

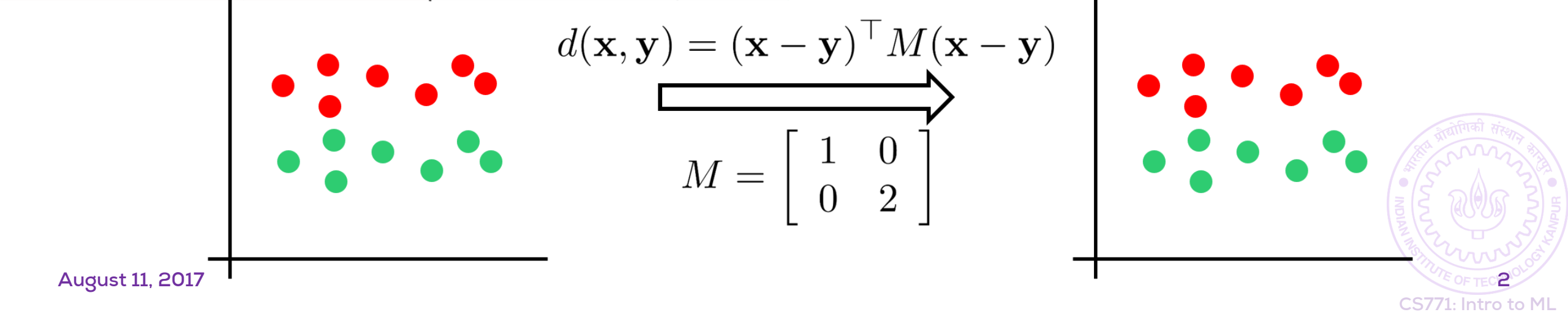
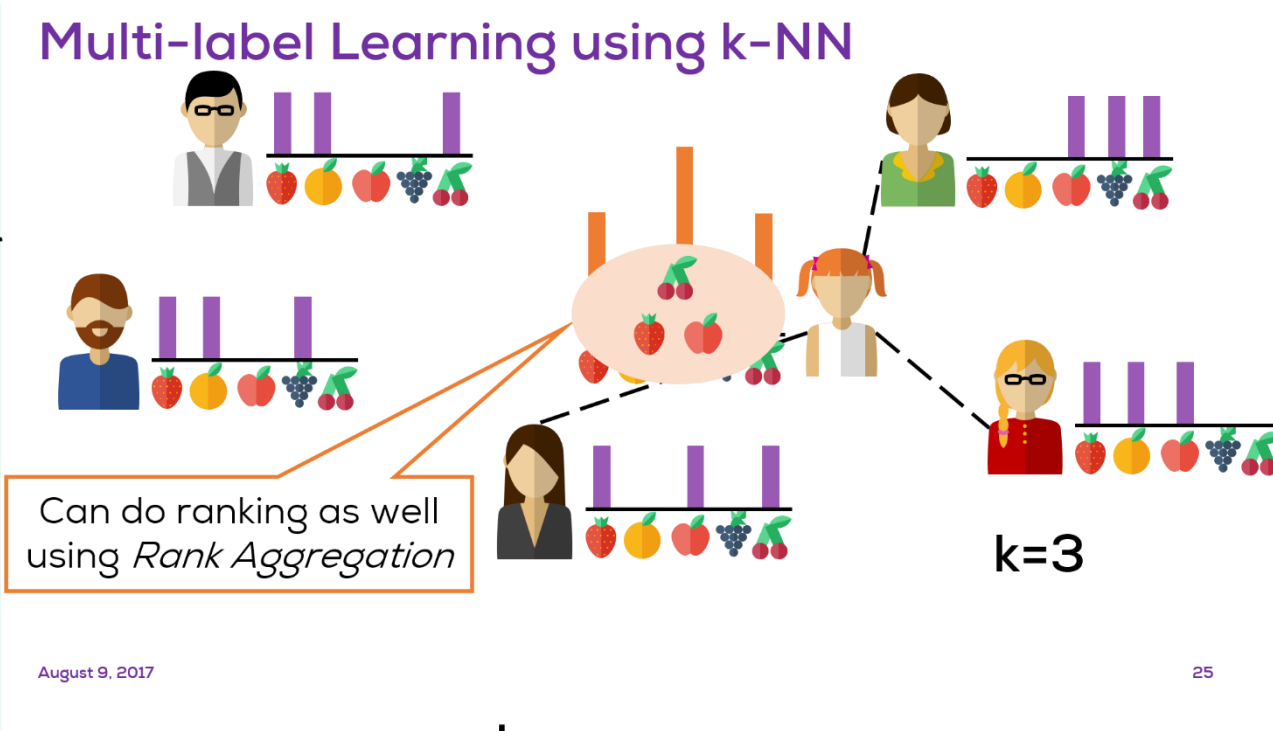
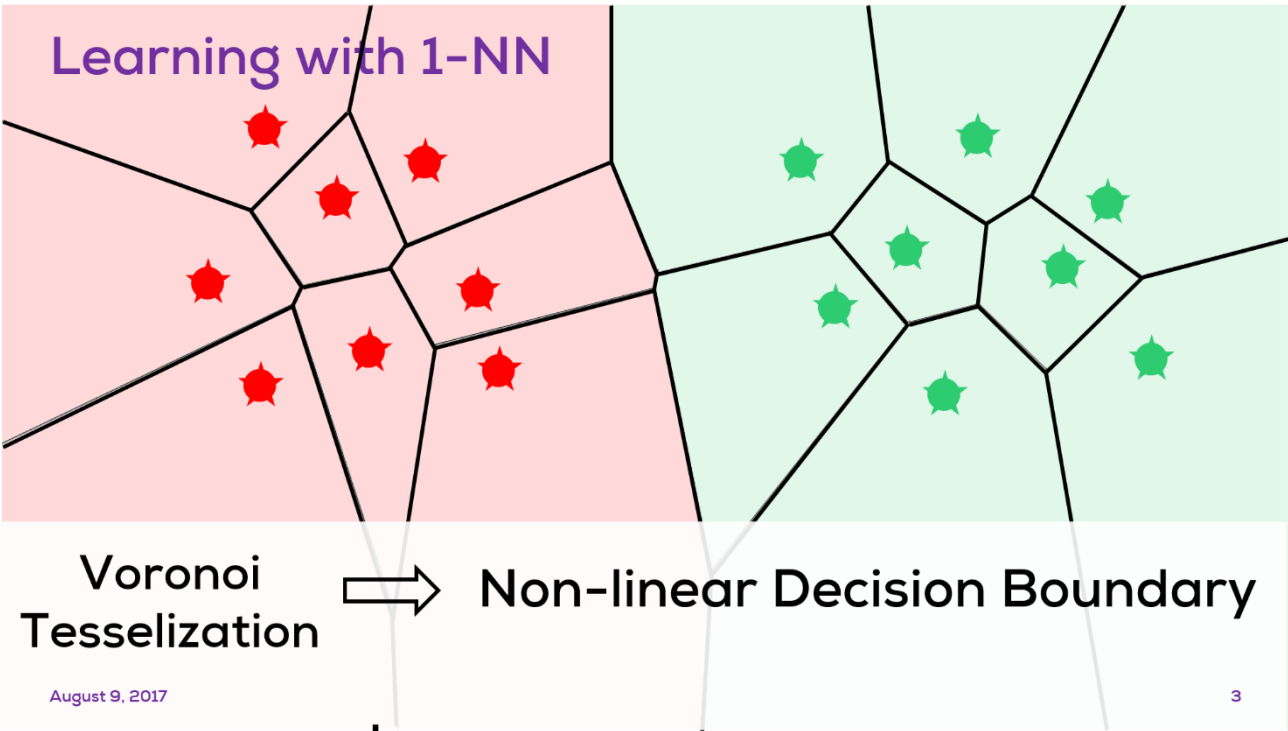
Local Methods-II

