

New Crossover Operator for Evolutionary Rule Discovery in XCS

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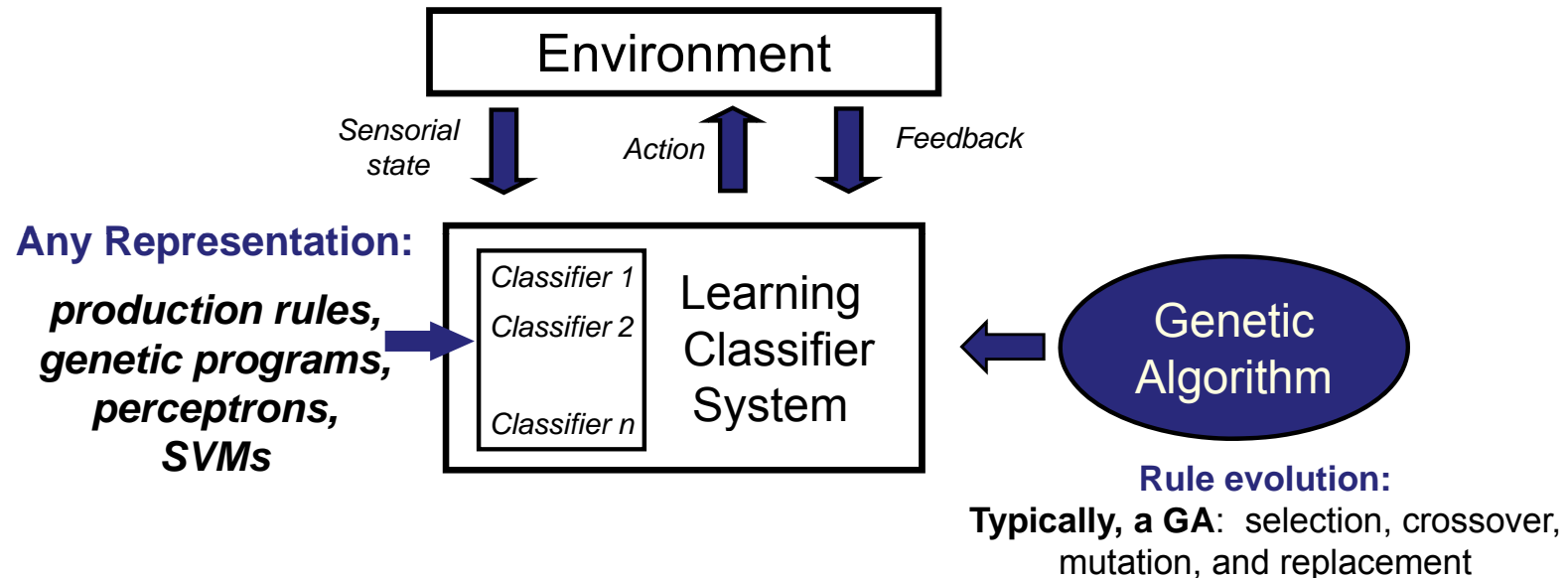
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Framework

- Michigan-style LCSs are mature learning techniques

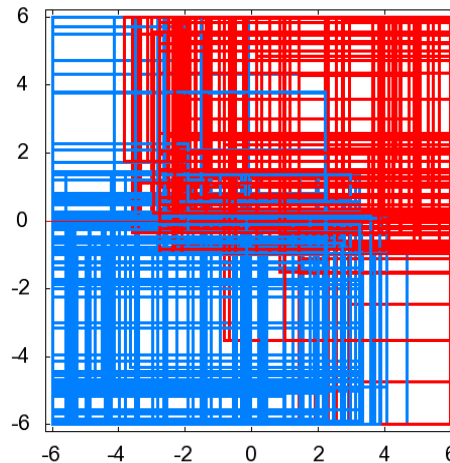
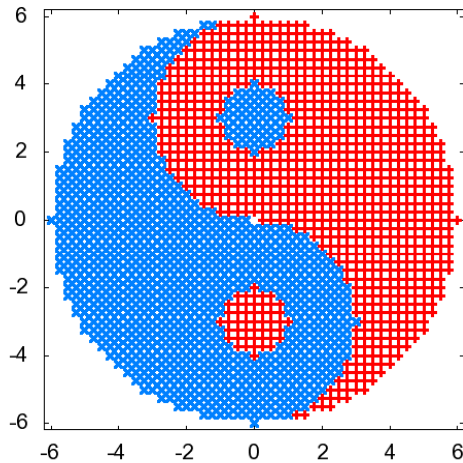


- XCS (Wilson, 95, 98)
 - By far, the most influential LCS

Motivation

□ Problems with continuous attributes

- Interval-based representation (Wilson, 2001)
- **IF** $v_1 \in [l_1, u_1]$ *and* $v_2 \in [l_2, u_2]$ *and* ... *and* $v_n \in [l_n, u_n]$ **THEN** class_i



- **2-point crossover**

➤ Too disruptive?

