

Introduction to Subgroup Discovery

NOMAD SUMMER
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MAX-PLANCK-GESELLSCHAFT

Two flavors of data science



Predictive modelling



Data analytics

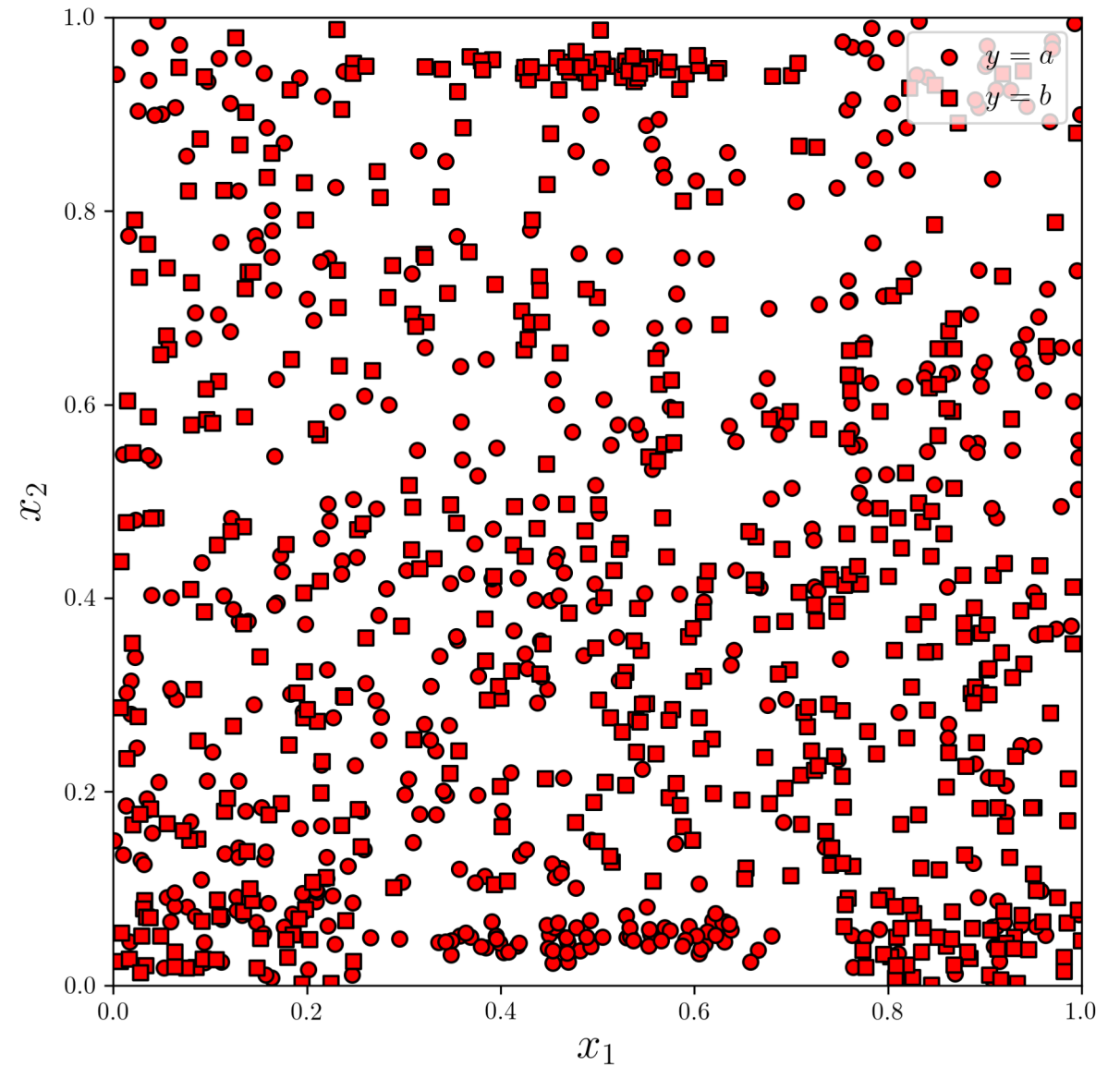
Basic setup

Given

Sample $S \subseteq P$

Target variable $y: P \rightarrow \{a, b, c, \dots\}$

Features $x_j: P \rightarrow X_j$



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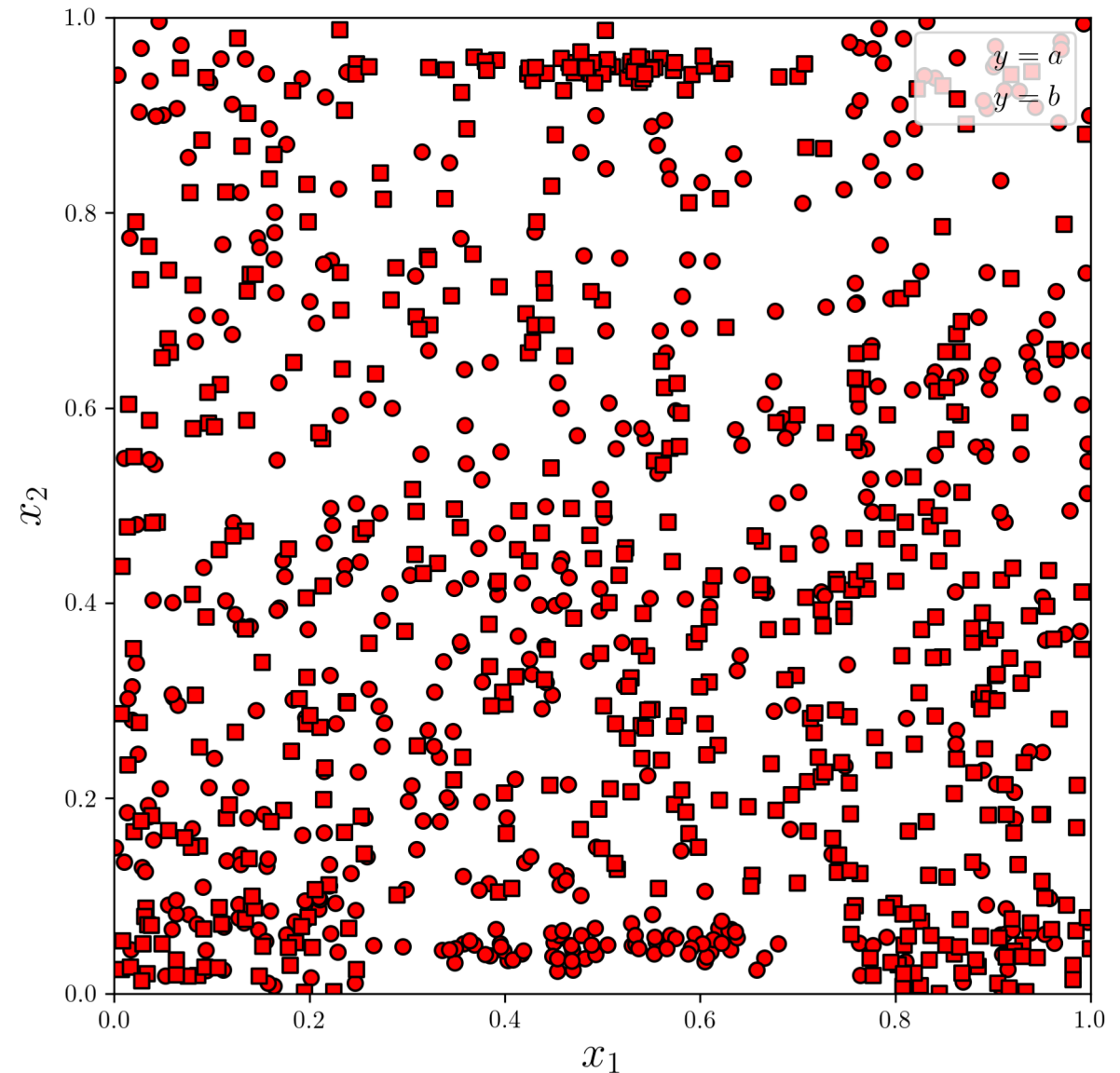
Predictive modeling

Model class $M \subseteq X \rightarrow Y$

Error measure $\text{err}: Y \times Y \rightarrow \mathbb{R}_+$

Minimize

$$f(m) = \sum_{i \in S} \text{err}(m(x(i)), y(i)) / |S| + \gamma \|m\|$$



Global models classify whole space

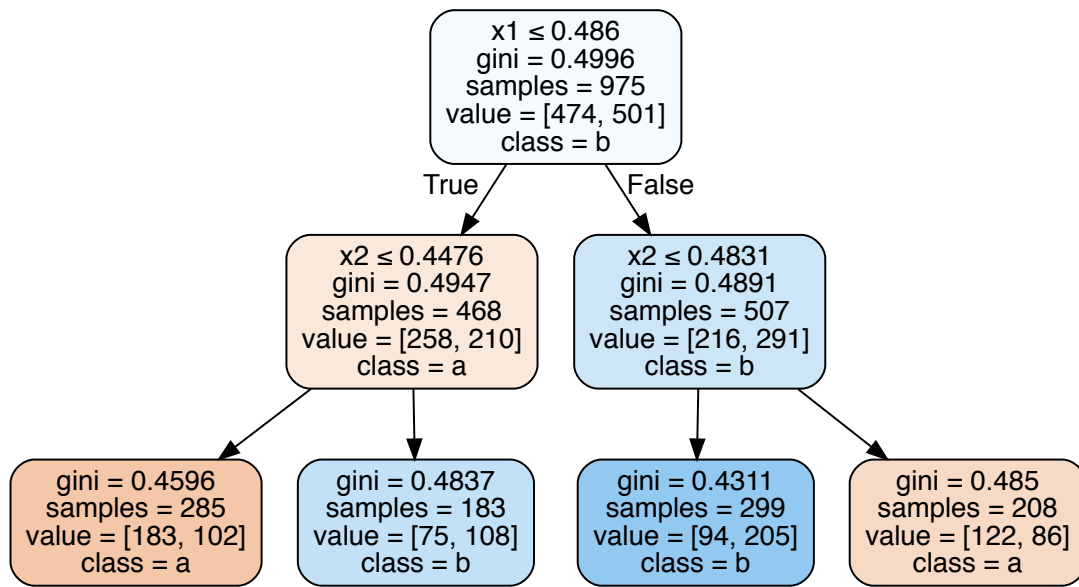
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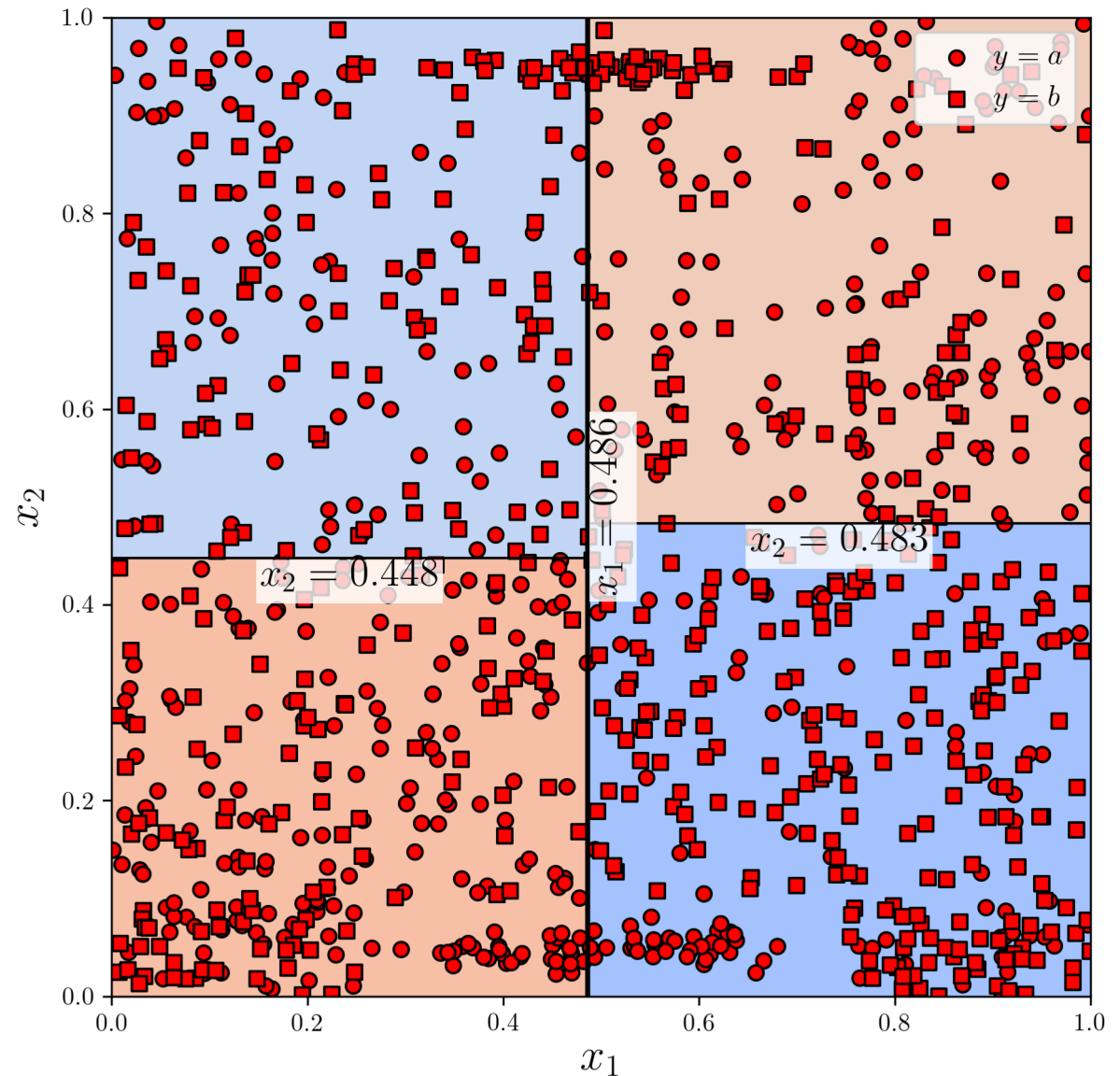
Target variable $y: P \rightarrow \{a, b, c, \dots\}$

