# Strategy Representation by Decision Trees with Linear Classifiers

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# Strategy

#### System:

S: state space

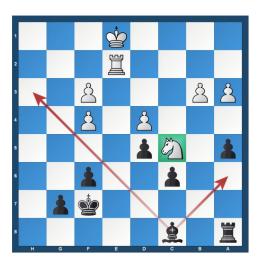
A: action space

Positional Strategy: resolve choices

 $\sigma$ : **S**  $\square$  **A** or

 $\sigma$ : S x A  $\square$  {tt, ff}

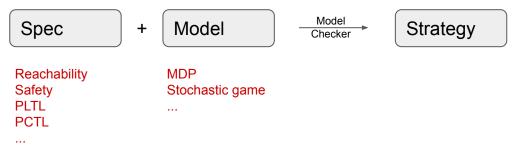
Focus: Synthesis of controllers



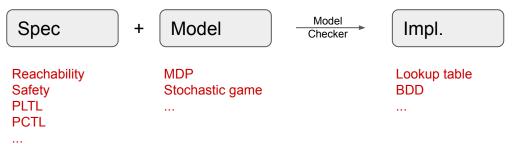
Model Checking Approach

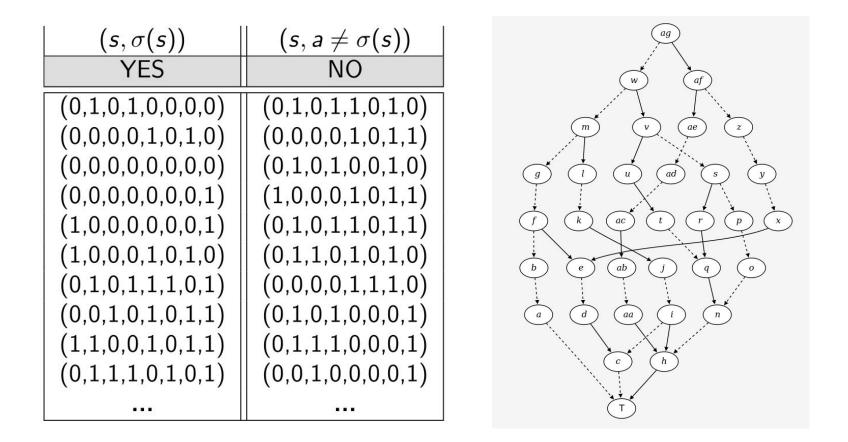


Model Checking Approach



Model Checking Approach





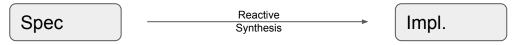
Lookup table

Binary Decision Diagram (BDD)

