

Invariant Generation.

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Summary

- ① Heuristics list for Cost Function.
- ② An Example of a Cost Function adopting continuous learning from Z3.
- ③ Heuristics for Mutating Invariant.

Heuristics for Cost Function.

Used heuristics by earlier work:

- "Smaller values" are better features and better thresholds.
- Choose features which maximize distance from TS and classifier.[???
- Given features, choose threshold which give more skewed partitions.

New Ideas:

- $I \wedge \neg B$ must be "close" to Q .
- Area of set not satisfying constraint - not sure how to compute this.
- Iterative converging methods.
 - Are there iterative solution finders to CHC equations - do they define an error function?
 - Get an ϵ -approximation to P , B , Q , and call a solution for this an ϵ -invariant.
- Learning from $Z3^1$.
 - Extract a continuous function from our learner ($Z3$) rather than a binary function.
 - Get k counterexamples. Define a continuous function over these k examples (area/centroid/ distance).

¹Need to figure out a way to mutate invariants

Learning from Z3 using s counterexamples: One Example.

Let the equations be (universally quantified over s,t):

$$P(s) \rightarrow I(s) \quad (1)$$

$$I(s) \wedge B(s) \wedge T(s, t) \rightarrow I(t) \quad (2)$$

$$I(s) \wedge \neg B(s) \rightarrow Q(s) \quad (3)$$

Let $J(I)$ be the cost function for a guess invariant. Then:

$$J(I) = K_1 J_1(I) + K_2 J_2(I) + K_3 J_3(I)$$

where $J_i(I)$ is the cost for violating equation (i). Get atmost s (try $s = 10$) counterexamples for each case.² Let z_i denote the ith counterexample for case 1 and case 3 and (z_i^1, z_i^2) for case 2.

$$J_1(I) = \max_{1 \leq i \leq s} d(z_i, I)$$

$$J_2(I) = \text{Penalty}(TS, I) \quad (?)$$

$$J_3(I) = \max_{1 \leq i \leq s} d(z_i, Q)$$

For $J_2(I)$ the 'variance' of the ICE examples seems like a good fit³.

²There will either be 0,1 or ∞ counterexamples for linear boundaries for (1) and (3).

³Computing $J_1(I)$ and $J_3(I)$ will require a method for non linear boundaries.

Mutating Invariants: Heuristics.

- In each iteration, randomly change some parameter of the invariant. If it decreases the cost, go to this invariant, else do not.
- If we fix the number of hyperplanes (to say k), then invariants close to $I(\theta_{11}, \theta_{12} \dots \theta_{1n}, \dots \theta_{k1}, \theta_{k2} \dots \theta_{kn}, c_1, \dots c_k)$ is $I(\theta_{11} \pm \epsilon, \theta_{12} \pm \epsilon \dots \theta_{1n} \pm \epsilon, \dots \theta_{k1} \pm \epsilon, \theta_{k2} \pm \epsilon \dots \theta_{kn} \pm \epsilon, c_1 \pm \epsilon, \dots c_k \pm \epsilon)$. But how to do this when k is variable? Other thought - should we do translations first, then rotations?