Features $x_j: P \rightarrow X_j$

Decision tree



Misses interesting local phenomena



Subgroup discovery focusses on local observations

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Sample $S \subseteq P$

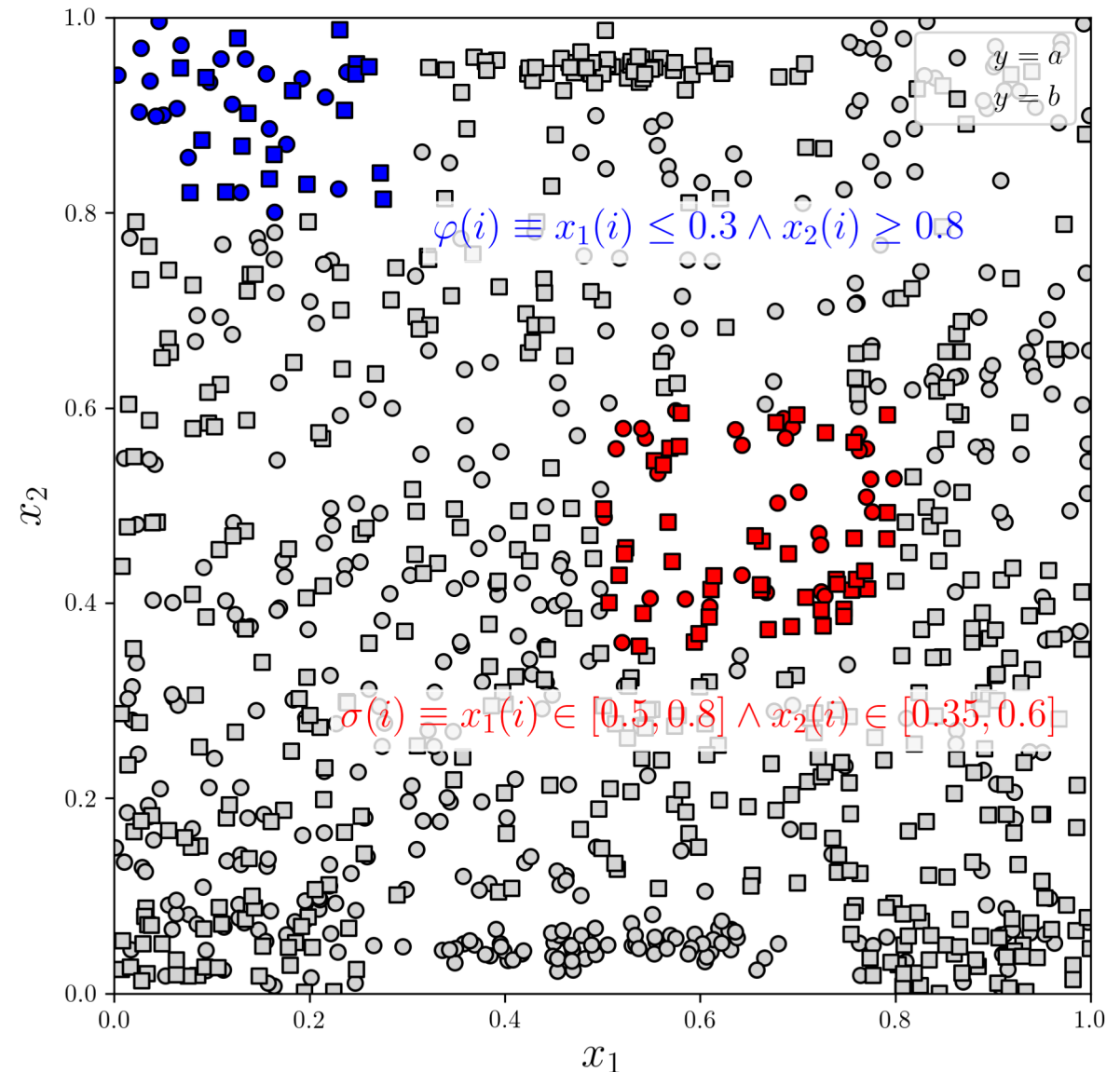
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Define

Propositions $\Pi_x = \{\pi_1, \dots, \pi_k\}$

Selection language $\mathcal{L}_x = \{\sigma(i) = \pi_{j_1}(i) \wedge \dots \wedge \pi_{j_l}(i)\}$



Subgroup discovery focusses on local observations

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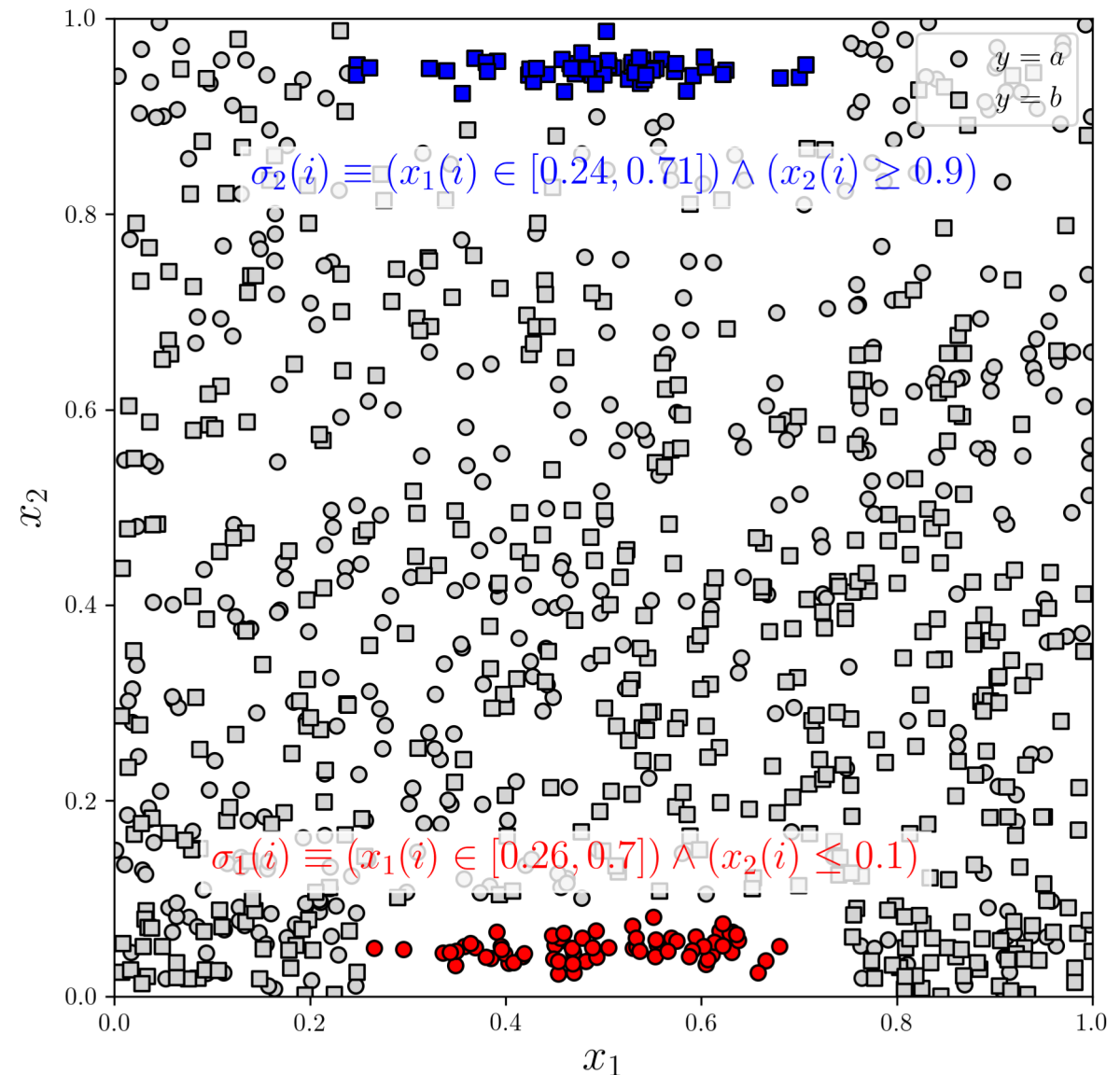
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Optimize