- **Mutation:** add a random uniform value

➤ Could we use more information?

□ Could we design better genetic operators?

- Not exactly clear the impact of crossover and mutation
- Systematic analysis
- Creative analysis: propose new operators

Purpose of the Work

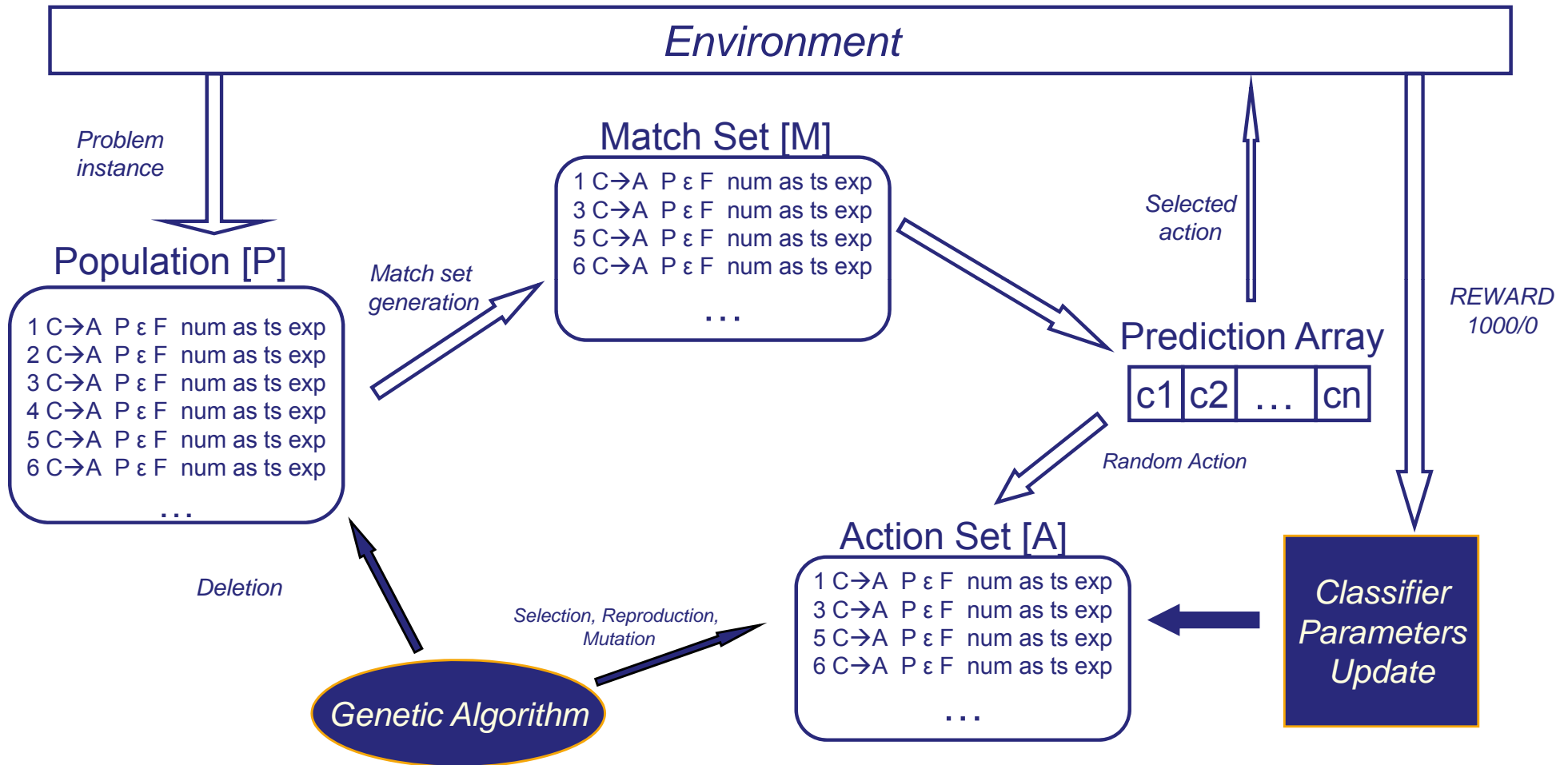
- **Previous work** (Morales et al., 2008a)
 - Design an XCS based on evolution strategies (ES)
 - Analyze the role of Gaussian mutation
 - **Conclusion:** Gaussian mutation (ES) enables XCS to approximate decision boundaries more accurately