CS771: Introduction to Machine Learning

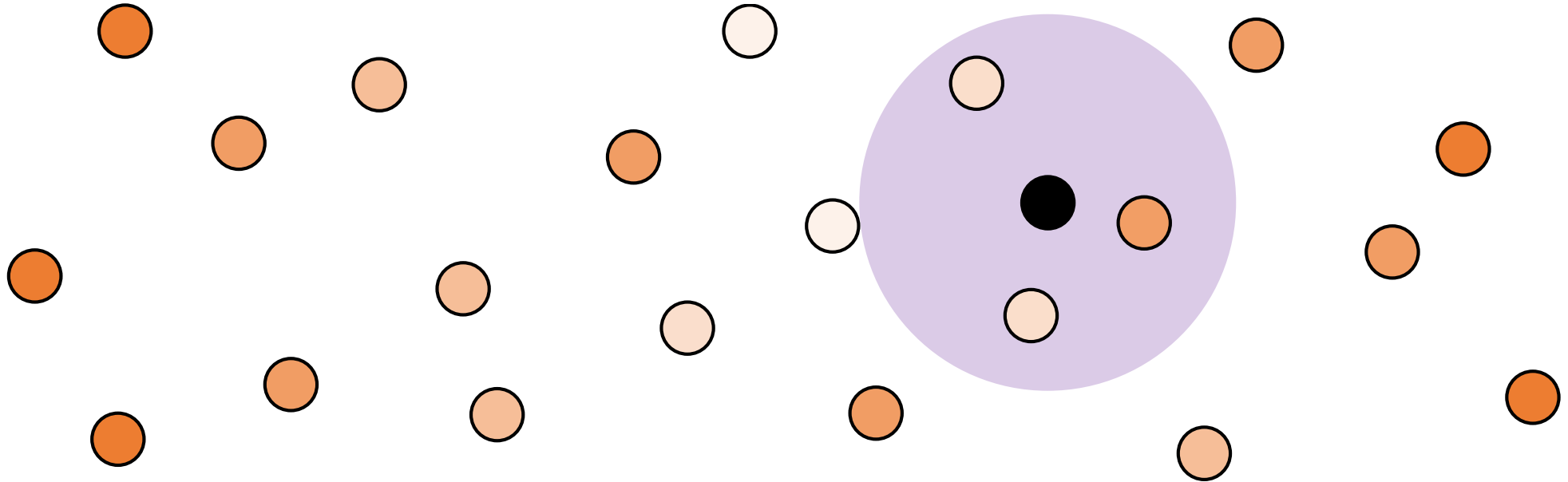
Purushottam Kar



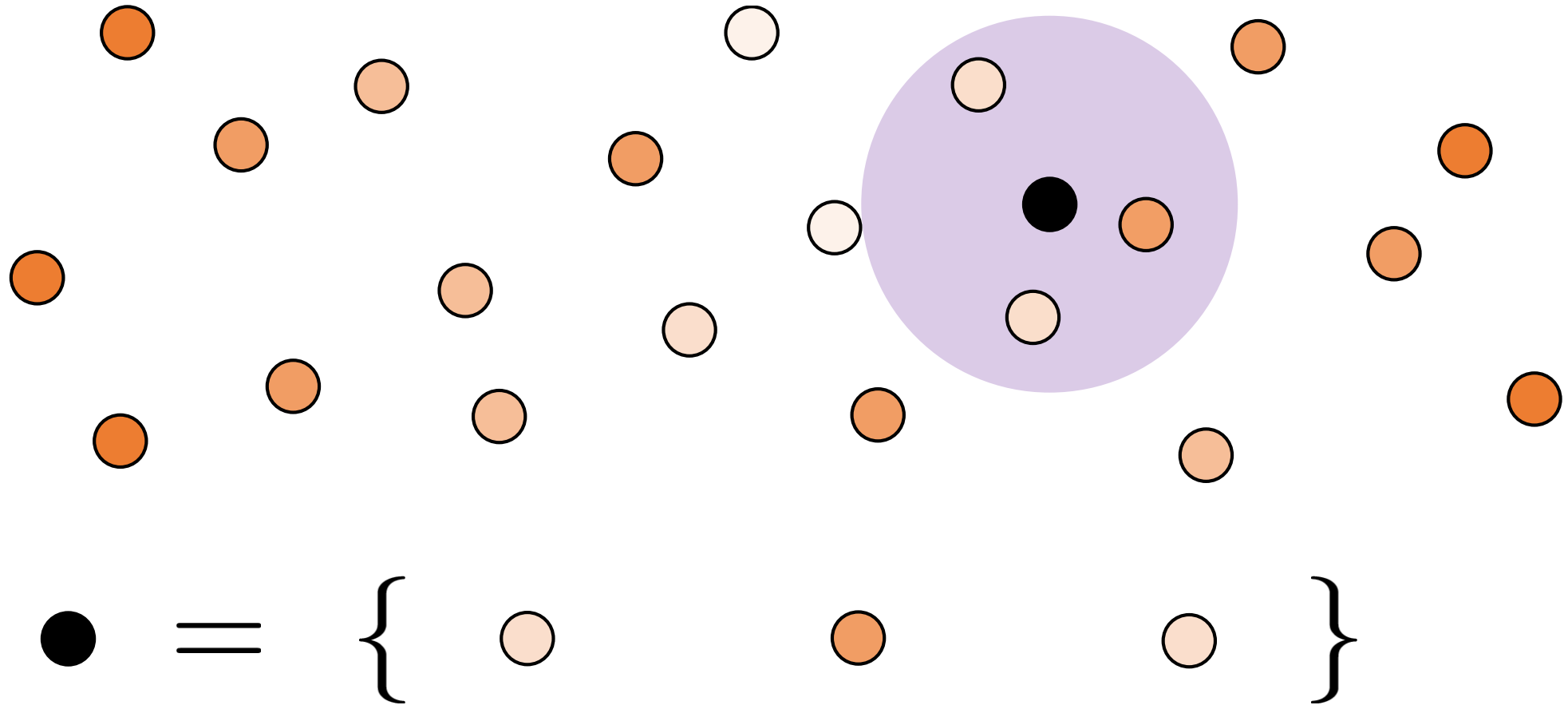
Recap



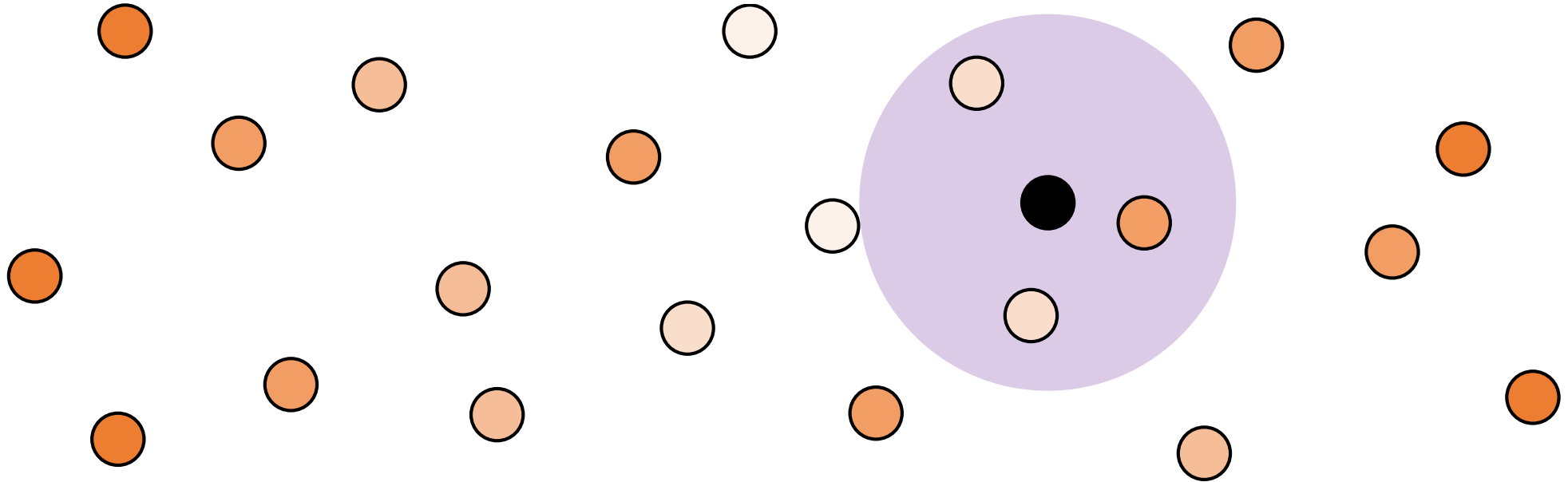
Regression with r-NN



Regression with r-NN

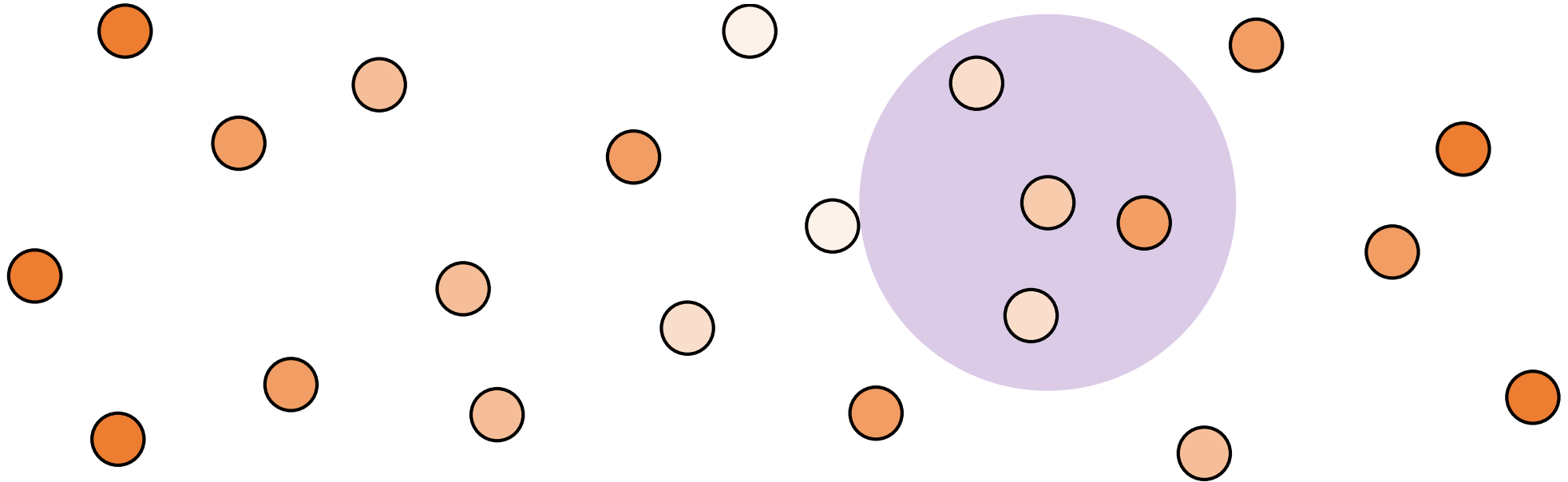


Regression with r-NN



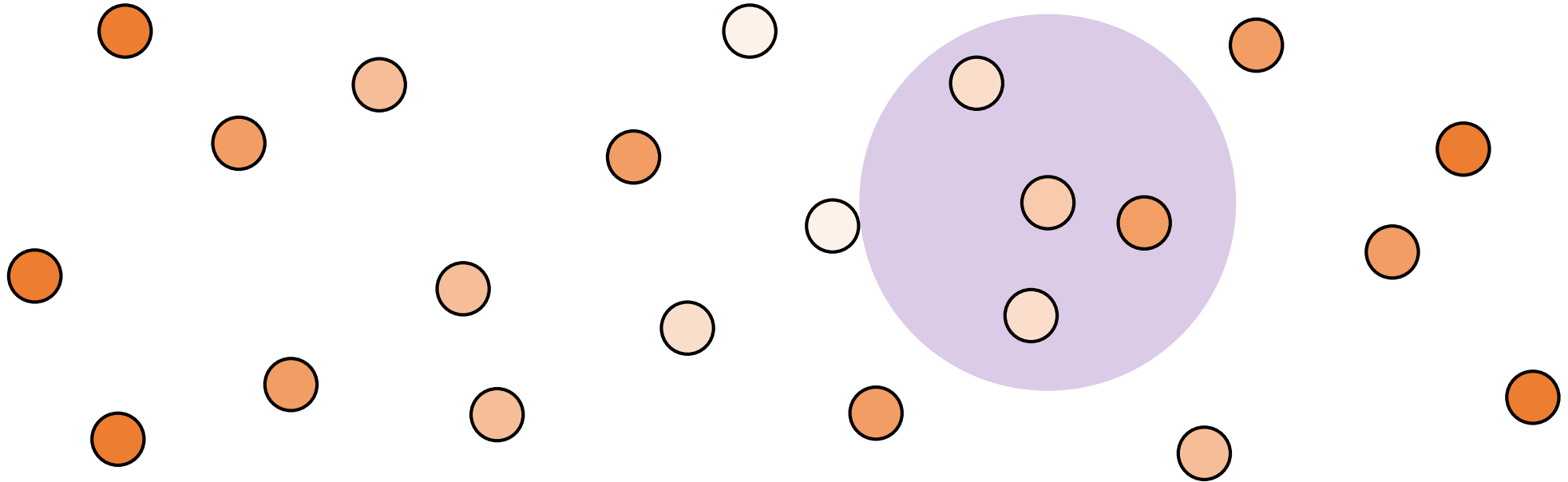