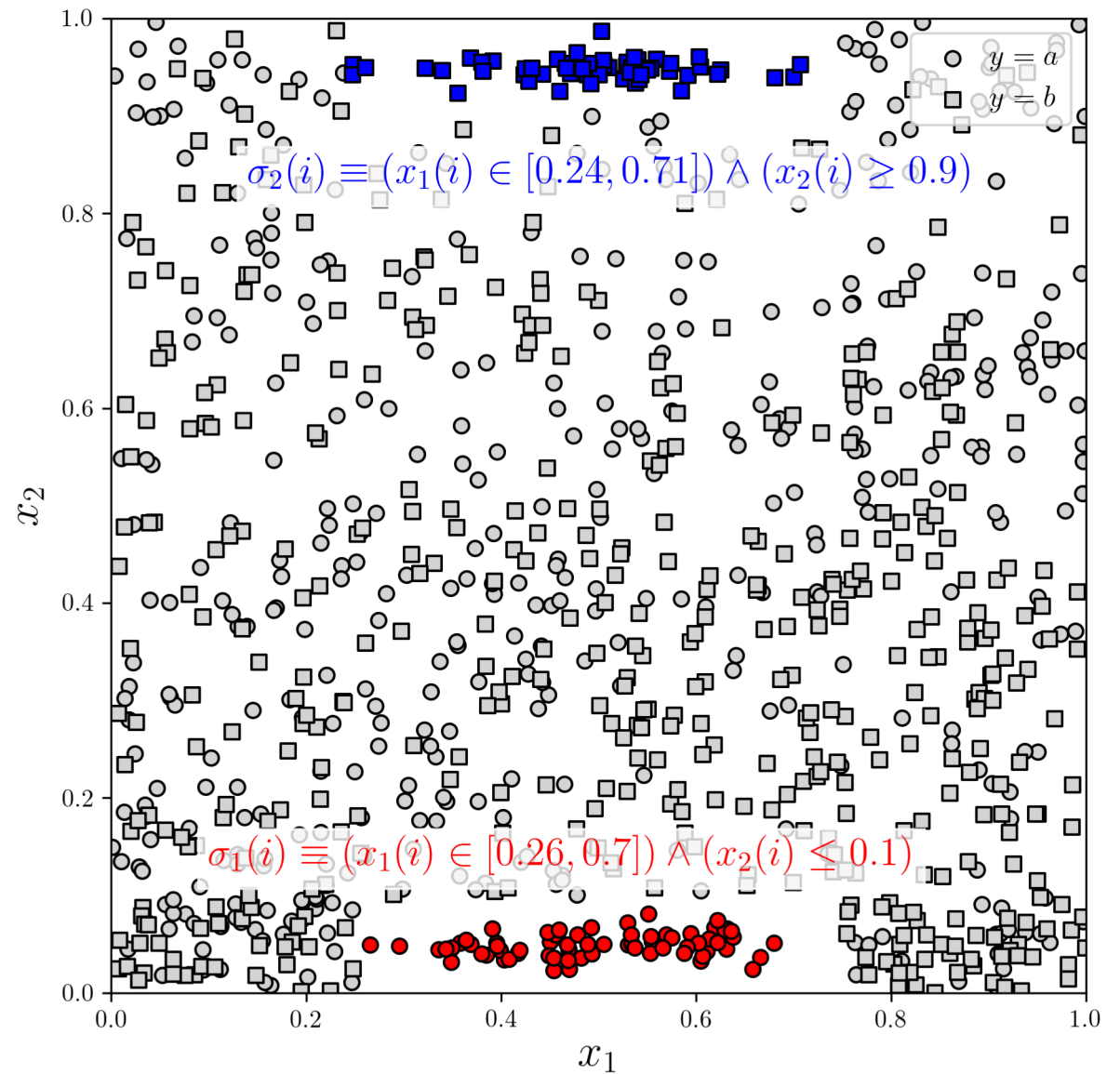
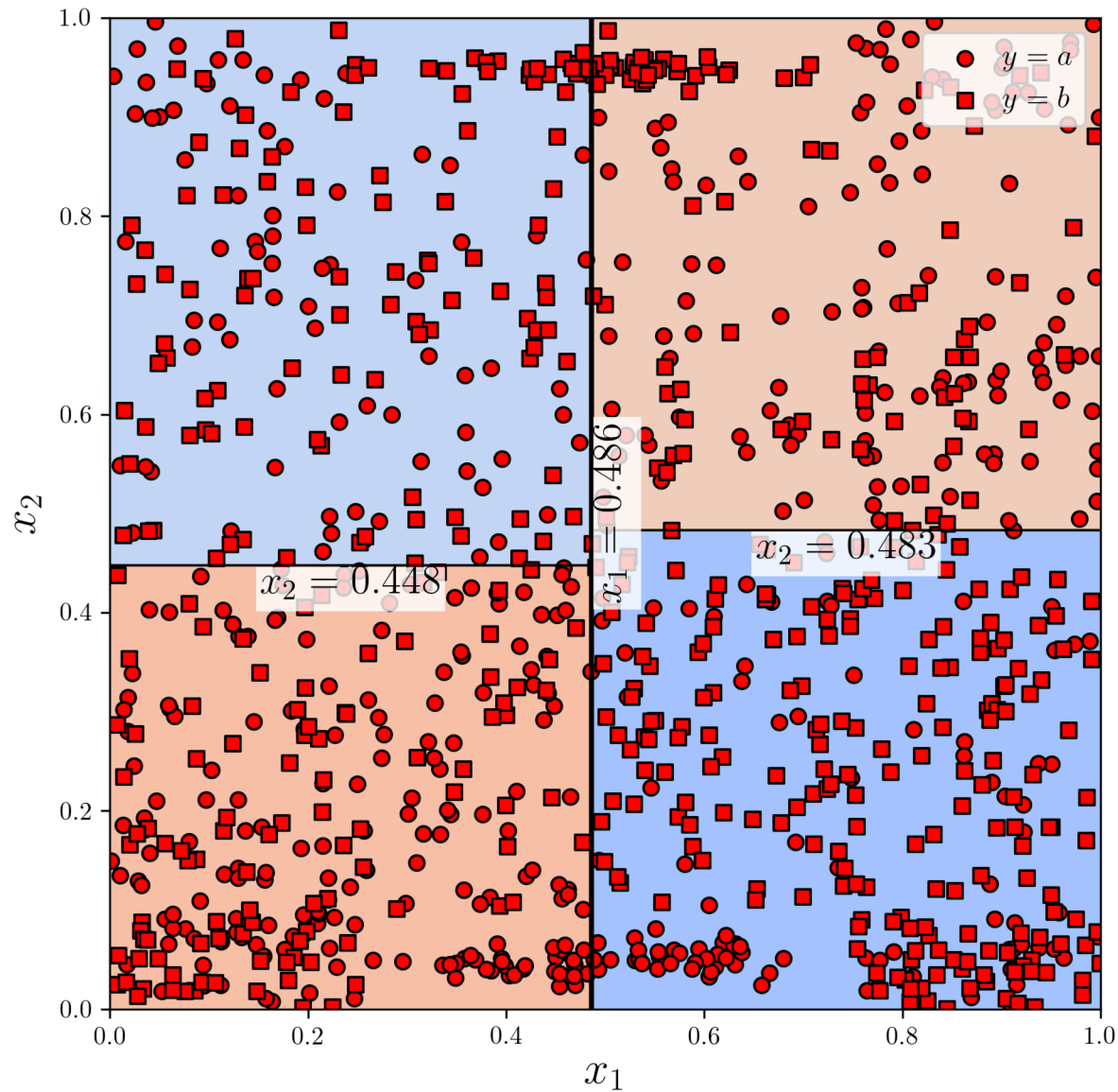
$$f(Q) = \text{cov}(Q)^y \text{eff}(Q)_+$$

with

- $Q = \{i \in S: \sigma(i) = \top\}$ **extension**
- $\text{cov}(Q) = |Q|/|S|$ **coverage**
- $\text{eff}(Q) = (H_y(S) - H_y(Q)) / H_y(S)$ **effect**
- $H_y(Q) = -\sum_v p_Q(y = v) \log p_Q(y = v)$ **entropy**



Subgroup discovery focusses on local observations



Application 1: octet binary crystal structures

Population

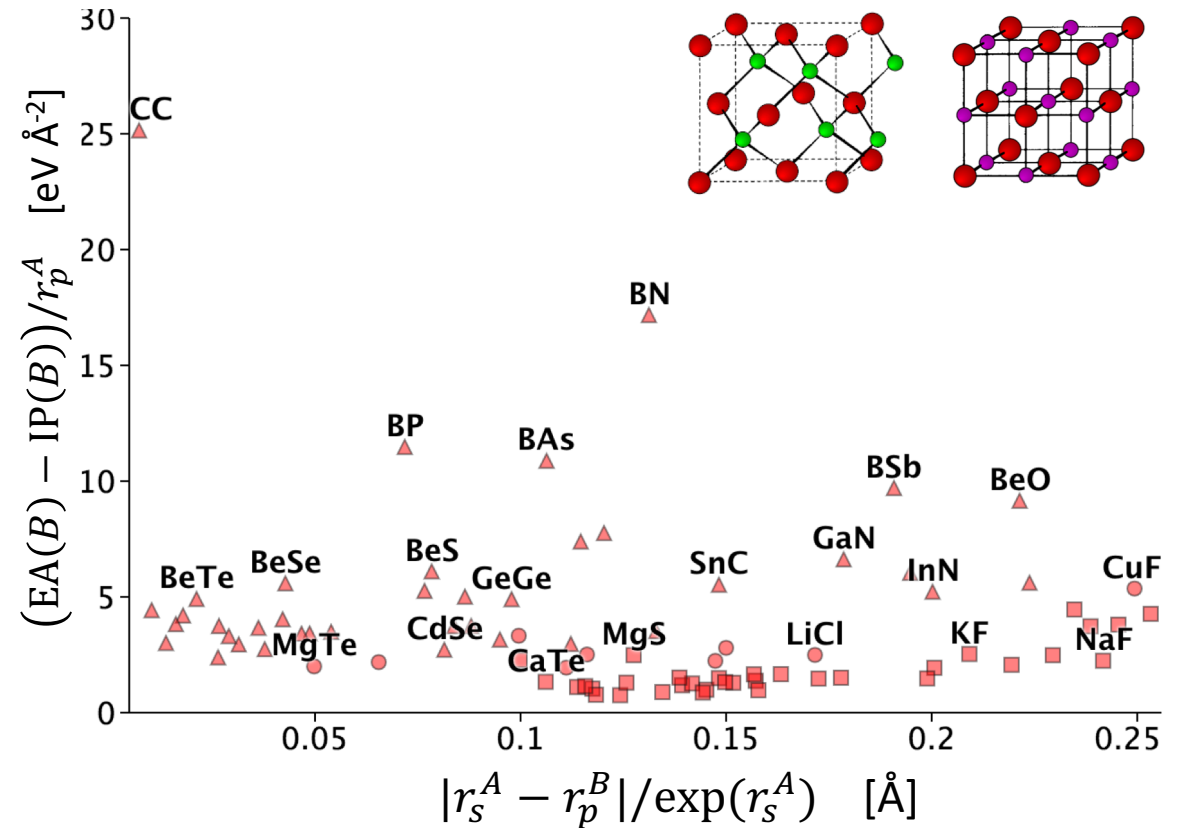
$$P = \{AB: A = \text{Ag, Al, Ba, ...} \wedge B = \text{Br, Cl, F, ...}\}$$

Target

$$y = \text{sign}(\Delta E) \text{ where } \Delta E = E_{\text{RS}} - E_{\text{ZB}}$$

Features

$$x \in \{IP^A, EA^A, r_s^A, r_p^A, r_d^A, IP^B, EA^B, r_s^B, r_p^B, r_d^B, IP^A - IP^B, EA^A - EA^B, |r_s^A - r_s^B|, \dots\}$$



Application 1: octet binary crystal structures

Population

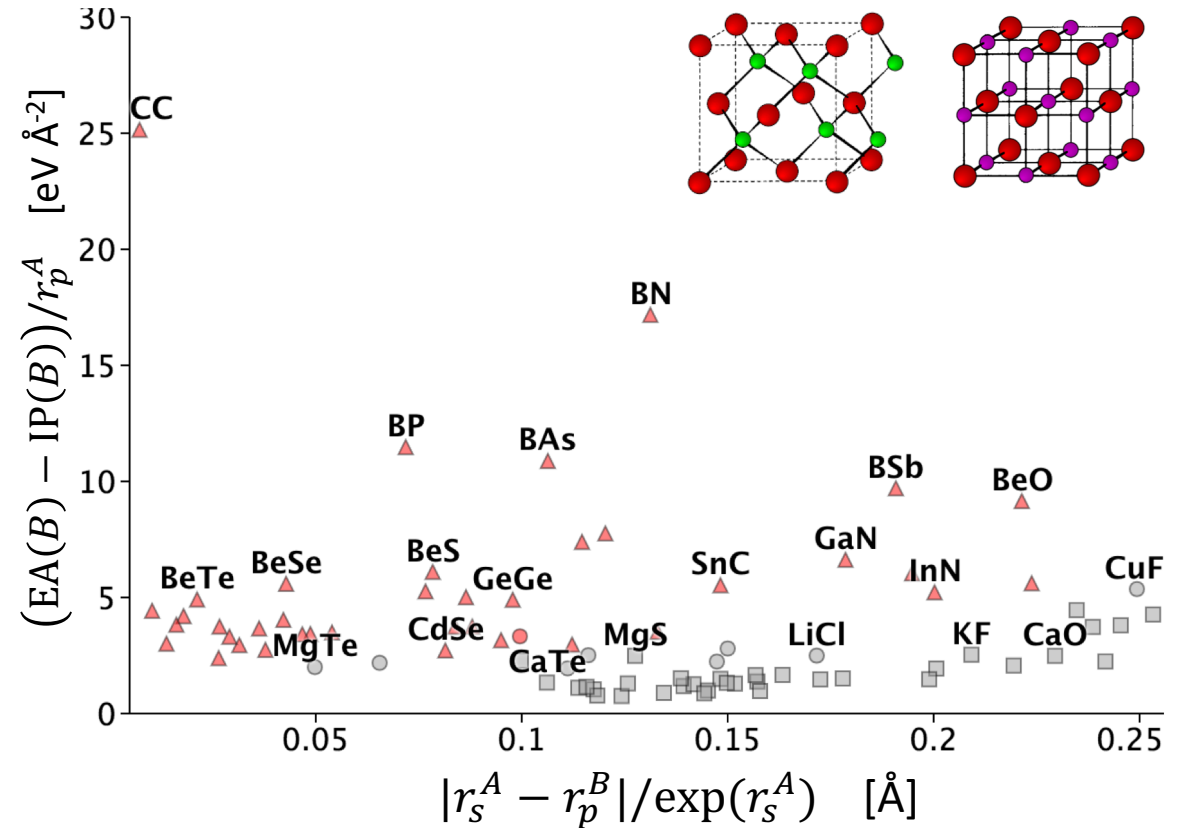
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Selector

$$\sigma_{\text{ZB}} \equiv (|r_p^A - r_p^B| \leq 1.15) \wedge (r_s^A \leq 1.27)$$