Model Checking Approach



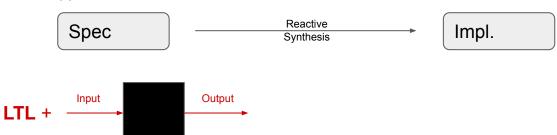
Reactive Synthesis Approach



Model Checking Approach



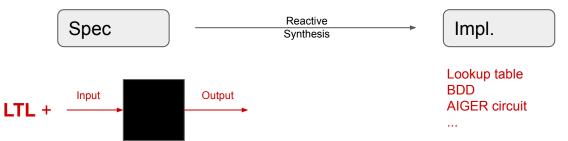
#### Reactive Synthesis Approach



Model Checking Approach



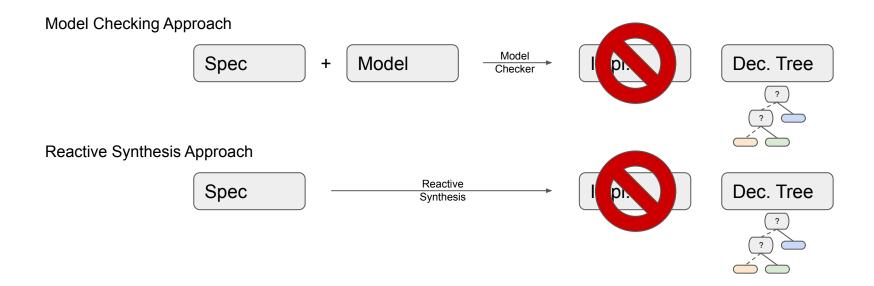
#### Reactive Synthesis Approach



#### Pitfalls of current implementations

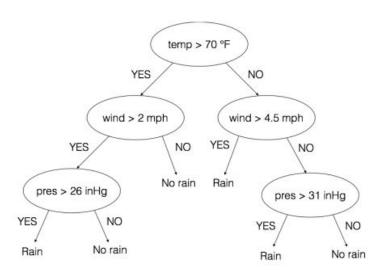
Large
Uninterpretable

Difficult to implement
Difficult to debug
Mistrust in methods



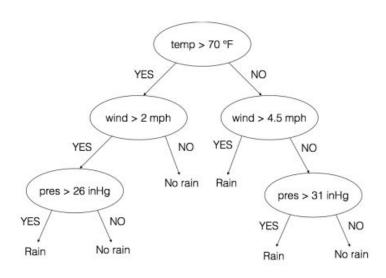
#### **Decision trees**

(from Machine Learning)



#### **Decision trees**

(from Machine Learning)



temp = 100 F wind = 5 mph pres = 32 inHg

#### Decision trees in formal methods

Garg et. al. POPL'16

- Program invariants from implication counterexamples

Neider et. al. (ACM TOCL, May 2018)

- Piece-Wise Functions against logical specifications

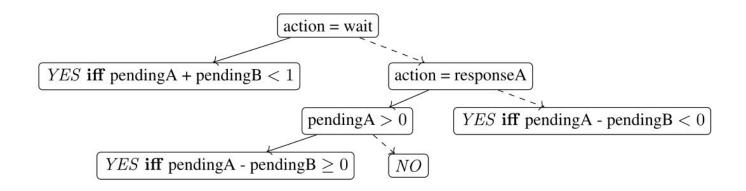
Brazdil et. al. (CAV 2015)

- Counter-example representation using dec. trees

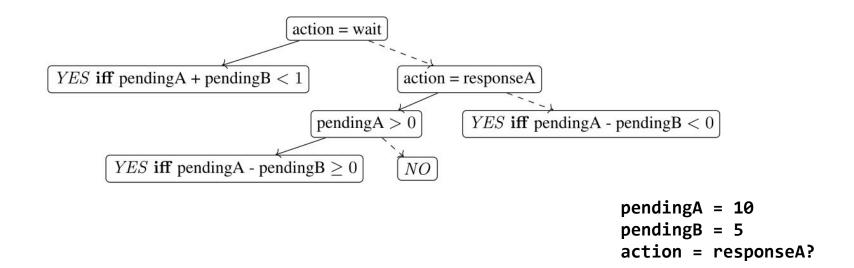
Brazdil et. al. (TACAS 2018)

- Dec. trees vs BDDs in reactive synthesis

#### Contribution: DT with linear classifiers

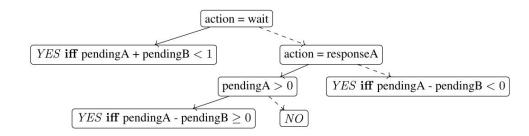


#### Contribution: DT with linear classifiers

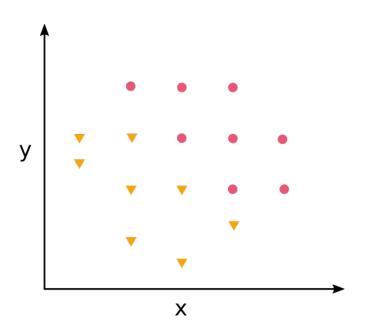


#### No such algorithm from ML?

- Want exact data representation
- No algorithm with LC + exact data reprst.
- Focus in ML on generalizing to data
- Not ideal for strategies