- **The purpose of this work is to**
 - Design a new crossover operator, *BLX crossover*, following the ideas observed in the ES-based XCS
 - Compare GA-based XCS and ES-based XCS with the new crossover operator.

Outline

- 1. Description of XCS**
- 2. The New BLX Crossover Operator**
- 3. Experimental Methodology**
- 4. Results**
- 5. Conclusions and Further Work**

Description of XCS



Genetic Operators

□ Selection

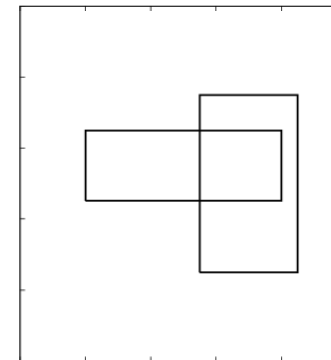
- Tournament selection
- Proportionate selection
- Truncation selection (ES)

□ Crossover:

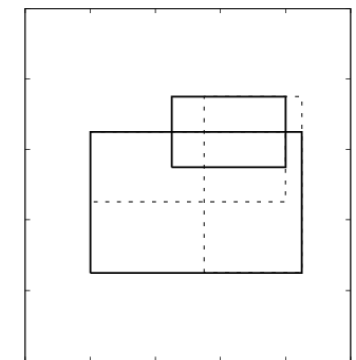
- Two-point crossover

$$\begin{array}{l} \langle [0.20, 0.80], [0.45, 0.55] \rangle \\ \langle [0.60, 0.85], [0.25, 0.75] \rangle \end{array} \Rightarrow \begin{array}{l} \langle [0.20, 0.85], [0.25, 0.55] \rangle \\ \langle [0.60, 0.80], [0.45, 0.75] \rangle \end{array}$$

Parents

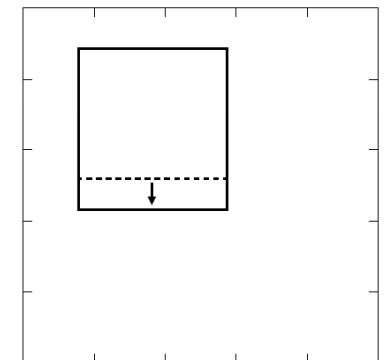


Offspring



□ Mutation:

- GA-based XCS: Add a uniform random value
- ES-based XCS: Add a Gaussian random value