Parameters

$$\text{cov} = 40/82 \quad \text{eff} = 1 \quad [H_y(\sigma_{\text{ZB}}) = 0, H_y(P) = 1]$$

Application 1: octet binary crystal structures

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Population

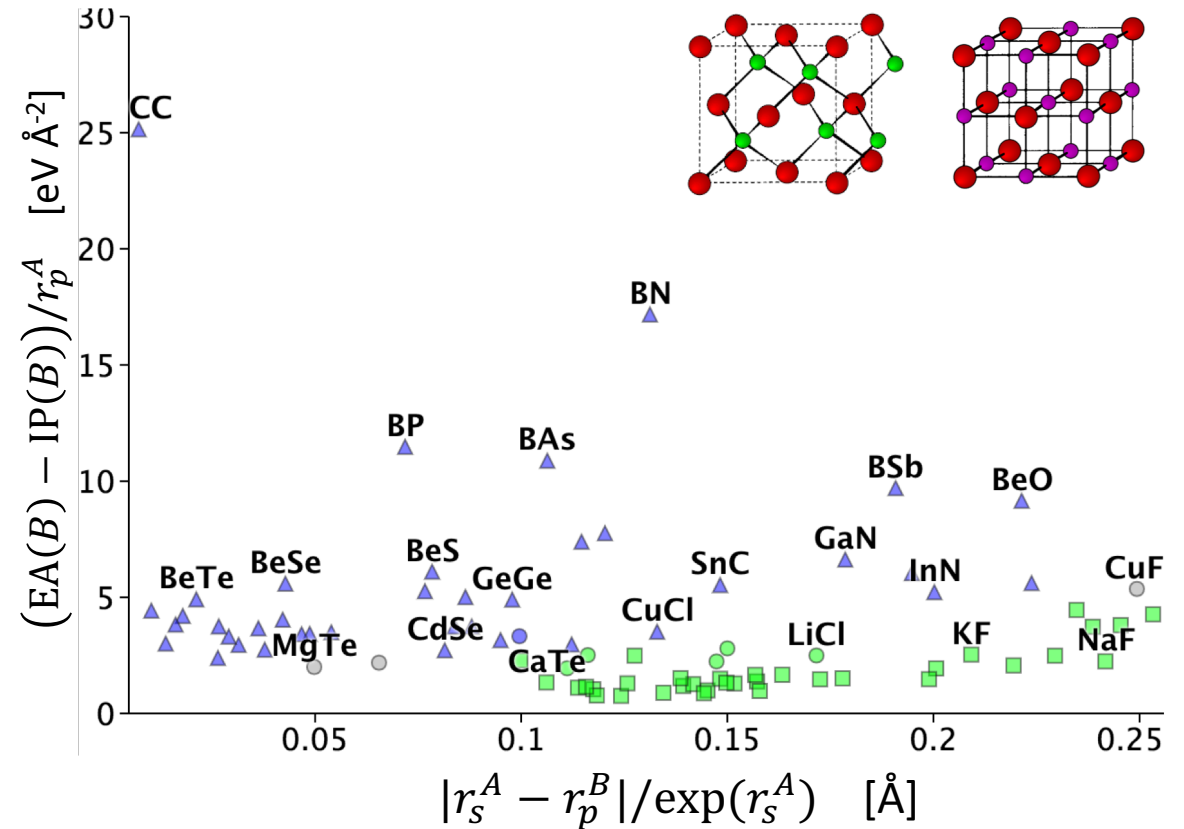
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$$\sigma_{\text{ZB}} \equiv (|r_p^A - r_p^B| \leq 1.15) \wedge (r_s^A \leq 1.27) \quad \sigma_{\text{RS}} \equiv (|r_p^A - r_p^B| \geq 0.91) \wedge (r_s^A \geq 1.22)$$

Parameters

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$$\text{cov} = 39/82 \quad \text{eff} = 1 \quad [H_y(\sigma_{\text{RS}}) = 0, H_y(P) = 1]$$

Application 1: octet binary crystal structures

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Population

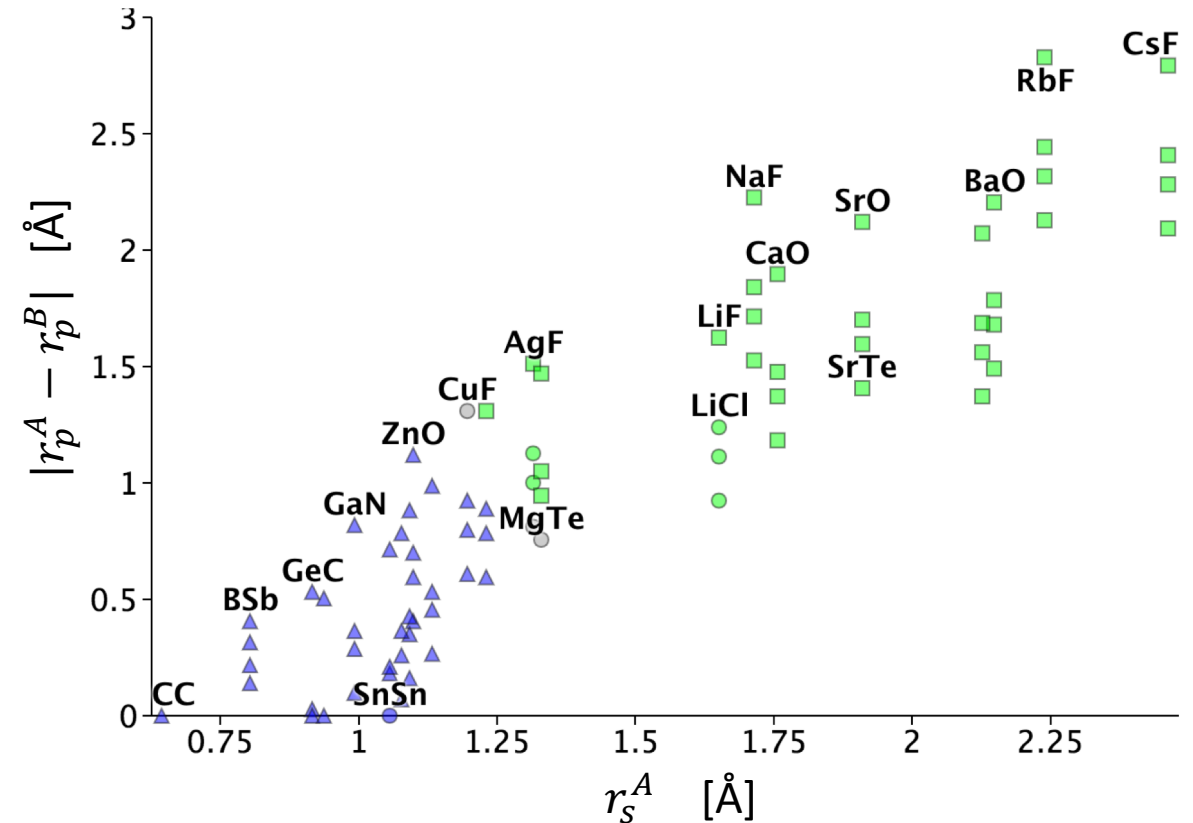
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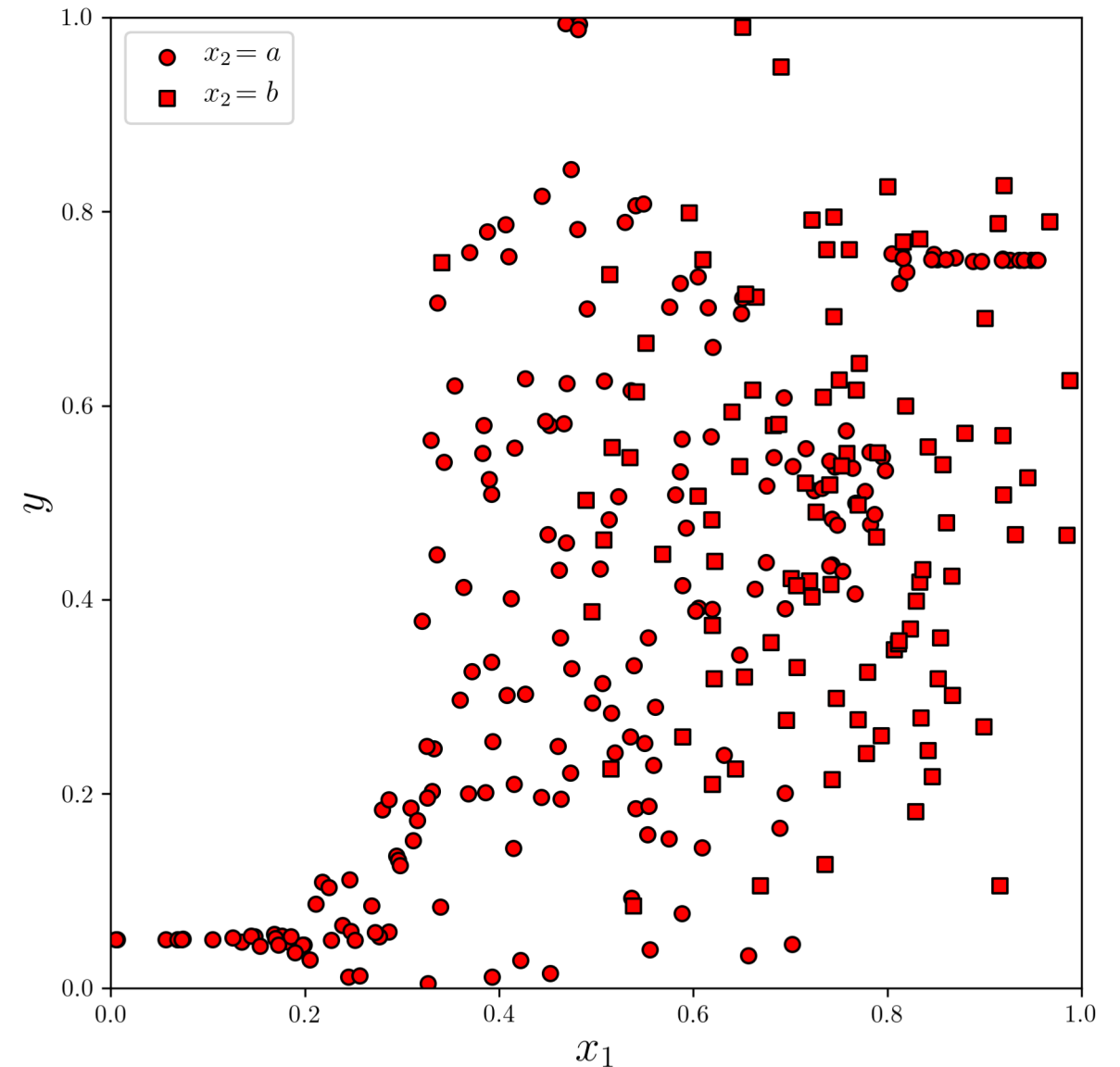
Subgroup discovery for real-valued targets

