## Making Woody Parallel

## Applying Parallel Computing to decision trees

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■ What is Woody?

■ Decision trees and Machine Learning



## Working with Woody Decision tree evaluation

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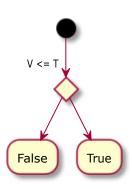
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A decision tree consists of nodes. Non-leaf nodes consist of a conditional check, where:

- V: Feature value checked for this node
- T: Threshold value of the node that is checked against





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- Python library run against a dataset
- Split data into training and test set
- 3 Fit on training set to create a forest of trees
- Run predictions with test set on this forest



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## Run as a library or run stand-alone?



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In order to pass the tree and test data from Woody to Futhark, we encoded each as a series of flat arrays.

- treeLeftid
- treeRightid
- treeFeature
- treeThres\_or\_leaf
- Xtest
- nXtest
- dXt.est.



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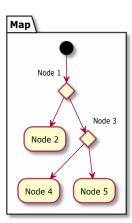
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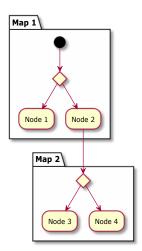
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 Layered provides a structured way of iterating over layers

- Many passes end early
- These are still considered by the maps



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```
let node_array = replicate n_preds 0
let is_not_leaf = (\ (node_id, _) -> treeLeftid[node_id] != 0)
let is_leaf = (\ (node_id, _) -> treeLeftid[node_id] == 0)
let next_node = (\ (node_id. data_row_start) ->
                  ((if (is_not_leaf (node_id, data_row_start)) then (if Xtest[
                         data_row_start + treeFeature[node_id]] <= treeThres_or_leaf[
                         node_id | then treeLeftid[node_id] else treeRightid[node_id])
                         else node_id), data_row_start))
let nodes = zip node_array data_row_starts
let leaves = []
let (_. leaves) = loop (nodes, leaves) for row in iota(depth) do
                  let new_nodes = (unsafe map next_node nodes)
                  in (unsafe filter is_not_leaf new_nodes, leaves ++ (unsafe filter
                         is_leaf new_nodes))
let result = map (\ (a, _) \rightarrow a) (radix_sort_by_key (\ (_, a) \rightarrow a) i32.num_bits i32.
      get_bit leaves)
in result
```



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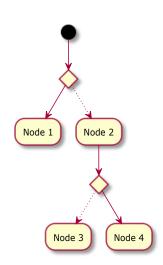
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```
let next node
  (row: [1f64)
  ((left, right, feature, thres) : (i32, i32, i32, f64)) : i32 =
  if row[feature] <= thres then left else right
let make_next_tree
   (tree: [](i32, i32, i32, f64))
   (row : []f64) : []i32 =
   map (next_node row) tree
let traverse
   (next_nodes: []i32): i32 =
   let (last, current) = (0, next_nodes[0])
   let (result, _) = loop (last, current) while current != 0 do (current, next_nodes
         [current])
   in result
let nodes = zip4 treeLeftid treeRightid treeFeature treeThres_or_leaf
let rows = unflatten nXtest dXtest Xtest
let next_nodes = unsafe map (make_next_tree nodes) rows
in unsafe map traverse next_nodes
```



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```
let repeated_criteria = flatten (replicate nXtest treeFeature)
```

- let flcr = map2 (+) repeated\_offsets repeated\_criteria

- let left\_or\_right = (\ b | r -> if b then | else r)
  - let repeatedLeft = flatten (replicate nXtest treeLeftid)
  - let repeatedRight = flatten (replicate nXtest treeRightid)
- let directions = map3 left\_or\_right threshold\_result repeatedLeft repeatedRight

in unsafe map traverse (unflatten nXtest treelength directions)



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## Comparing with Woody

- Only the covtype dataset.
- Troubles with swig: Measurements may be impacted

## Command to measure futhark code:

```
cat tmp_tree | ./futharkProgram -t measurements/futharkProgramTimes -r 10 > measurements/futharkProgram.txt
```

## Measuring Woody:

```
times = []
for i in range(n):
    cpu_pred_start_time = time.time()
    cpu_pred = super(WoodClassifier, model).predict_single_tree(Xtest)
    cpu_pred_stop_time = time.time()
    micro = int((cpu_pred_stop_time - cpu_pred_start_time) * 1000000)
    times.append(micro)
return times
```



# Experiment Results: Varying train data size

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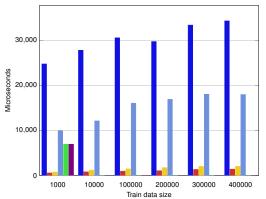
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## Microseconds used predicting a single tree as average of 10 runs







## Experiment Results: Varying number of predictions

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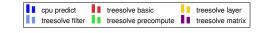
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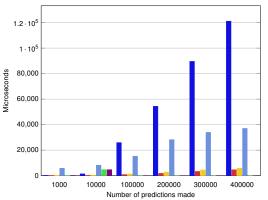
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## Microseconds used predicting a single tree as average of 10 runs







## Experiment Evaluation

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- Basic and layer are fast.
- GPU code scales better with bigger tests



# Experiment Future Research

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- Matrix and Precompute might be promising
- GPU code scales better with bigger tests



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We have proposed a number of approaches to parallelising the evaluation of decision trees using Futhark. Our findings show that as a whole this parallelisation is promising for the performance of decision tree evaluation on large datasets.



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