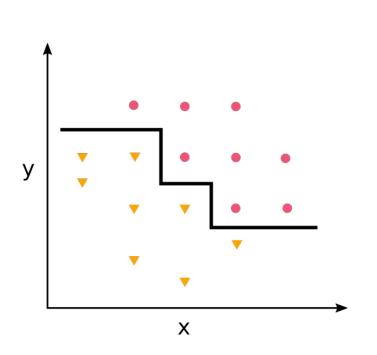


# Why linear classifiers?

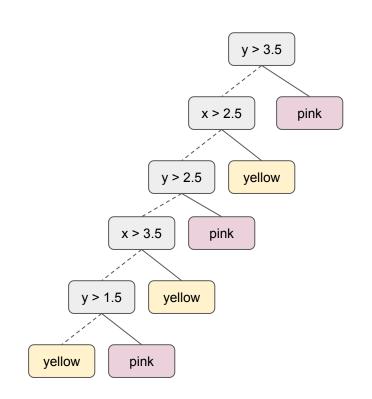


- Each point state
- Two actions: circle/triangle

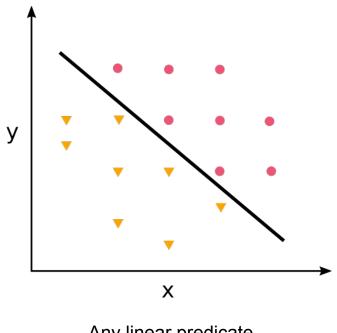
## Why linear classifiers?



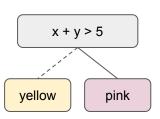
Standard DT: Only axis-parallel predicates

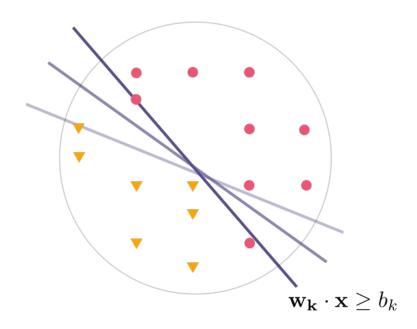


# Why linear classifiers?

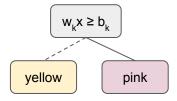


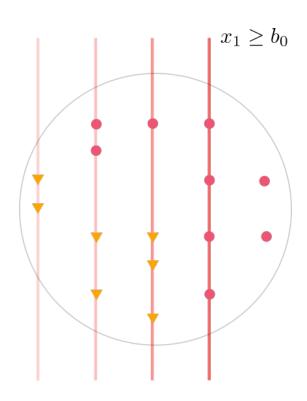
Any linear predicate





Step 1: Check if dataset is linearly separable (If yes, we are done)

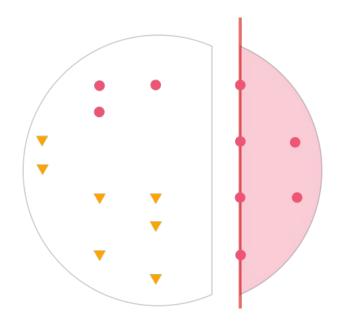




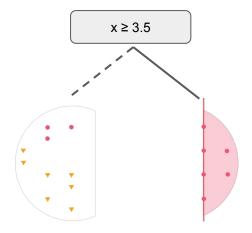
**Step 2: Try out each axis-parallel split...** 

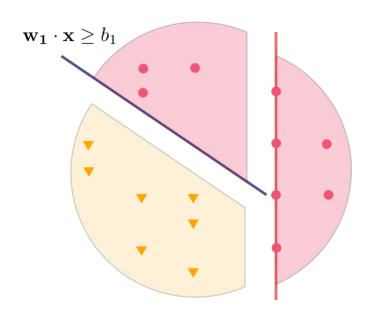
$$x_1 \ge 1$$
  $x_1 \ge 2$   $x_1 \ge 3$  .....  $x_1 \ge b_0$  .....

$$x_2 \ge 1$$
  $x_2 \ge 2$   $x_2 \ge 3$  ....

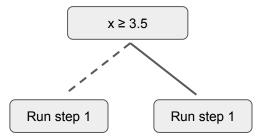


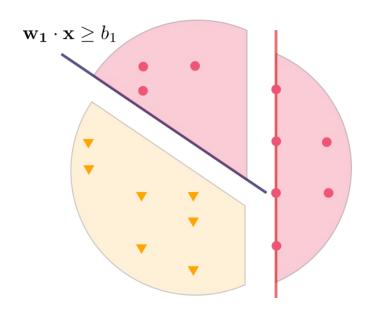
Step 2: Try out each axis-parallel split...
until the best split is found!