Given

Sample $S \subseteq P$

Target variable $y: P \rightarrow \mathbb{R}$

Features $x_j: P \rightarrow X_j$



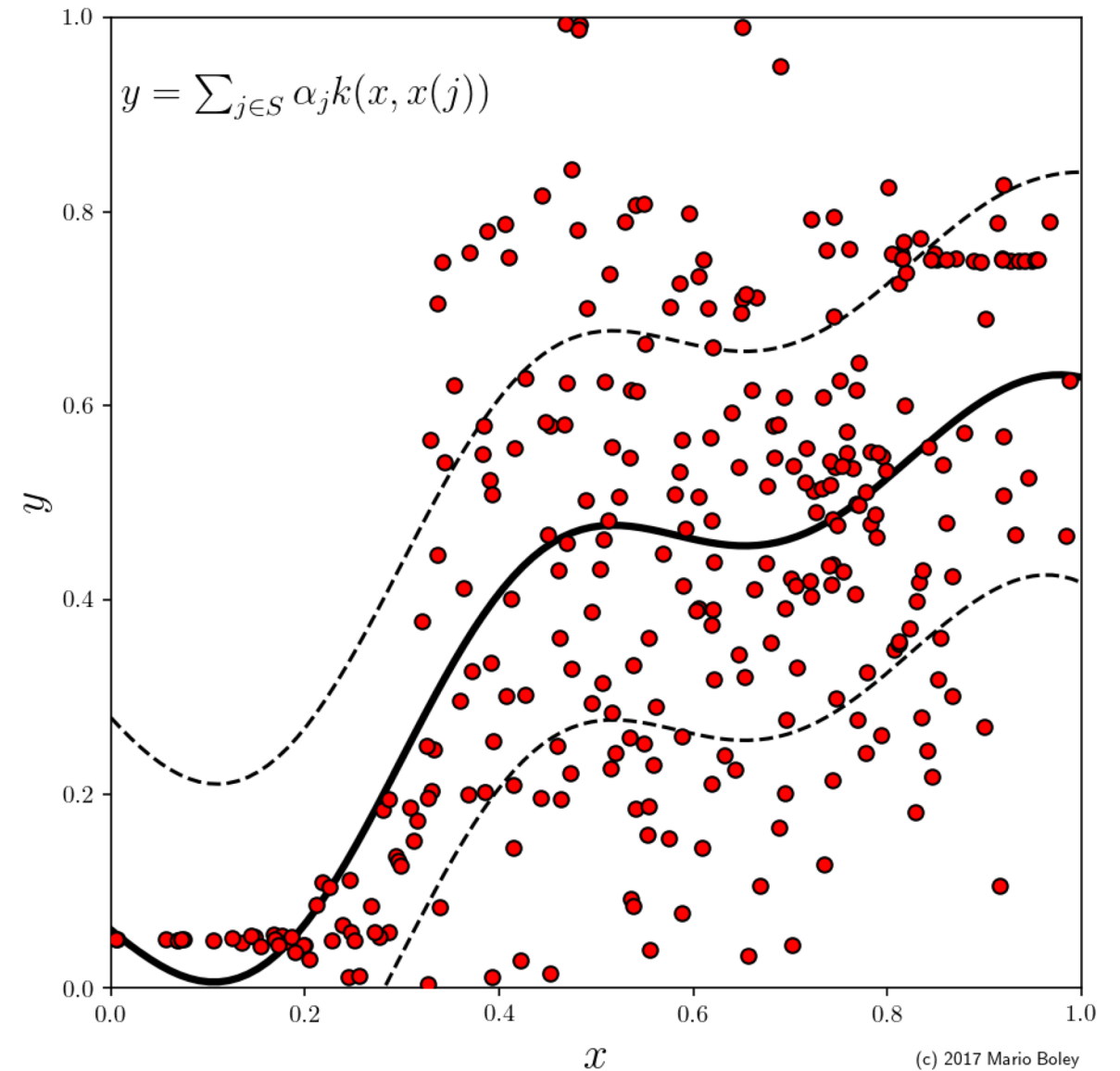
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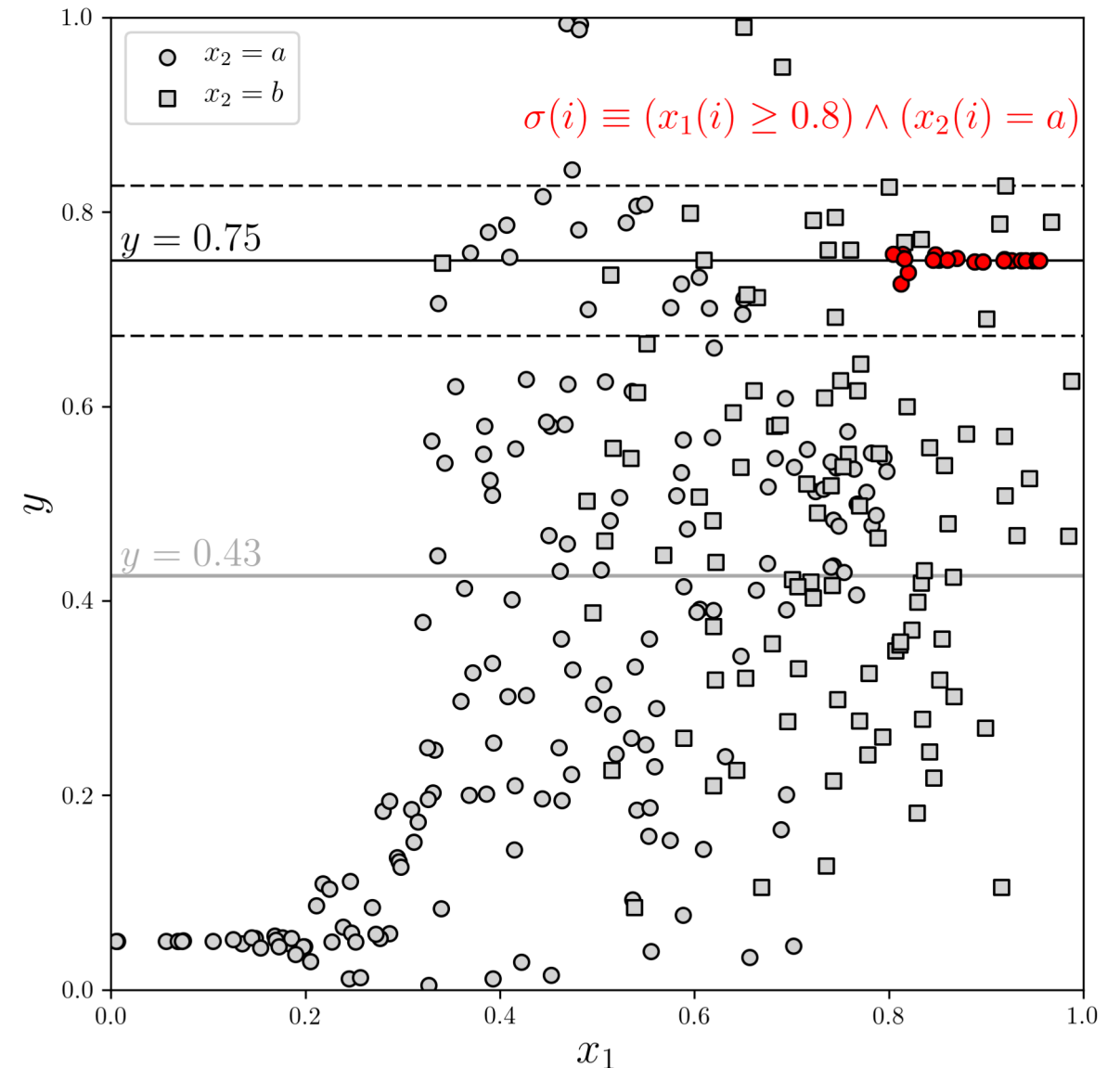
Selection language $\mathcal{L}_x = \{\sigma(i) = \pi_{j_1}(i) \wedge \dots \wedge \pi_{j_l}(i)\}$

Optimize

$$f(Q) = \text{cov}(Q)^y \text{eff}(Q)_+$$

with

- $Q = \{i \in S: \sigma(i) = \top\}$ extension
- $\text{cov}(Q) = |Q|/|S|$ coverage
- $\text{eff}(Q) = (s_y(S) - s_y(Q)) / s_y(S)$ effect
- $s_y(Q) = \sqrt{\sum_{i \in Q} (\bar{y} - y(i))^2 / (|Q| - 1)}$ std. dev.



Application 2: Au structure/property relationship

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Population

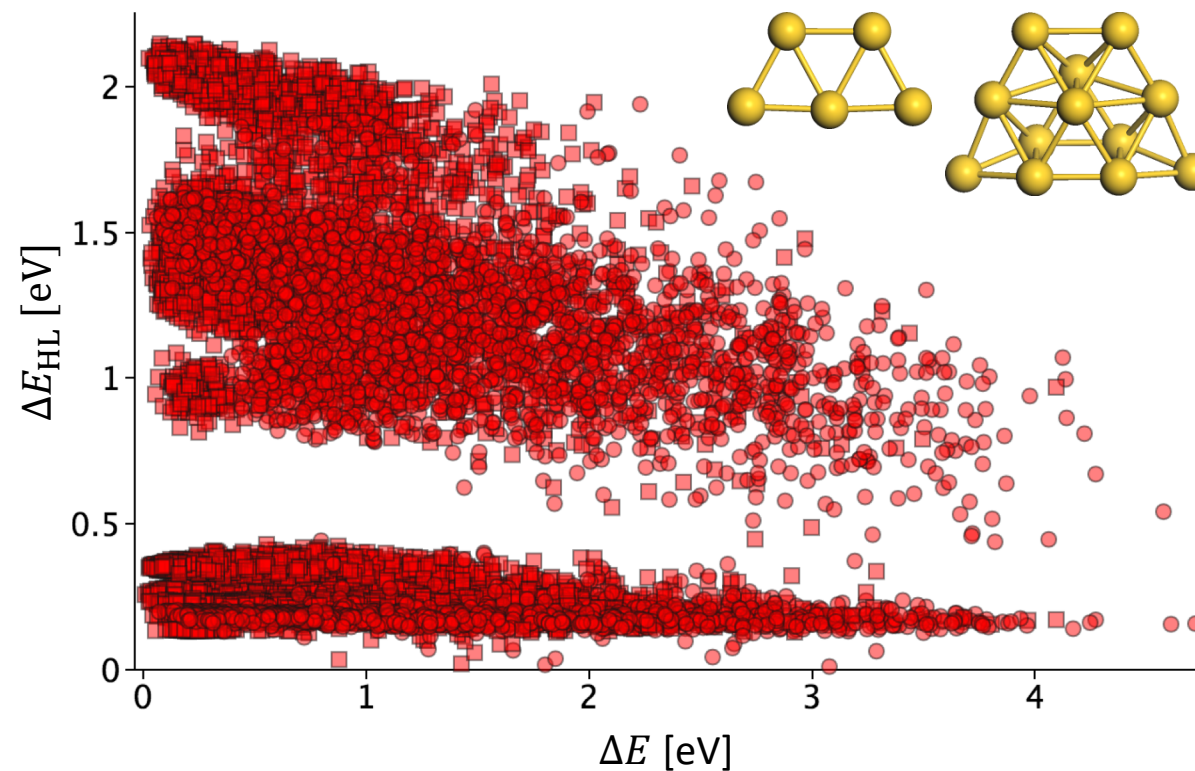
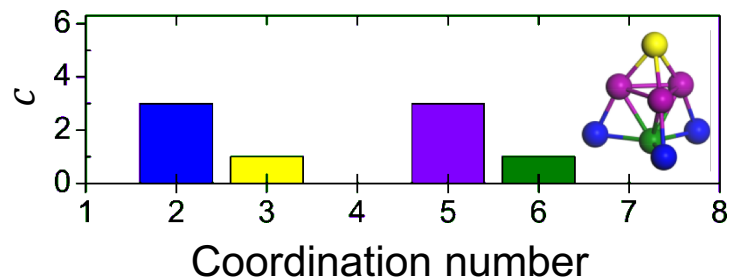
$$P = \{c: c \text{ conf. of Au5} - \text{Au14}\}$$

Target

$$y = \Delta E_{\text{HL}} \text{ HOMO-LUMO energy gap}$$

Features

$$x \in \{a, c_1, c_2, c_3, c_4, c_5, c_6, r, \text{shape}, \text{Mo}_{\text{co}}, \text{Me}_{\text{co}}\}$$