$$\bullet = \left\{ \frac{1}{3} \bullet + \frac{1}{3} \bullet + \frac{1}{3} \bullet \right\}$$

Regression with r-NN



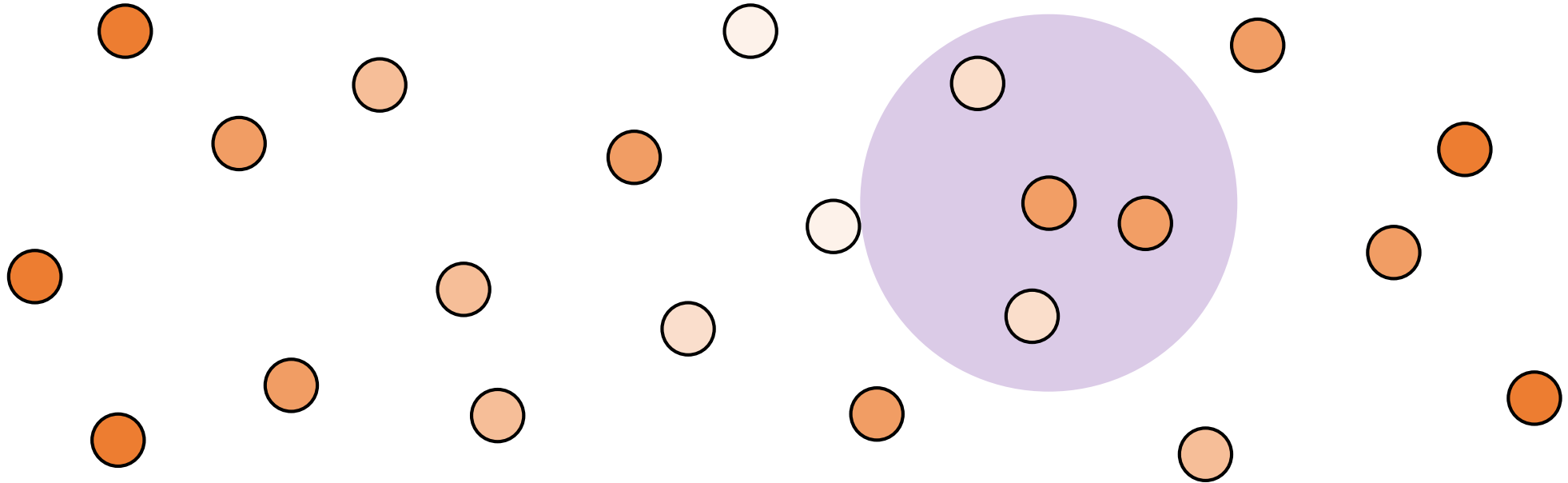
$$\text{light orange circle} = \left\{ \frac{1}{3} \text{light orange circle} + \frac{1}{3} \text{dark orange circle} + \frac{1}{3} \text{light orange circle} \right\}$$