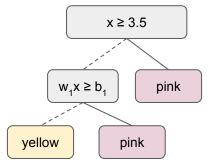


**Step 3: for each child, continue from step 1.** 



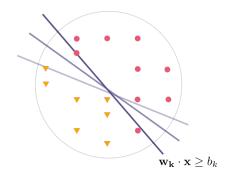


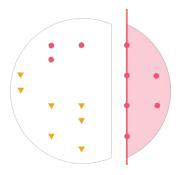
**Step 3: for each child, continue from step 1.** 

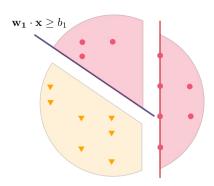


#### Summary of algorithm

- 1. Check if linearly separable (if yes, we are done)
- 2. Find find best axis-parallel split
- 3. For each children, go to step 1



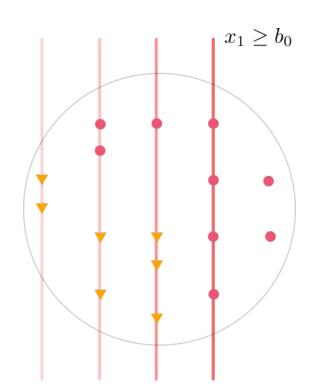




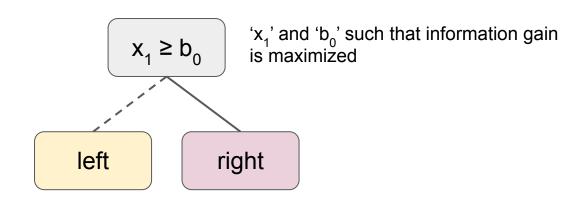
# Finding best split

#### Common measures used in ML

- Information gain
- Gini index



## Finding best split: Information Gain



Information gain = entropy(parent) - ( entropy(left) + entropy(right) )

# Finding best split: something better?

# Area under receiver operator characteristics (AUROC)

- Devised during WWII
- Now used in machine learning

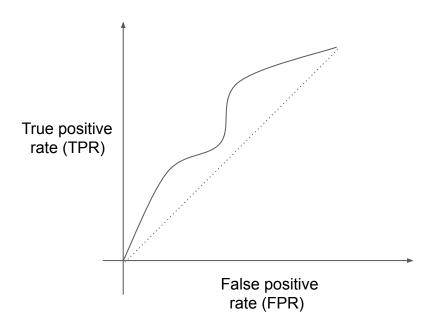
#### Finding best split: AUROC

#### **Area under ROC**

```
Given: dataset and family of classifiers wx \ge b
```

```
for each b:
```

```
TPR = % actual +ve identified as +ve
FPR = % actual -ve identified as -ve
plot (TPR, FPR)
```



#### Finding best split: AUROC

#### Area under ROC

Given: dataset and family of classifiers wx ≥ b

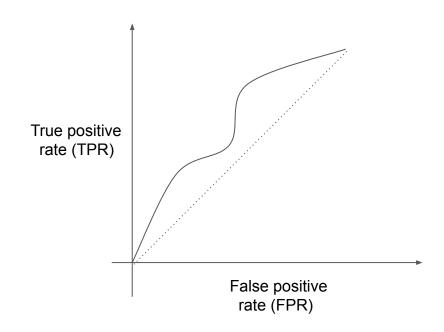
#### Intuition

Point above 45 deg line - good Point below 45 deg line - bad

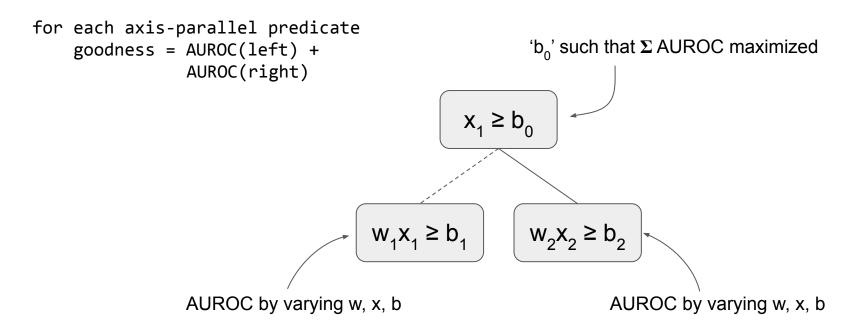
Area under ROC curve = 1 (perfect)

Area under ROC curve = 0.5 (average)

Area under ROC curve = 0 (predicts exact opposite!)