Application 2: Au structure/property relationship

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Population

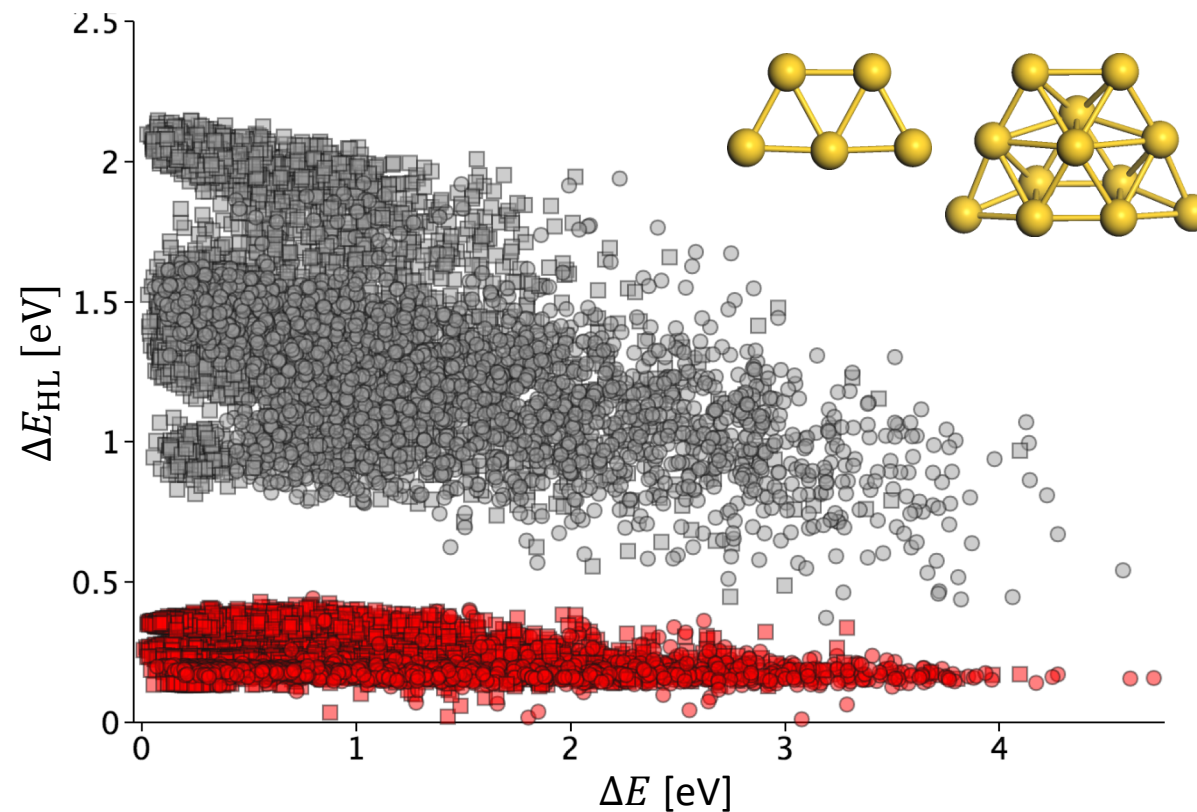
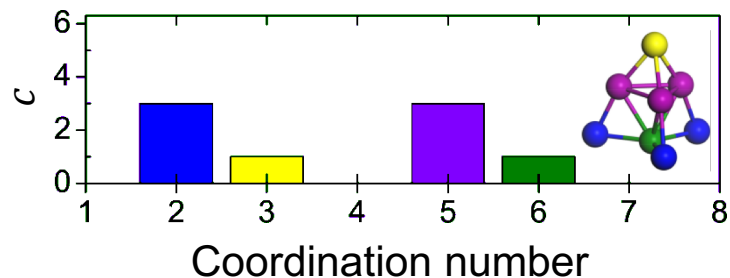
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Selector

$$\sigma(i) \equiv \text{odd}(a(i))$$

Parameters

$$\text{cov}(\sigma) = 0.5 \quad \text{eff}(\sigma) = 0.9$$

$$[s_y(Q) = 0.06, s_y(S) = 0.58]$$

$$[\bar{y}(Q) = 0.22, \bar{y}(S) = 0.42]$$

Application 2: Au structure/property relationship

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Population

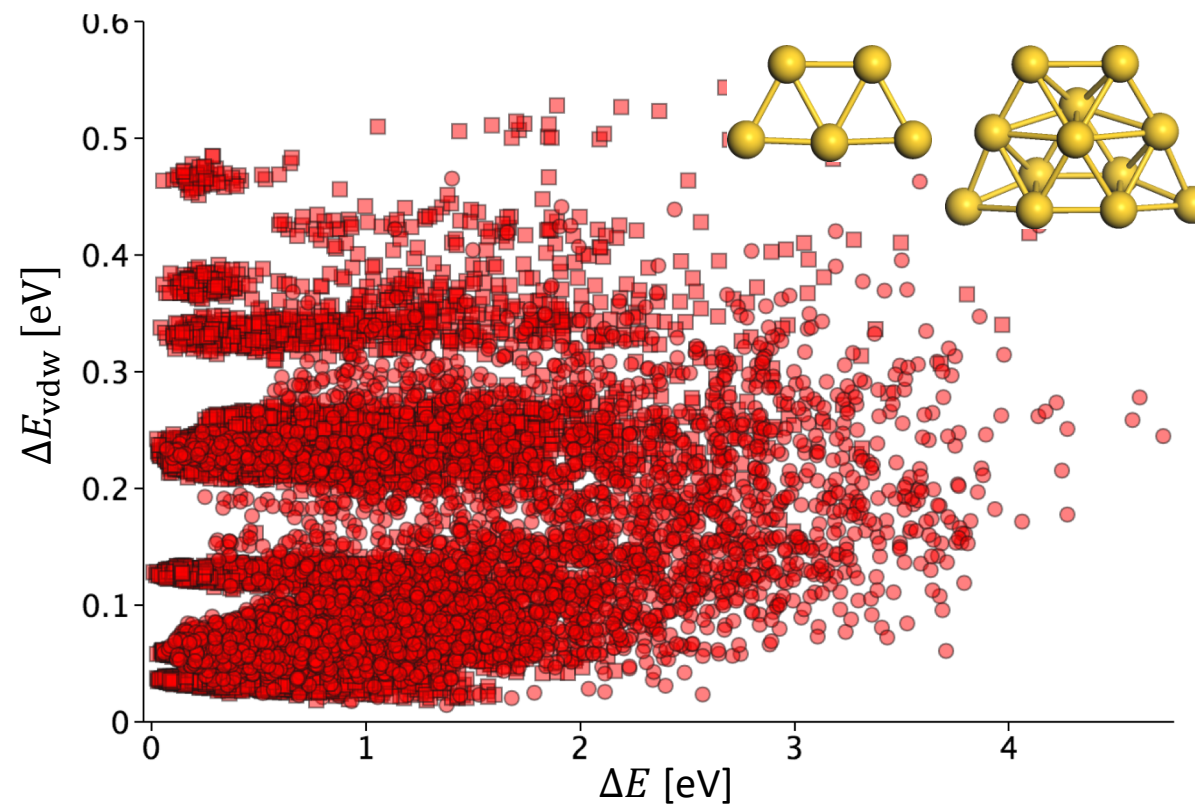
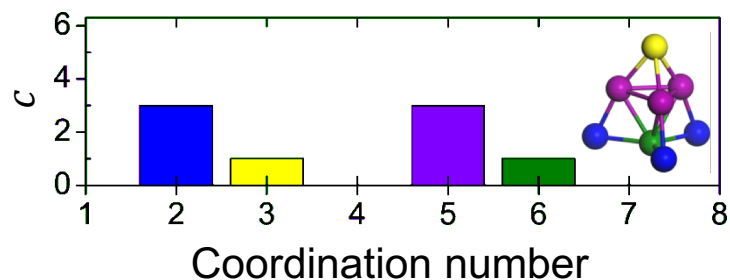
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Application 2: Au structure/property relationship

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Population

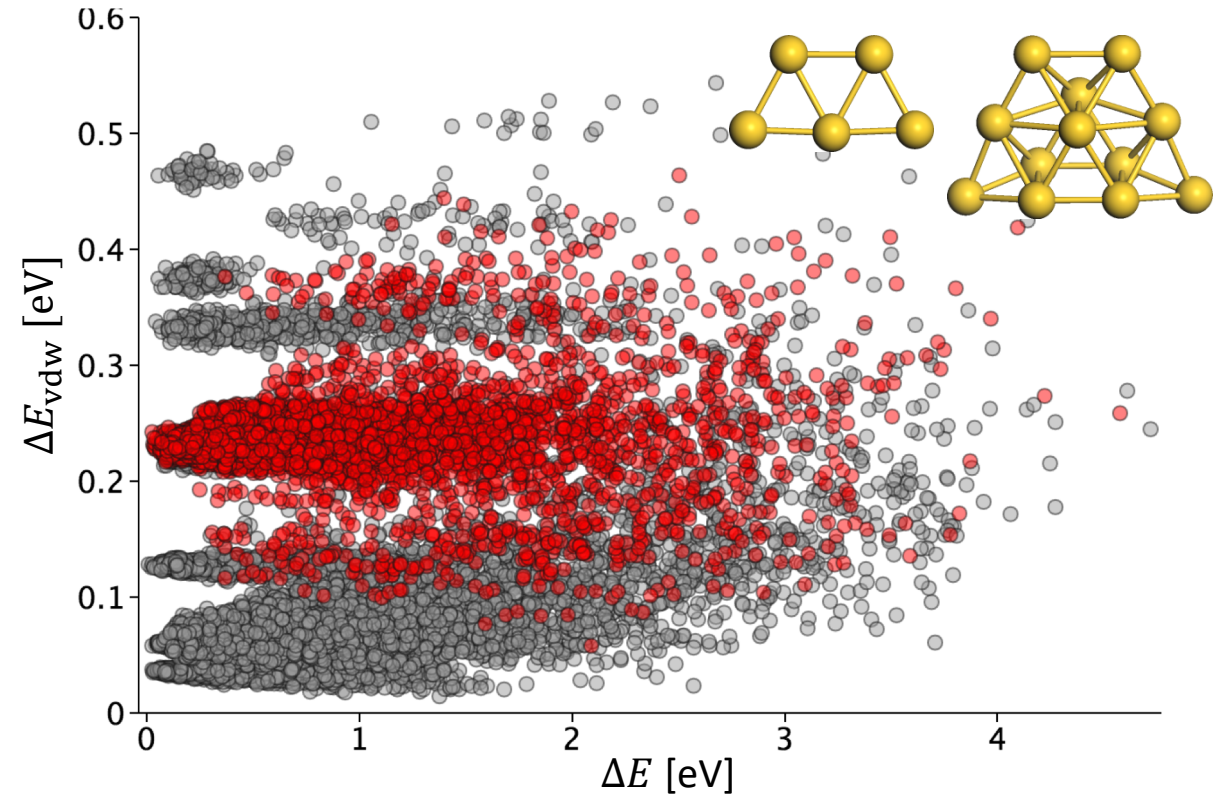
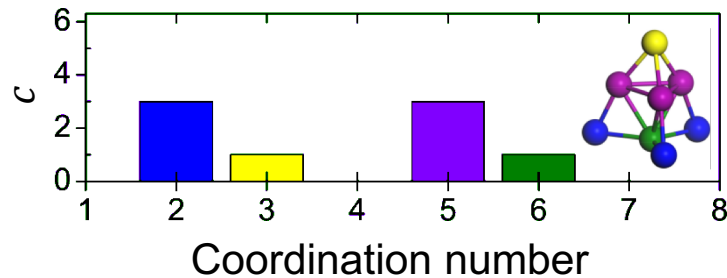
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Features

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Selector

$$\sigma(i) \equiv a(i) \in [8,12] \wedge c_2(i) > 0.17 \wedge c_6(i) < 0.28 \wedge r(i) > 0.86$$

Parameters

$$\text{cov}(\sigma) = 0.2 \quad \text{eff}(\sigma) = 0.68 \quad [s_y(Q) = 0.03, s_y(S) = 0.09]$$
$$[\bar{y}(Q) = 0.23, \bar{y}(S) = 0.13]$$