Regression with Weighted r-NN



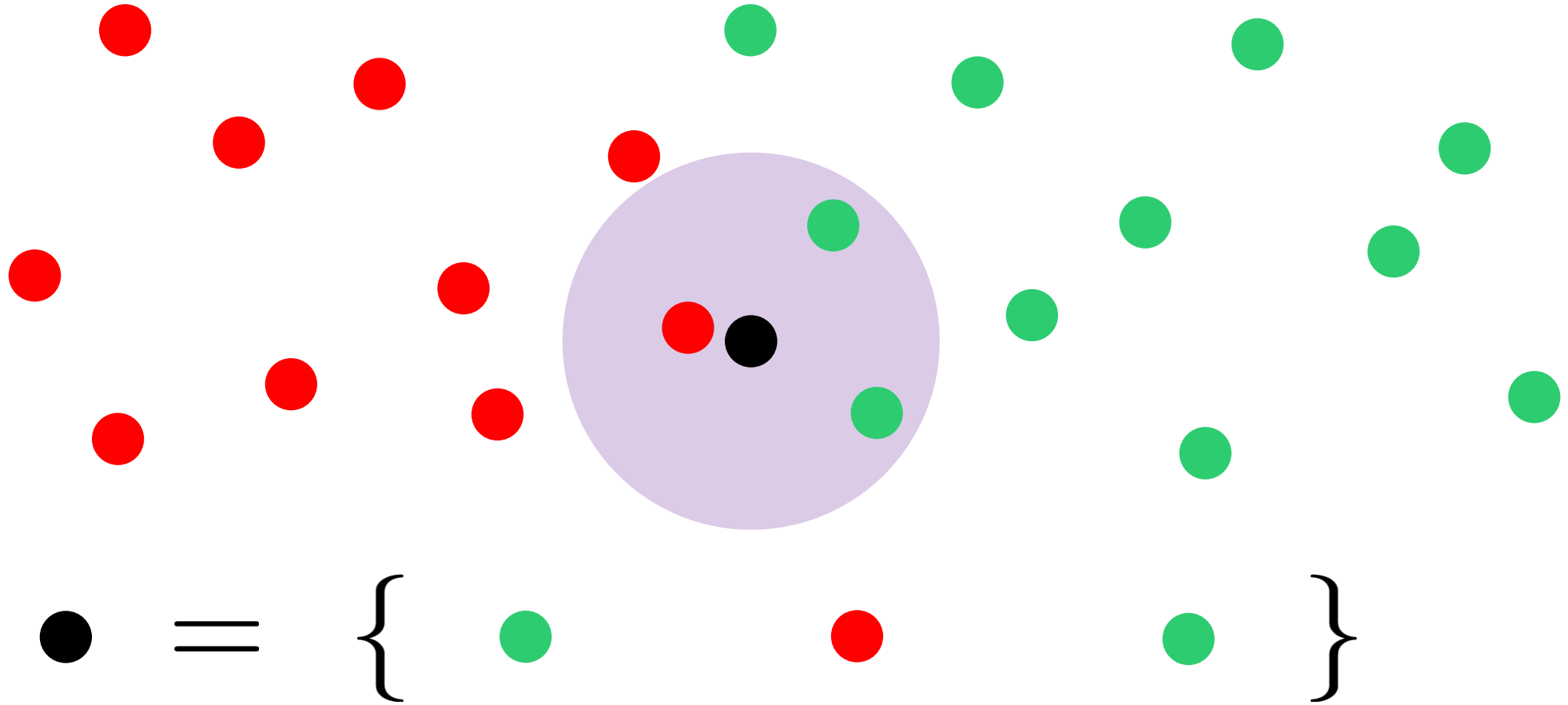
$$\text{orange circle} = \left\{ \frac{1}{5} \text{light orange circle} + \frac{3}{5} \text{orange circle} + \frac{1}{5} \text{light orange circle} \right\}$$

Regression with Weighted r-NN

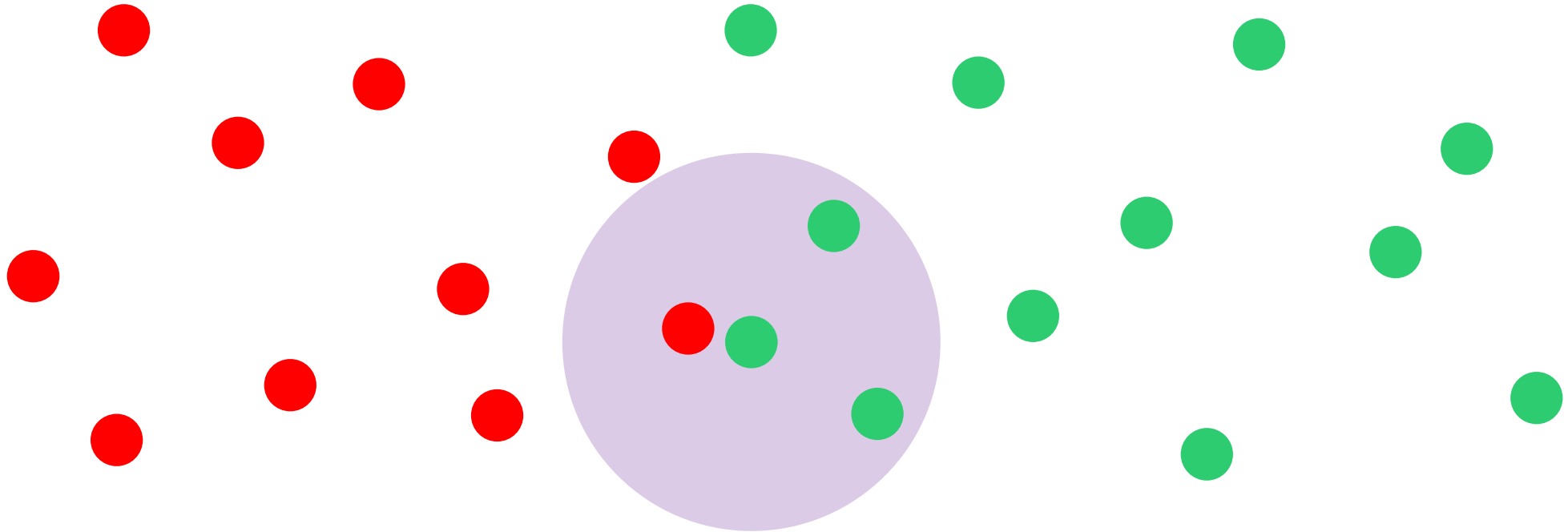


Can use a similar trick
with k-nn too!

