

Weight Learning from Cost Matrix in Weighted Least Squares Model Based on Genetic Algorithm

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October 15, 2018

Background



$$\mathbf{C} = \begin{bmatrix} 0 & \cdots & c_{1m} \\ \vdots & \vdots & \vdots \\ c_{m1} & \cdots & 0 \end{bmatrix}_{m \times m}$$

Motivation

- The existing cost-sensitive learning methods cannot guarantee the overall misclassification cost attains its minimum value.
 - The weights in Weighted Least Squares model are generally unknown and finding these weights is usually difficult.
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Contribution

- We propose a weight learning method for constructing a cost-sensitive classifier.
 - We complete the transition from the misclassification cost matrix (MCM) to the weights in the weighted least square model.
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Model Construction

- Objective Function:

$$\begin{aligned} \text{Min: } E = & w_1 \sum_{i=1}^{N_1} \left(\sum_{j=1}^{\tilde{N}} \beta_j g(\alpha_j \cdot \mathbf{x}_i + b_j) - \mathbf{t}_i \right)^2 \\ & + w_2 \sum_{i=N_1+1}^{N_2} \left(\sum_{j=1}^{\tilde{N}} \beta_j g(\alpha_j \cdot \mathbf{x}_i + b_j) - \mathbf{t}_i \right)^2 \\ & + \dots \\ & + w_m \sum_{i=N_{m-1}+1}^{N_m} \left(\sum_{j=1}^{\tilde{N}} \beta_j g(\alpha_j \cdot \mathbf{x}_i + b_j) - \mathbf{t}_i \right)^2 \end{aligned}$$

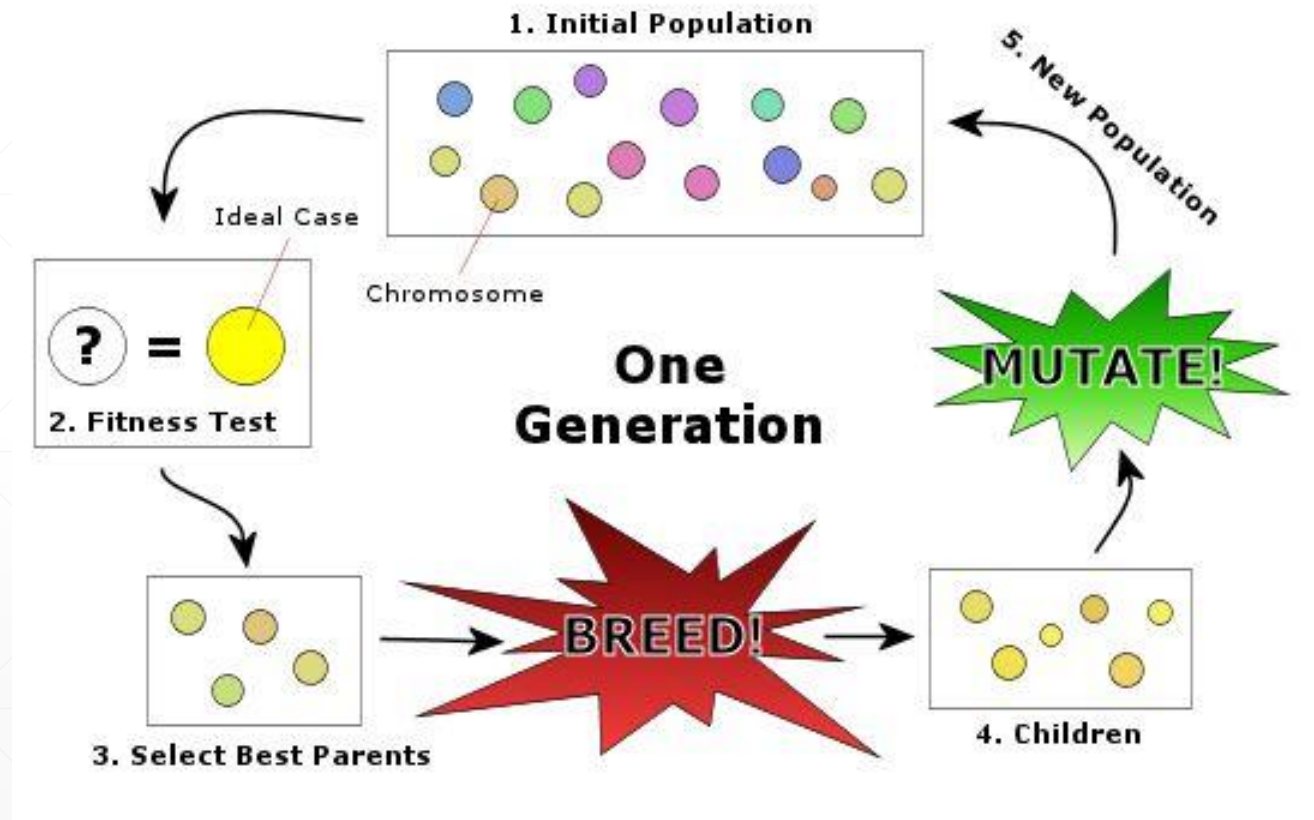
- Relational Expression between Output Weights and Weights of the Least Squares Model:

$$\hat{\boldsymbol{\beta}} = \begin{pmatrix} w_1 \mathbf{H}_1 \\ \vdots \\ w_m \mathbf{H}_m \end{pmatrix}^+ \begin{pmatrix} w_1 \mathbf{T}_1 \\ \vdots \\ w_m \mathbf{T}_m \end{pmatrix}$$