Subgroup discovery with multiple targets

Given

Sample $S \subseteq P$

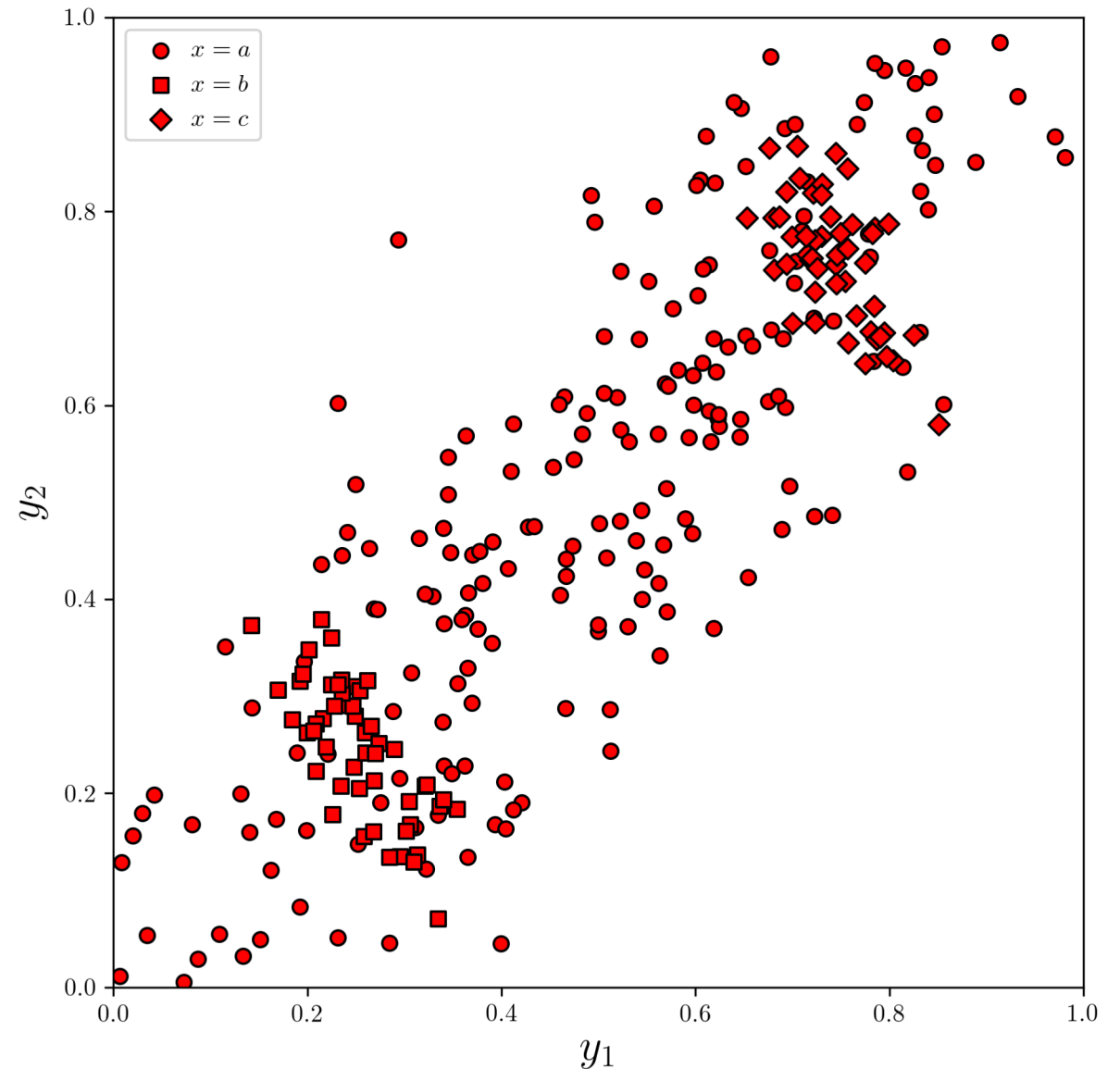
Target variable $y_1, y_2: P \rightarrow \mathbb{R}$

Features $x_j: P \rightarrow X_j$

Define

Propositions $\Pi_x = \{\pi_1, \dots, \pi_k\}$

Selection language $\mathcal{L}_x = \{\sigma(i) = \pi_{j_1}(i) \wedge \dots \wedge \pi_{j_l}(i)\}$



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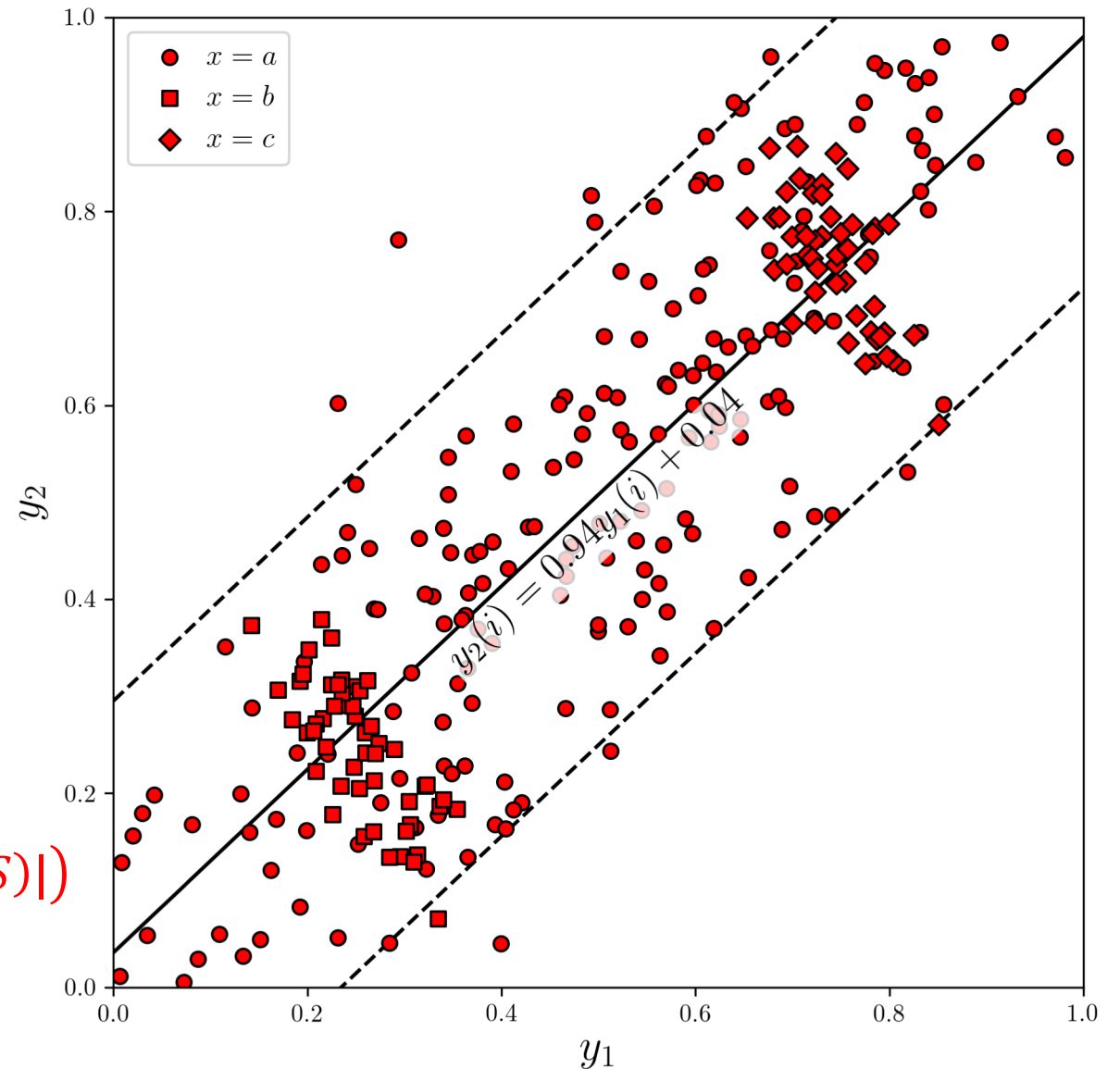
Selection language $\mathcal{L}_x = \{\sigma(i) = \pi_{j_1}(i) \wedge \dots \wedge \pi_{j_l}(i)\}$

Optimize

$$f(Q) = \text{cov}(Q)^y \text{eff}(Q)_+$$

with

- $Q = \{i \in S: \sigma(i) = T\}$
- $\text{cov}(Q) = |Q|/|S|$
- $\text{eff}(Q) = (|r_{y_1, y_2}(Q)| - |r_{y_1, y_2}(S)|) / (1 - |r_{y_1, y_2}(S)|)$
- $r(Q) = \frac{1}{|Q|-1} \sum_{i \in Q} \left(\frac{\bar{y}_1(Q) - y_1(i)}{s_{y_1}(Q)} \right) \left(\frac{\bar{y}_2(Q) - y_2(i)}{s_{y_2}(Q)} \right)$



Subgroup discovery with multiple targets

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Sample $S \subseteq P$

Target variable $y_1, y_2: P \rightarrow \mathbb{R}$

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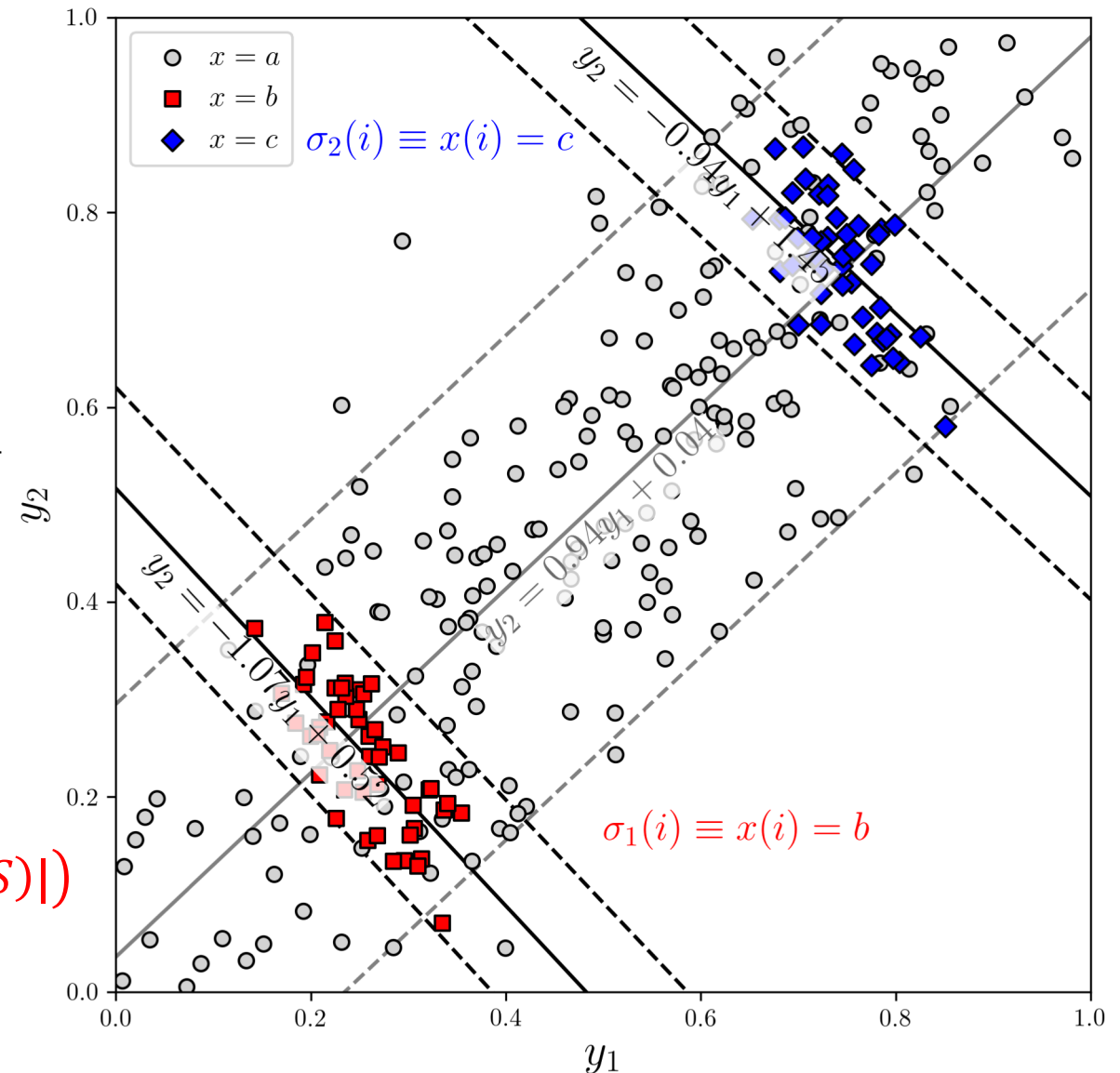
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Application 2: Au structure/property relationship