Classification with Weighted r-NN

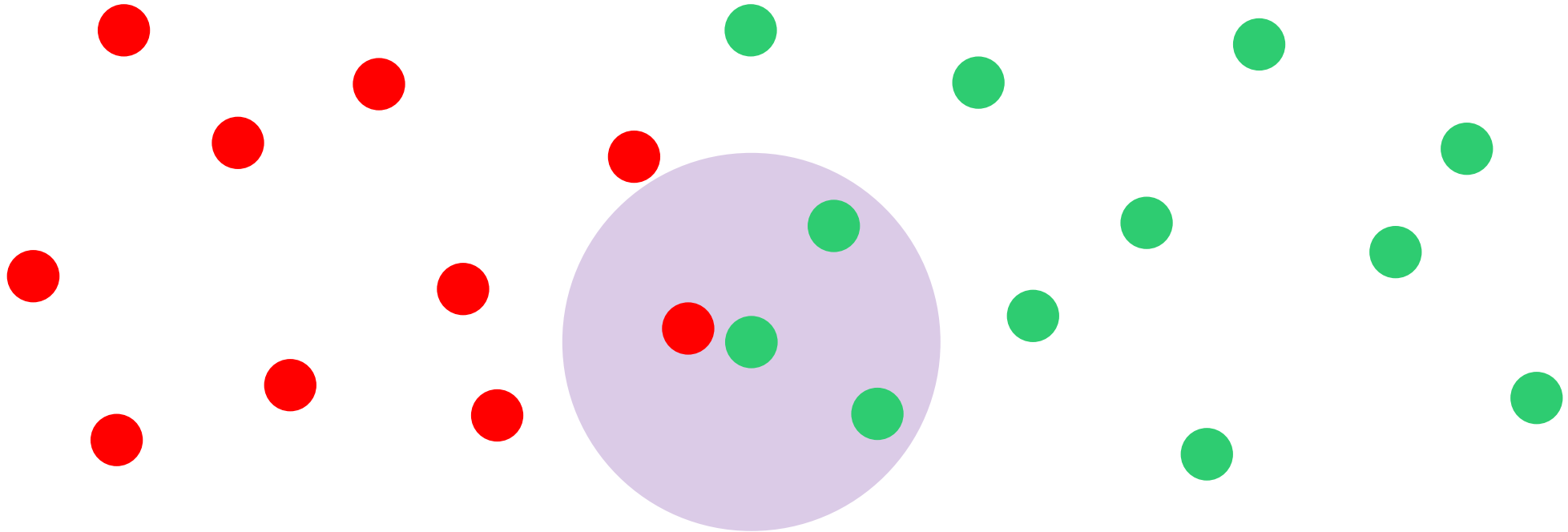


Classification with Weighted r-NN



$$\text{green dot} = \left\{ \frac{1}{3} (+1) + \frac{1}{3} (-1) + \frac{1}{3} (+1) \right\}$$

Classification with Weighted r-NN



$$\bullet = \left\{ \frac{1}{5} (+1) + \frac{3}{5} (-1) + \frac{1}{5} (+1) \right\}$$

All it takes are 20 questions

Guess the Movie!



पबल्लु

MAY						
Sun	Mo	Tu	We	Th	Fri	Sa
26	27	28	29	30	1	2
3	4	5	6	7	8	9
10	11	12	13	14	15	16
17	18	19	20	21	22	23
24	25	26	27	28	29	30
31	1	2	3	4	5	6

Original Language

Year of Release
2010,2011,...2017



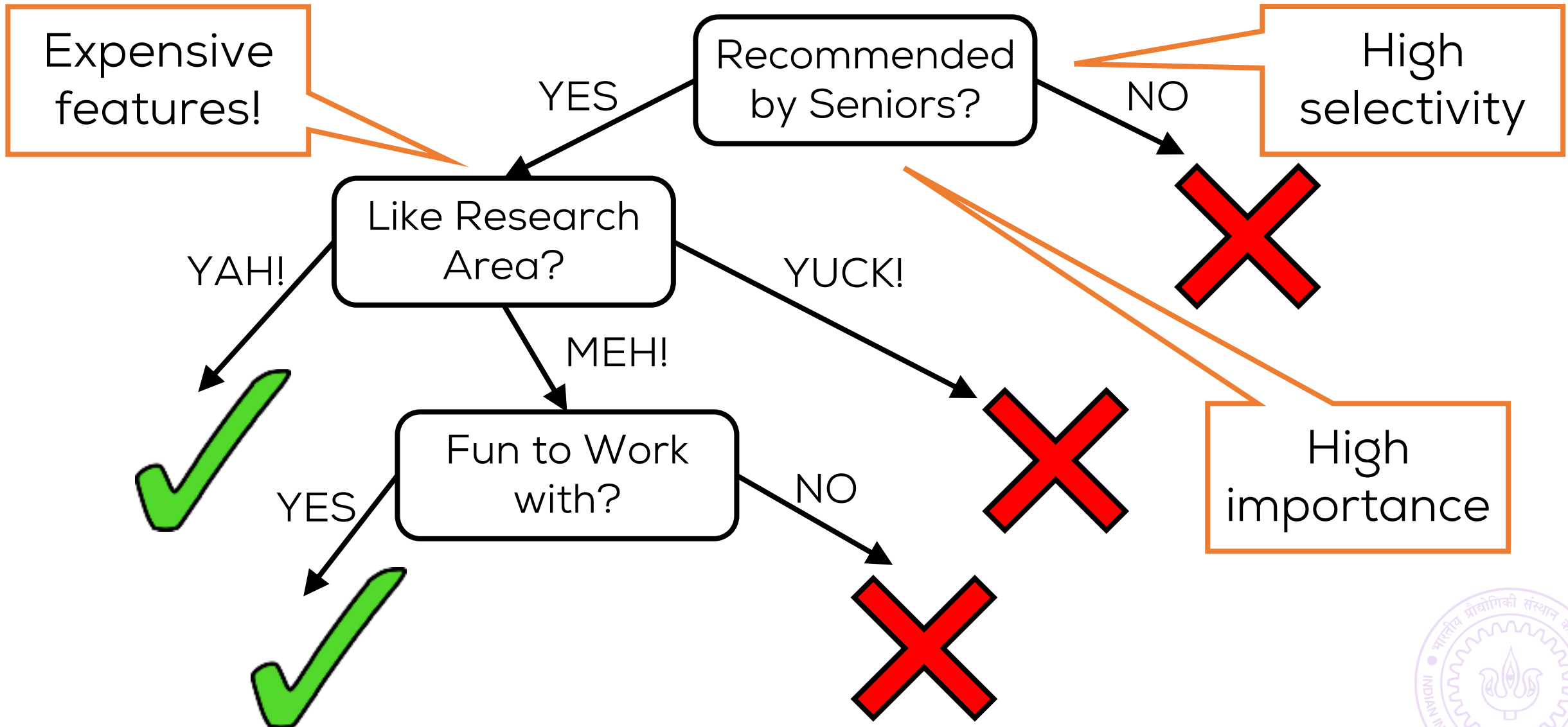
Box-Office Collection

Low (< INR 100 cr)

Medium (INR 100-1000 cr)

High (> 1000 cr)

Choose an Adviser!



Learning with Decision Trees

... and Decision Forests

Decision Trees

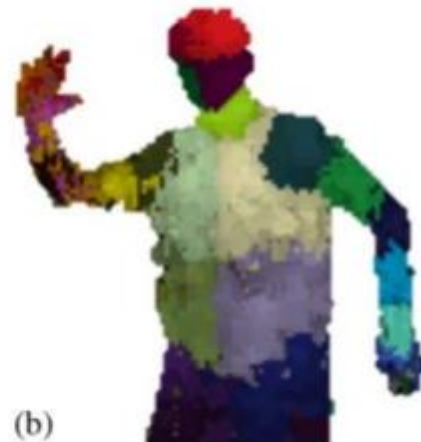
- Very versatile
- Easy to implement
- Extremely fast at test time
- Easy to interpret by humans
- Can be voluminous
- Can overfit badly
- Very very popular