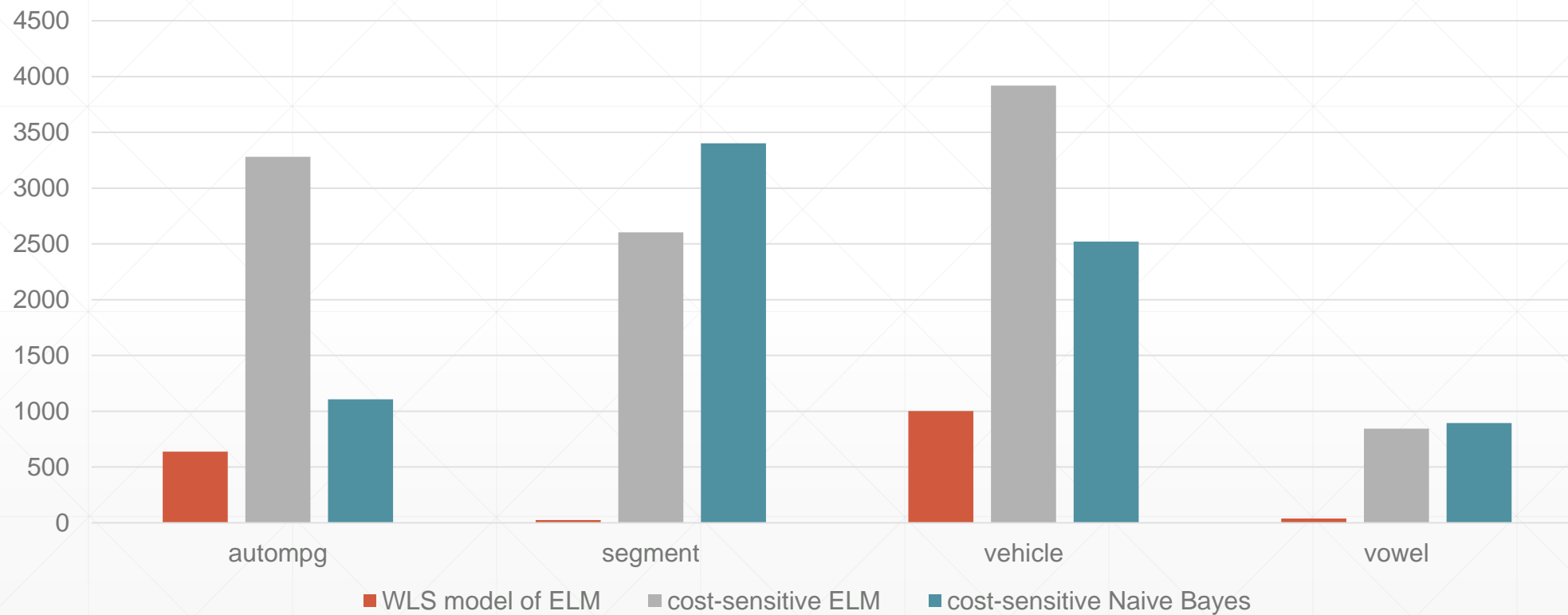
where $\mathbf{H} = \begin{bmatrix} g(\boldsymbol{\alpha}_1 \cdot \mathbf{x}_1 + b_1) & \cdots & g(\boldsymbol{\alpha}_{\tilde{N}} \cdot \mathbf{x}_1 + b_{\tilde{N}}) \\ \vdots & \vdots & \vdots \\ g(\boldsymbol{\alpha}_1 \cdot \mathbf{x}_N + b_1) & \cdots & g(\boldsymbol{\alpha}_{\tilde{N}} \cdot \mathbf{x}_N + b_{\tilde{N}}) \end{bmatrix}_{N \times \tilde{N}}$ is the

hidden layer output matrix, “+” is the Moore-Penrose generalized inverse operator.

Genetic Algorithm



Experimental Validation



Conclusion

- We propose a weight learning method for constructing a WLS model of ELM.
 - We complete the transition from the misclassification cost matrix (MCM) to the weights in the weighted least square model.
 - We conduct comparative experiments to verify the validity of the proposed method.
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