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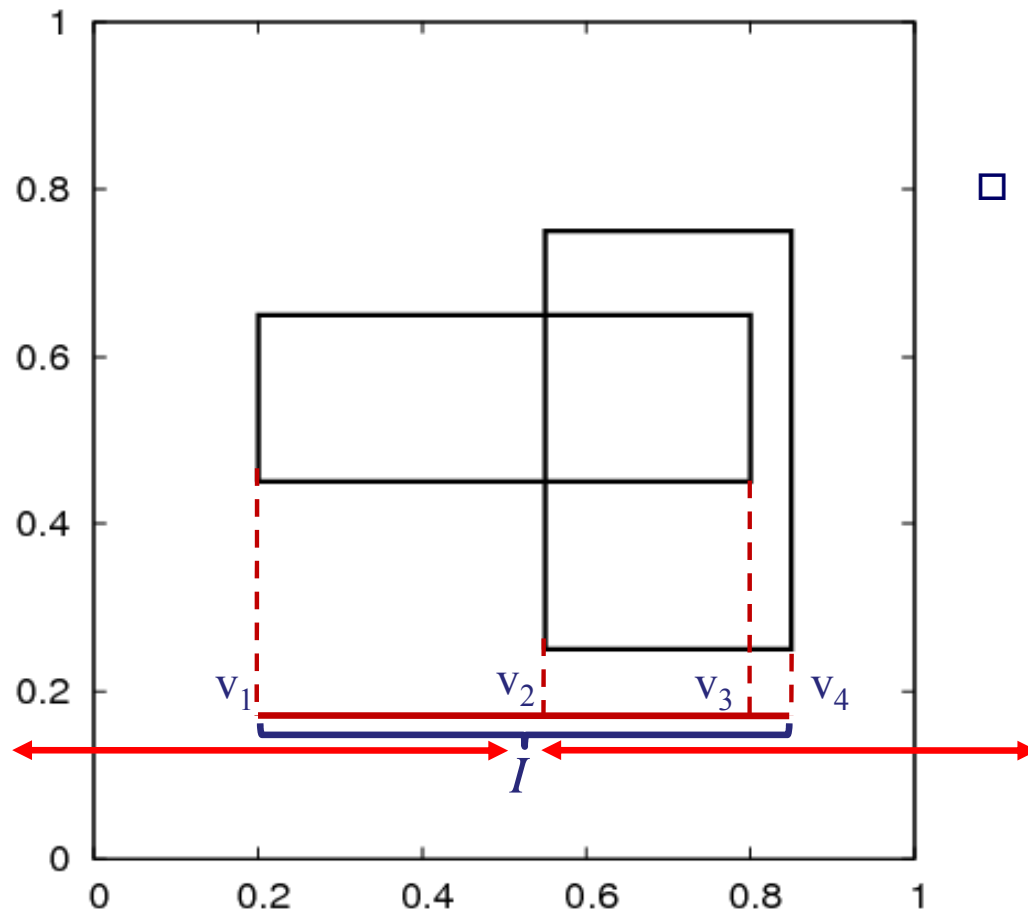
BLX-Crossover Operator

- **Problem with two-point crossover:**
 - Offspring maintain parent's boundaries
 - Mutation is responsible for modifying these boundaries

- **Advantage of evolution strategies:**
 - Mutation is more focused
 - Mutation is applied more often

- **Aim of BLX-Crossover**
 - Combine the *innovation* power of 2-point crossover with also *local search* by permitting moving the boundaries

BLX-Crossover Operator



- For each offspring $o1$ and $o2$, the intervals of each attribute $[l_i, u_i]$ are generated as:

- $\alpha = \text{rand}(0,0.5)$
- $l_{o1} = v_1 + \alpha I \text{ rand}\{-1, 1\}$
- $l_{o2} = v_1 + (1-\alpha)I \text{ rand}\{-1, 1\}$
- $u_{o1} = v_4 + \alpha I \text{ rand}\{-1, 1\}$
- $u_{o2} = v_4 + (1-\alpha)I \text{ rand}\{-1, 1\}$

- **Key aspects:**

- Boundaries *move according to I* , the distance between parents bounds
- When one offspring is practically equal to the parents ($\alpha \approx 0$), the other is very different from them ($\alpha \approx 0.5$)

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Experimental Methodology

- Experiments on 12 real-world data sets from the UCI repository
 - 10-fold cross validation

Id.	dataset	#Inst	#Fea	#Re	#In	#No	#Cl	%MisInst
<i>bal</i>	Balance	625	4	4	0	0	3	0
<i>bpa</i>	Bupa	345	6	6	0	0	2	0
<i>gls</i>	Glass	214	9	9	0	0	6	0
<i>h-s</i>	Heart-s	270	13	13	0	0	2	0
<i>irs</i>	Iris	150	4	4	0	0	3	0
<i>pim</i>	Pima	768	8	8	0	0	2	0
<i>tao</i>	Tao	1888	2	2	0	0	2	0
<i>thy</i>	Thyroid	215	5	5	0	0	3	0
<i>veh</i>	Vehicle	846	18	18	0	0	4	0
<i>wbcd</i>	Wisc. breast-cancer	699	9	0	9	0	2	2.3
<i>wdbc</i>	Wisc. diagnose breast-cancer	569	30	30	0	0	2	0
<i>wne</i>	Wine	178	13	13	0	0	3	0

Experimental Methodology

- **XCS configured as:**
 - Iter=100000, $N = 6400$, $\theta_{GA} = 50$, $P_{cross} = 0.8$, $P_{mut} = 0.04$,
 $r_0 = 0.6$, $m_0 = 0.1$
- **Comparison of:**
 - XCS-GA with proportionate and tournament selection
 - XCS-ES with proportionate, tournament and truncation selection
 - XCS-GA with two point-crossover
- **Why this comparison?**
 - Does BLX approach XCS-GA to XCS-ES?
 - Is BLX better than 2-point crossover in XCS?