Gaming – Kinect for Xbox 360

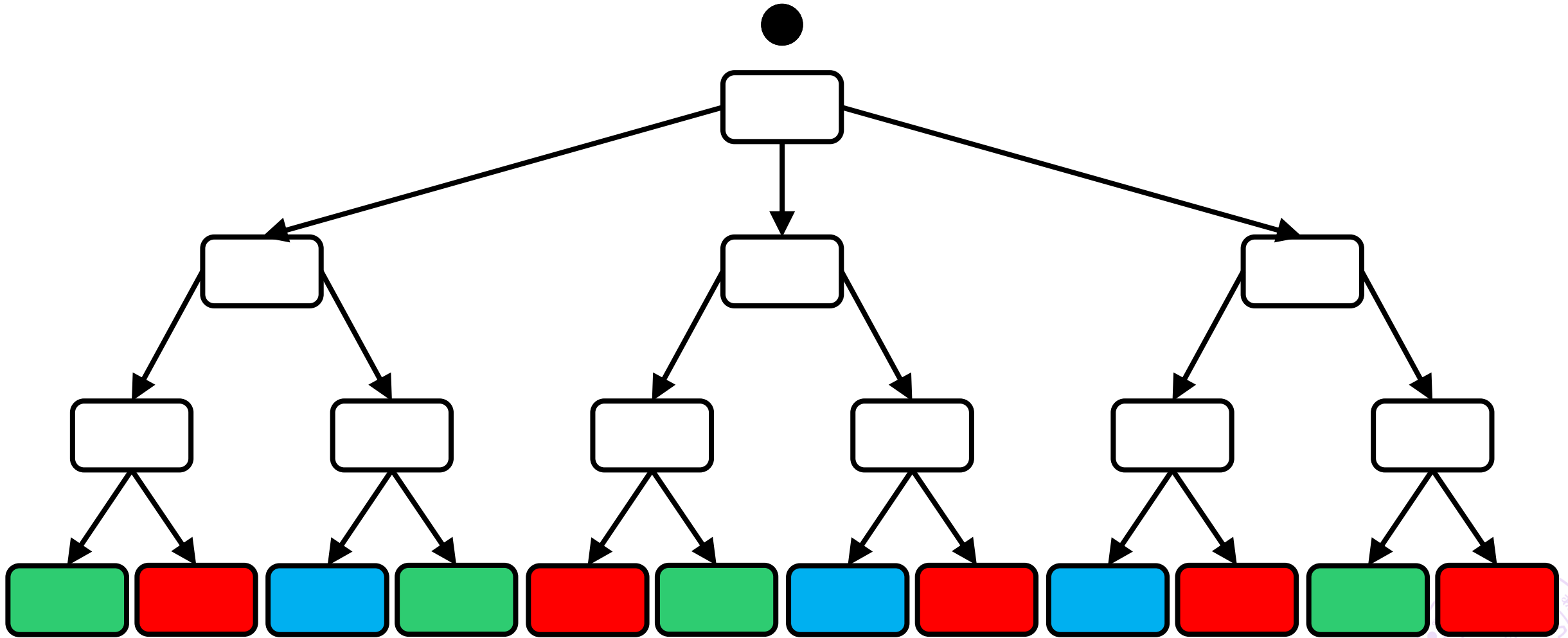


Depth map

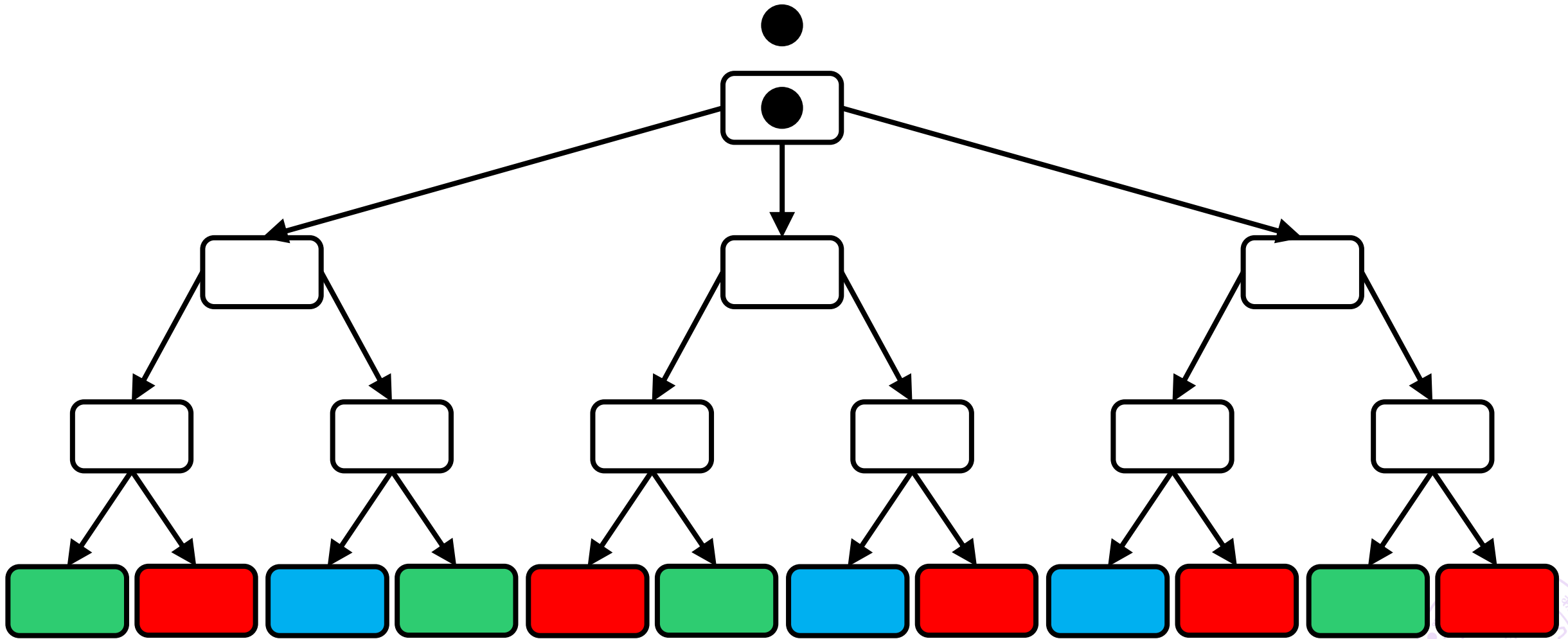


Body part classification

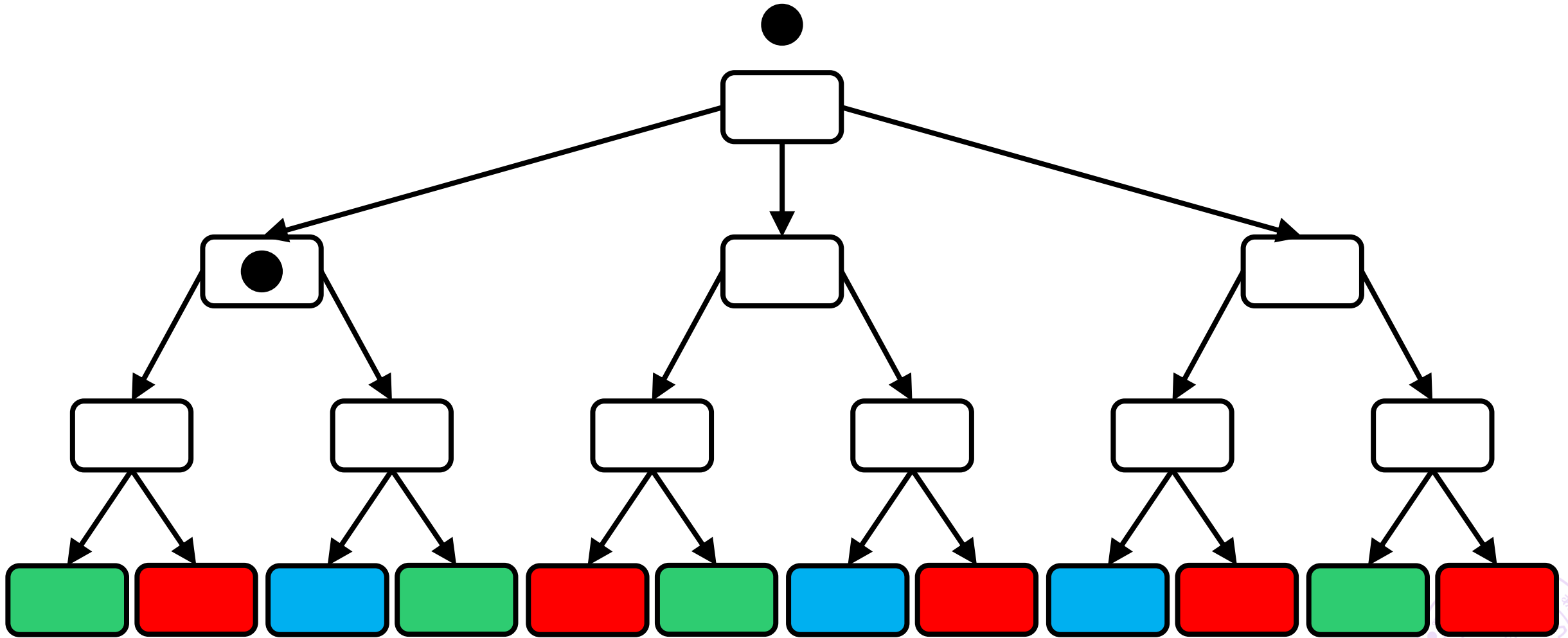