## Finding best split: AUROC



#### Experiments

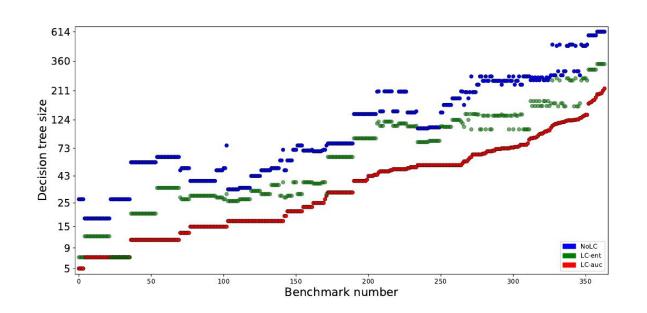
#### **Reactive synthesis**

- 1. Scheduling of washing cycles (SYNTCOMP 2015)
- 2. LTL synthesis
- 3. LTL synthesis with randomized environment

#### Model checking

4. PCTL on MDPs

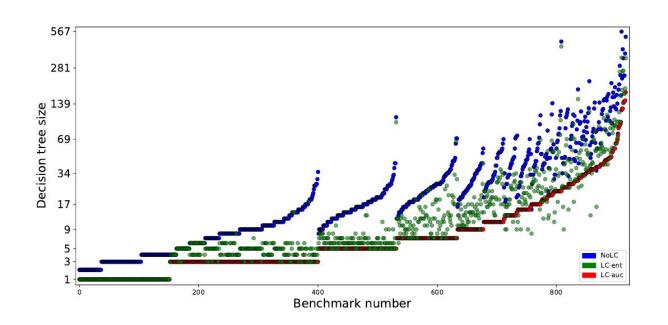
## Experiments: Washing Cycle (SYNTCOMP)



#### Mean

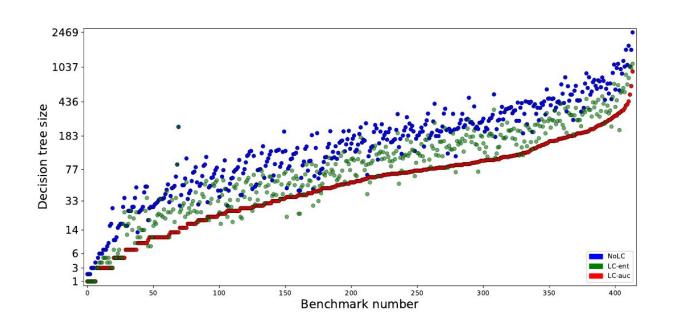
LC-entropy is 59% of NoLC LC-auc is 33% of NoLC

## Experiments: LTL synthesis



Mean LC-ent is 51% of NoLC LC-auc is 36% of NoLC

## Experiments: LTL synthesis with randomized env.



#### Mean LC-ent is 58% of NoLC LC-auc is 38% of NoLC

## **Experiments: MDP**

Model	# Train	# Dim	NoLC	LC+Entropy	LC+AUROC
coin-5	451204	13	2572	1626	566
csma-2-4	79580	13	54	41	59
firewire-5	90389	12	102	85	72
leader-4	38016	22	152	92	45
mer-30	1408932	19	1373	1332	126
wlan-2	275140	14	288	206	353
zeroconf	268326	24	413	330	376

#### Summary

- Representing strategies using DT with linear classifiers
- New splitting measure (AUROC)

#### **Future work**

- Representing AIGER circuits as DT
- Linear predicates in all nodes?
- More complex predicates?