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Results

- Test Accuracy**
 - XCS-GA with tournament
 - XCS-ES with proportionate
 - XCS-GA with proportionate
 - XCS-ES with tournament
 - XCS-ES with truncation
 - Original XCS

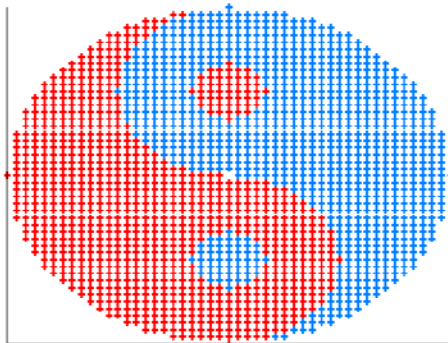
Data set	XCS-GA ps	XCS-ES ps	XCS-GA ts	XCS-ES ts	XCS-ES tr	XCS-GA tp
<i>bal</i>	86.29 (1)	85.39 (5)	85.44 (4)	85.55 (2.5)	85.55 (2.5)	83.20 (6)
<i>bpa</i>	69.18 (1)	65.89 (6)	68.11 (3.5)	68.11 (3.5)	66.76 (5)	68.21 (2)
<i>gls</i>	69.94 (5)	70.25 (4)	71.65 (2)	71.49 (3)	69.16 (6)	72.12 (1)
<i>h-s</i>	39.51 (4)	38.64 (6)	38.76 (5)	39.88 (3)	43.21 (2)	46.91 (1)
<i>irs</i>	95.11 (4)	95.11 (4)	95.33 (1.5)	95.11 (4)	94.89 (6)	95.33 (1.5)
<i>pim</i>	75.22 (1)	74.43 (4)	74.65 (2.5)	73.83 (5)	74.65 (2.5)	72.53 (6)
<i>tao</i>	96.80 (1)	96.77 (2)	96.68 (3)	96.61 (4)	96.57 (5)	91.22 (6)
<i>thy</i>	97.36 (1)	96.90 (3)	97.05 (2)	96.43 (5)	96.74 (4)	95.81 (6)
<i>veh</i>	69.34 (4)	69.23 (6)	69.74 (3)	70.21 (2)	69.31 (5)	71.79 (1)
<i>wbcd</i>	95.66 (5)	96.28 (3)	96.33 (1.5)	95.99 (4)	96.33 (1.5)	94.85 (6)
<i>wdbc</i>	91.04 (5)	91.33 (3)	91.62 (2)	91.97 (1)	90.63 (6)	91.09 (4)
<i>wne</i>	95.88 (4)	96.07 (3)	96.25 (2)	96.82 (1)	95.13 (6)	95.50 (5)
Rank	3.00	4.08	2.67	3.17	4.29	3.79
Pos	2	5	1	3	6	4

Results

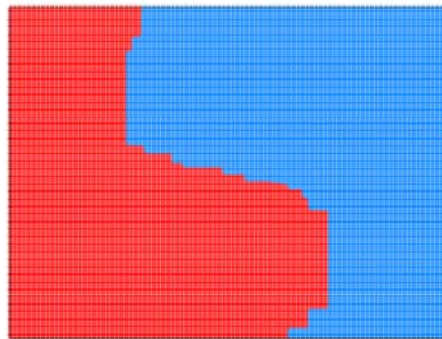
- **Two key observations:**

- BLX crossover enables XCS to fit complex boundaries more accurately
- BLX may prevent XCS from over-fitting in complex domains

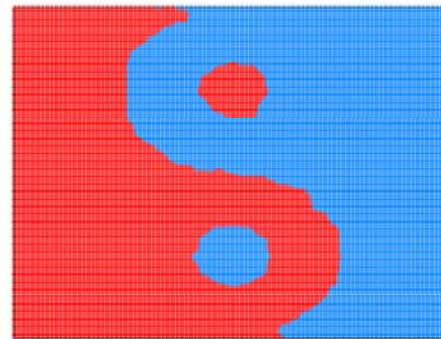
- **Fit complex boundaries: the TAO problem**