Classification with Decision Trees



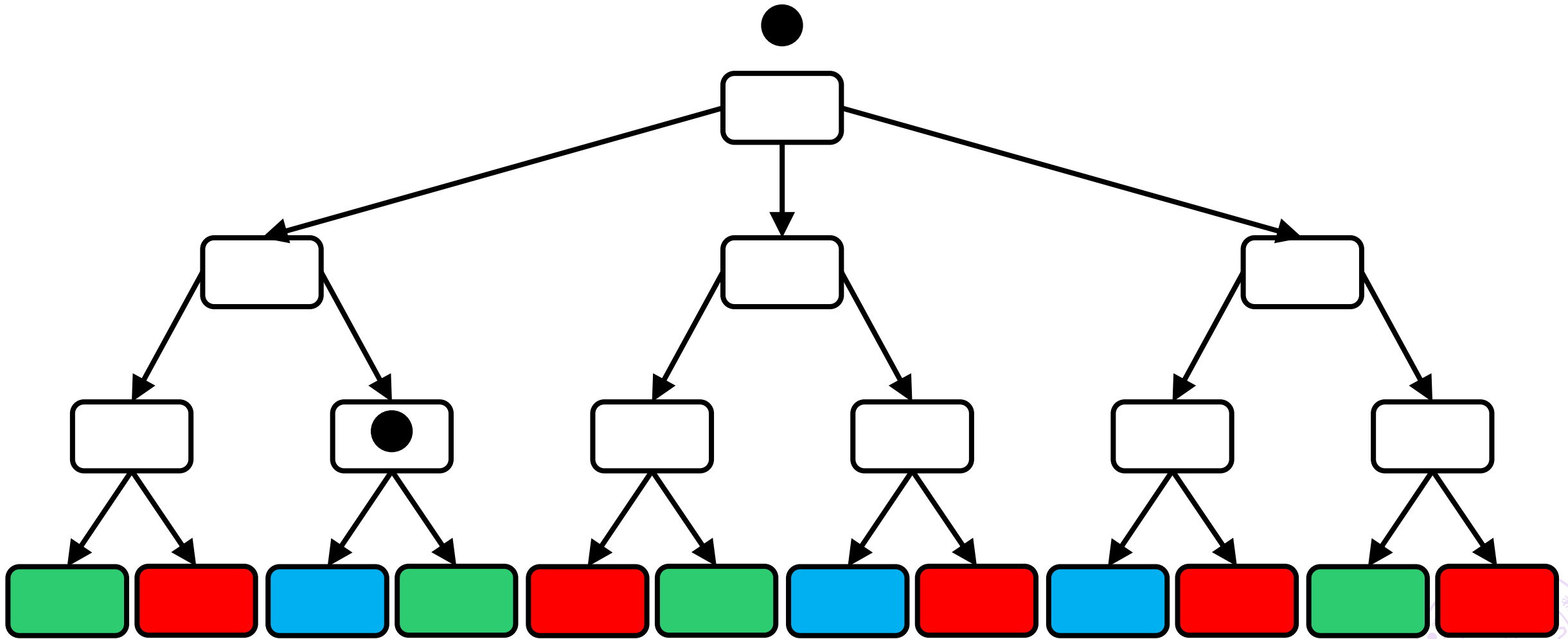
Classification with Decision Trees



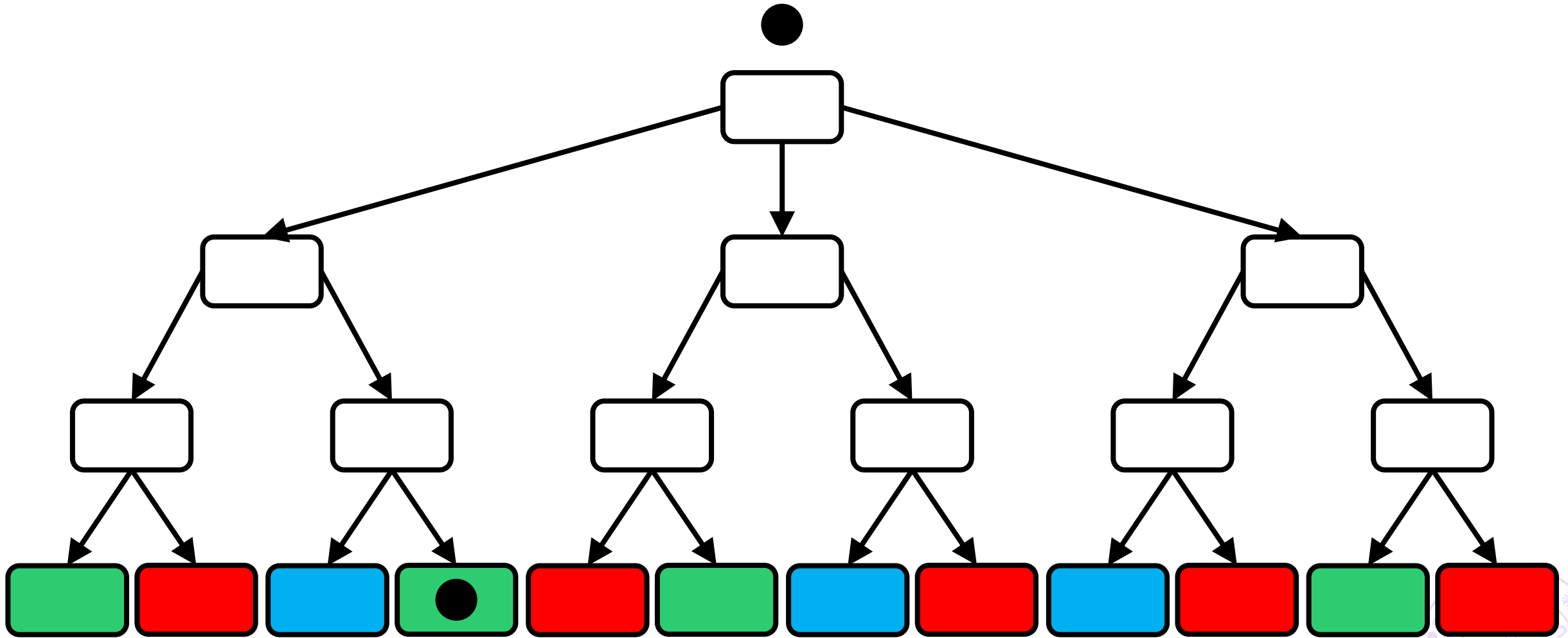
Classification with Decision Trees



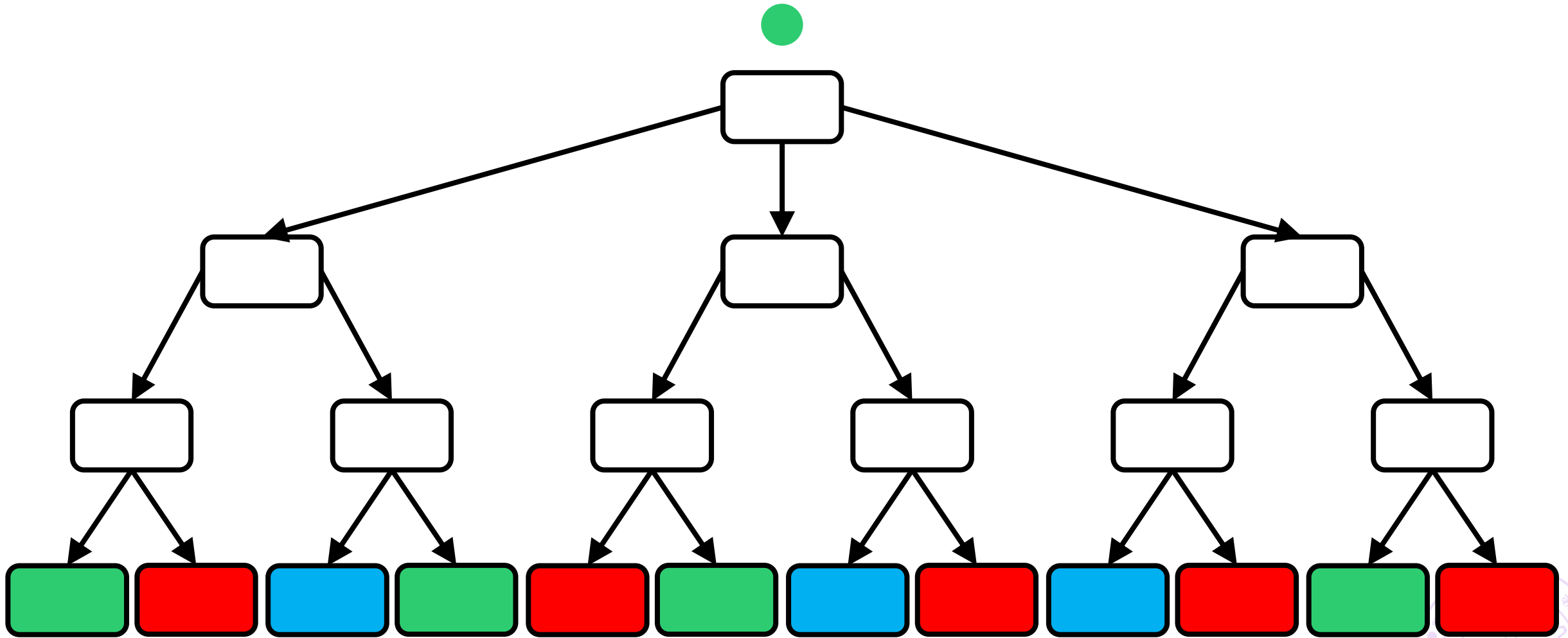
