a feature vector set: $\mathcal{D} = \{\mathbf{x}_i\}_{i=1}^N$.



Data set \mathcal{D} .

Crisp/fuzzy set clustering

Let \mathcal{D} be a feature data set: $\mathcal{D} = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n\}$.

- **Crisp clustering** is the partition of \mathcal{D} into m sets $\mathcal{C}_1, \dots, \mathcal{C}_m$, satisfying the following conditions:

- $\mathcal{C}_i \neq \emptyset, \quad i = 1, \dots, m,$
- $\bigcup_{i=1}^m \mathcal{C}_i = \mathcal{D},$
- $\mathcal{C}_i \cap \mathcal{C}_j = \emptyset, \quad i \neq j, \quad i, j = 1, \dots, m.$

- Feature vectors \mathbf{x}_i in a cluster \mathcal{C}_i are ‘similar’, while they are ‘dissimilar’ to the ones of other clusters $\mathcal{C}_j, i \neq j$.