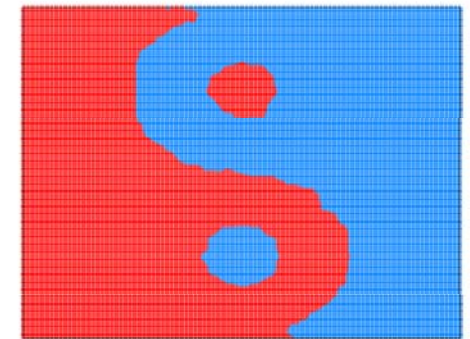
Domain



Original XCS



XCS-GA BLX

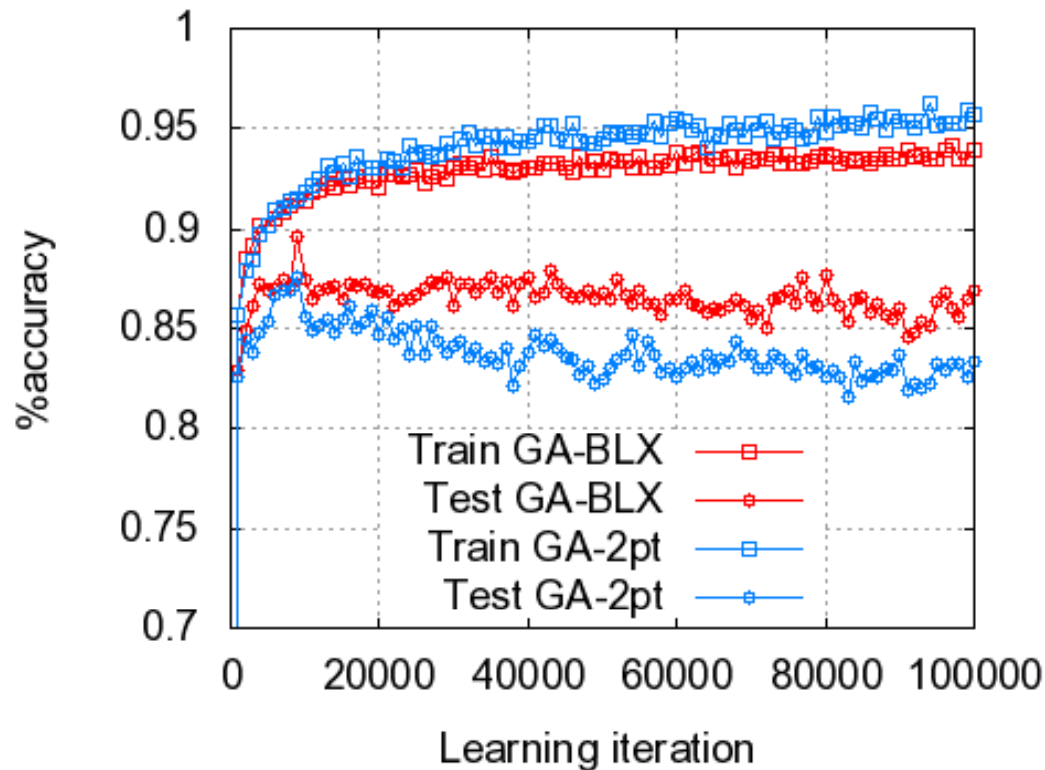


XCS-ES BLX

- The freedom provided by BLX enables the system to create new classifiers that fit the curved boundaries

Results

□ Prevention from overfitting: The BAL problem



- To increase the training accuracy, two-point crossover starts to over-fit the training examples in the BAL problem

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Conclusions

- **On average, BLX yields more accurate models than 2-point crossover**
- **BLX crossover enhances XCS-GA**
 - Morales et al. (2008) showed that XCS-ES resulted in more accurate models than XCS-GA.
 - BLX approaches XCS-GA results to XCS-ES results.
- **Two important observations:**
 - BLX helps fit curved boundaries.
 - BLX may prevent over-fitting.
- **The overall work clearly shows the importance of further research on GA operators.**

Further Work

- **Well, BLX is good! But, always?**
 - On average, yes!
 - Specific problems may not benefit from BLX

- **May evolution tell me when to use one operator or another?**
 - Existing studies on self-adaptation mutation for ternary rules
 - Search for evolution signals
 - Combine different operators
 - Let classifiers decide which operator to use
 - Characterize learning domains

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