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Population

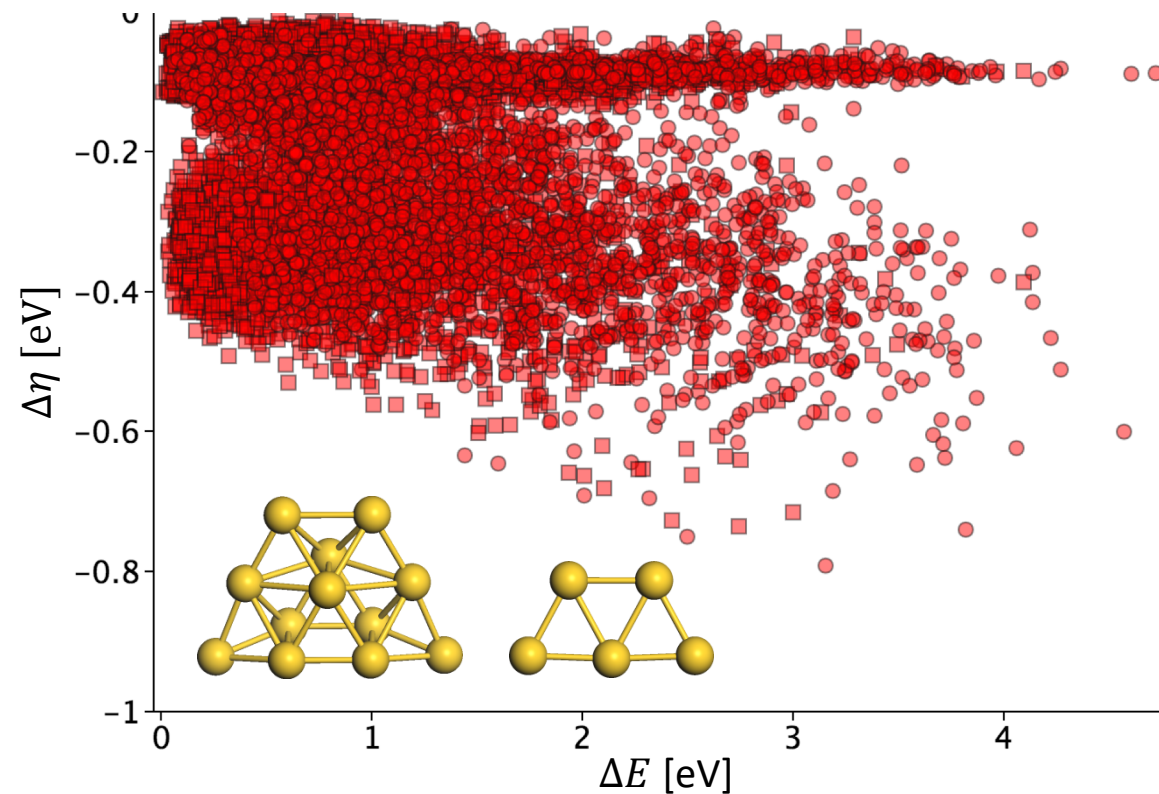
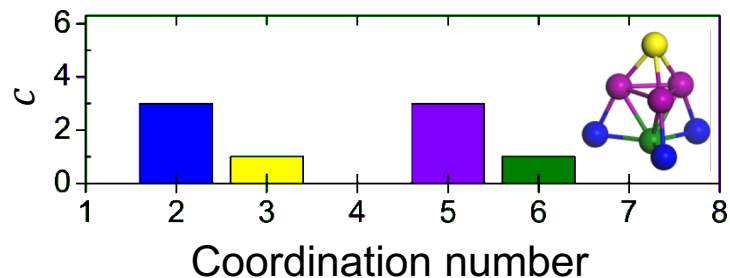
$$P = \{c: c \text{ conf. of Au5} - \text{Au14}\}$$

Targets

$$y_1 = \Delta E, y_2 = \Delta\eta \text{ chem. hardness}$$

Features

$$x \in \{a, c_1, c_2, c_3, c_4, c_5, c_6, r, \text{shape}, \text{Mo}_{\text{CO}}, \text{Me}_{\text{CO}}\}$$



Application 2: Au structure/property relationship

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Population

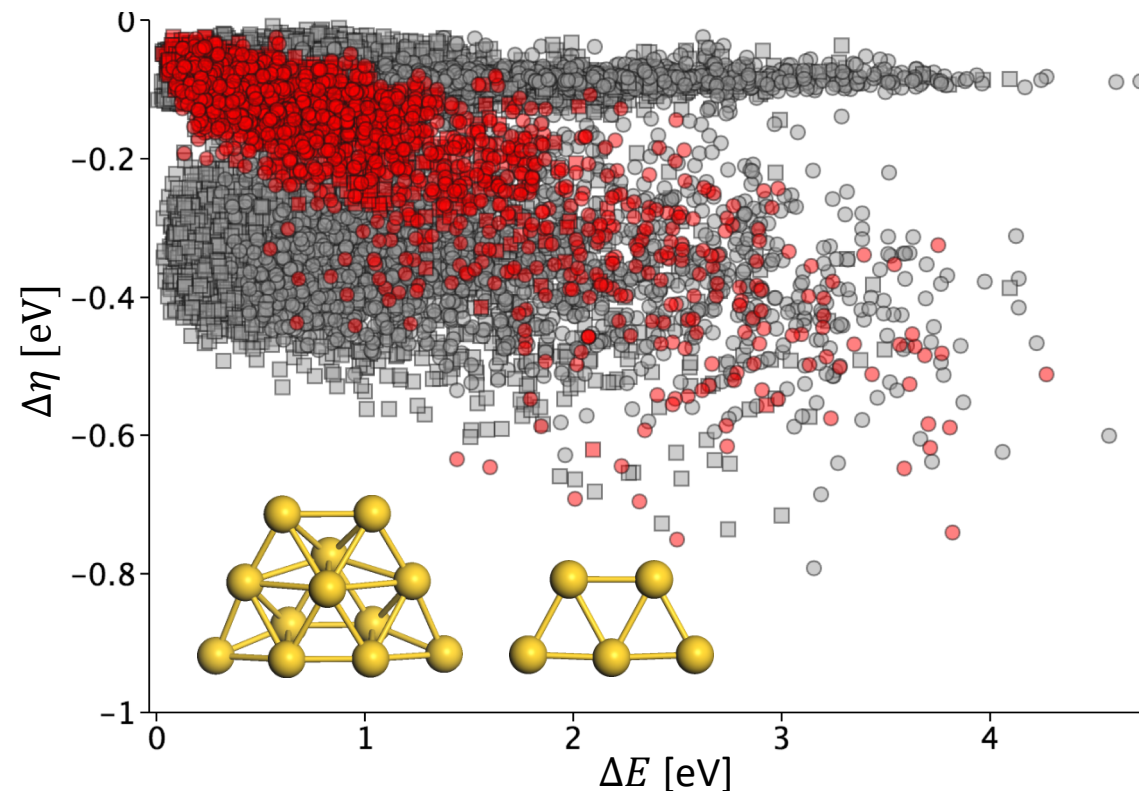
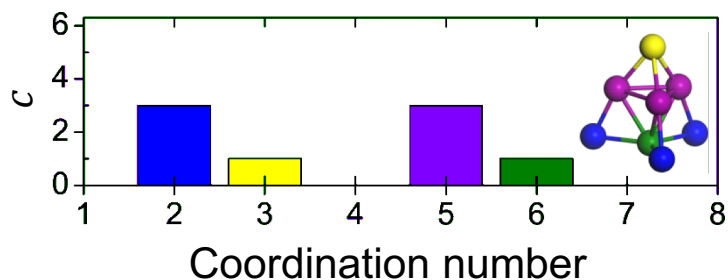
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Selector

$$\sigma(i) \equiv \text{even}(a(i)) \wedge (c_5(i) \leq 0.24) \wedge (\Delta E_{\text{vdw}}(i) \leq 0.18)$$

Parameters

$$\text{cov}(\sigma) = 0.2 \quad \text{eff}(\sigma) = 0.74 \quad [r(Q) = -0.81, r(S) = -0.27]$$

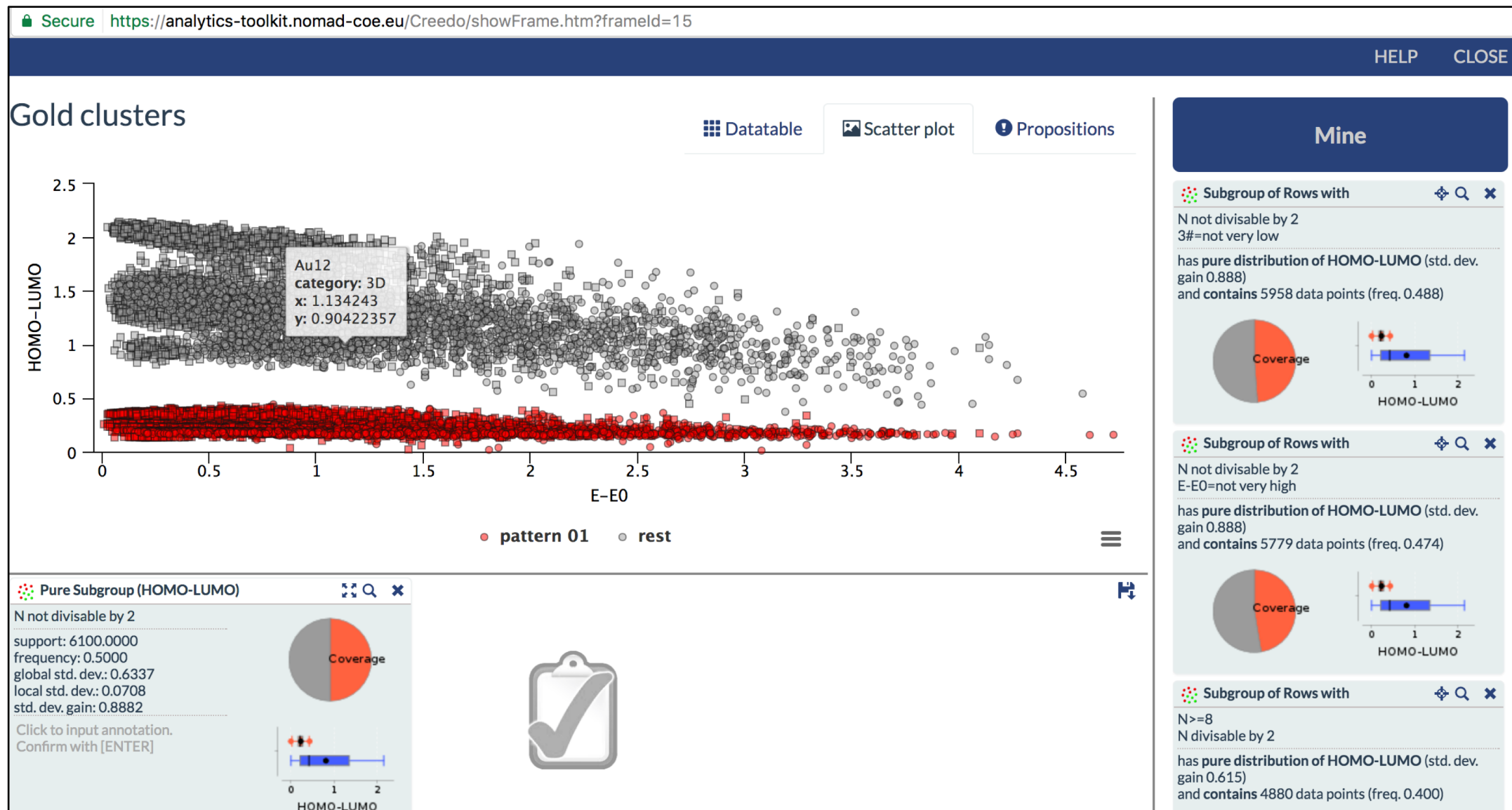
Topics

- Basic concepts: selectors, extensions, objective functions
- Application I: octet binary crystal structures
- A glimpse beyond: numeric and multiple targets
- Application II: Au structure/property relationship

References

- Boley, www.realkd.org: **The power of saying ‘I don’t know’**
- Atzmueller, WIREs Data Mining Knowl Discov, 2015: Subgroup discovery – advanced review
- Friedman and Fisher, Stat Comput, 1999: Bump hunting in high-dimensional data
- **Goldsmith et al., New J Phys, 2017: Uncovering structure-property relationships by subgroup discovery**
- Boley et al., Data Min Knowl Disc, 2017: Identifying consistent statements about numerical data with dispersion-corrected subgroup discovery

Try it out for yourself



Try it out for yourself

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Secure <https://analytics-toolkit.nomad-coe.eu/Creedo/showFrame.htm?frameId=15>

Crystal structure prediction

On-the-fly data analysis for the NOMAD Archive

Predicting energy differences between crystal structures

Tutorial on compressed sensing for materials property prediction

Discovering simple descriptors for crystal-structure classification

Evaluating the (dis)similarity of crystalline, disordered, and molecular compounds

Building structure maps for crystal-structure classification

0 1 2
HOMO-LUMO

gam0.019
and contains 4880 data points (freq. 0.400)