Classification with Decision Trees



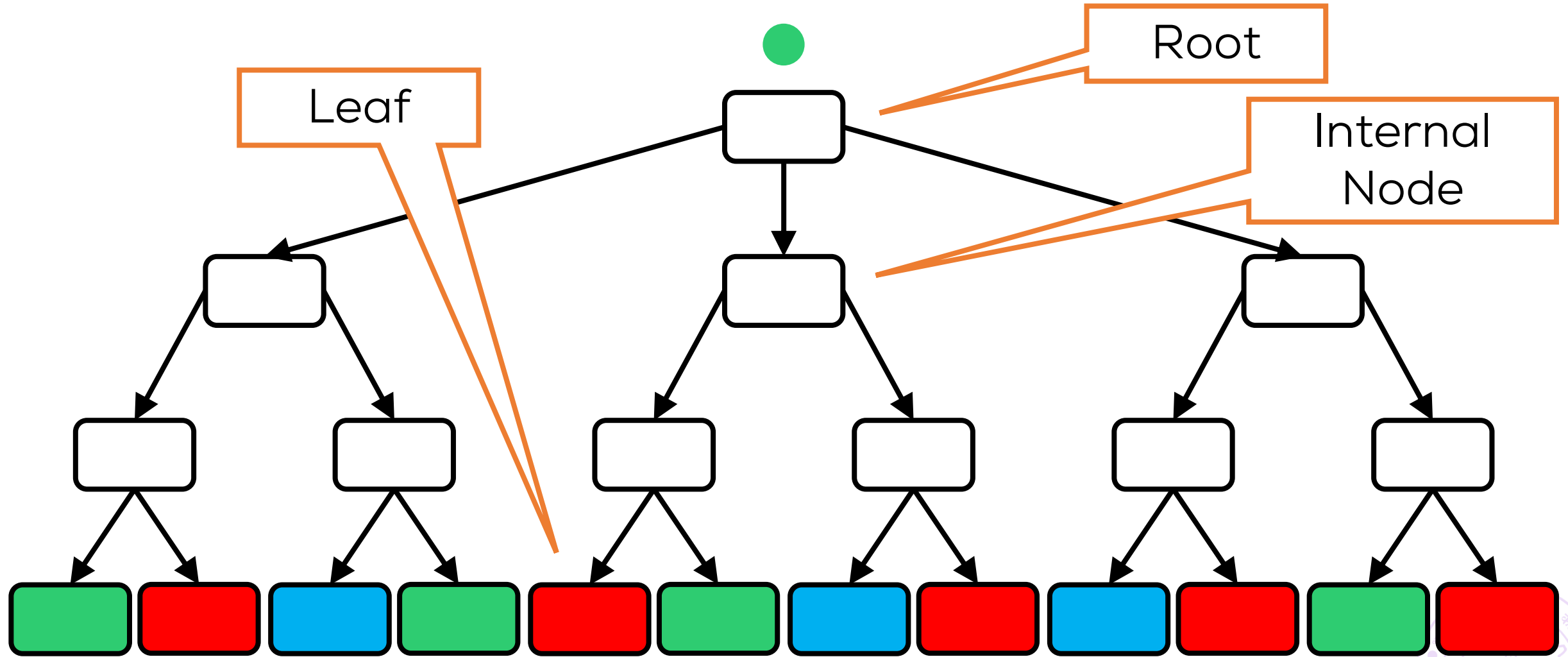
Classification with Decision Trees



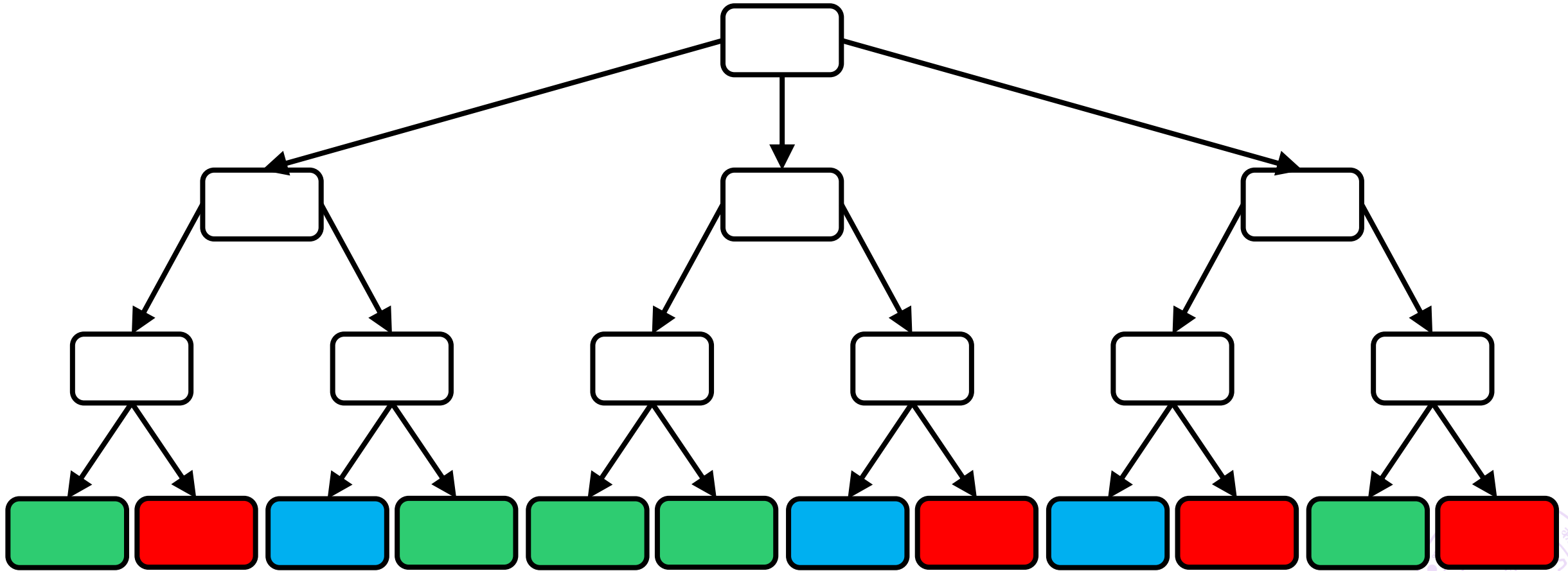
Classification with Decision Trees



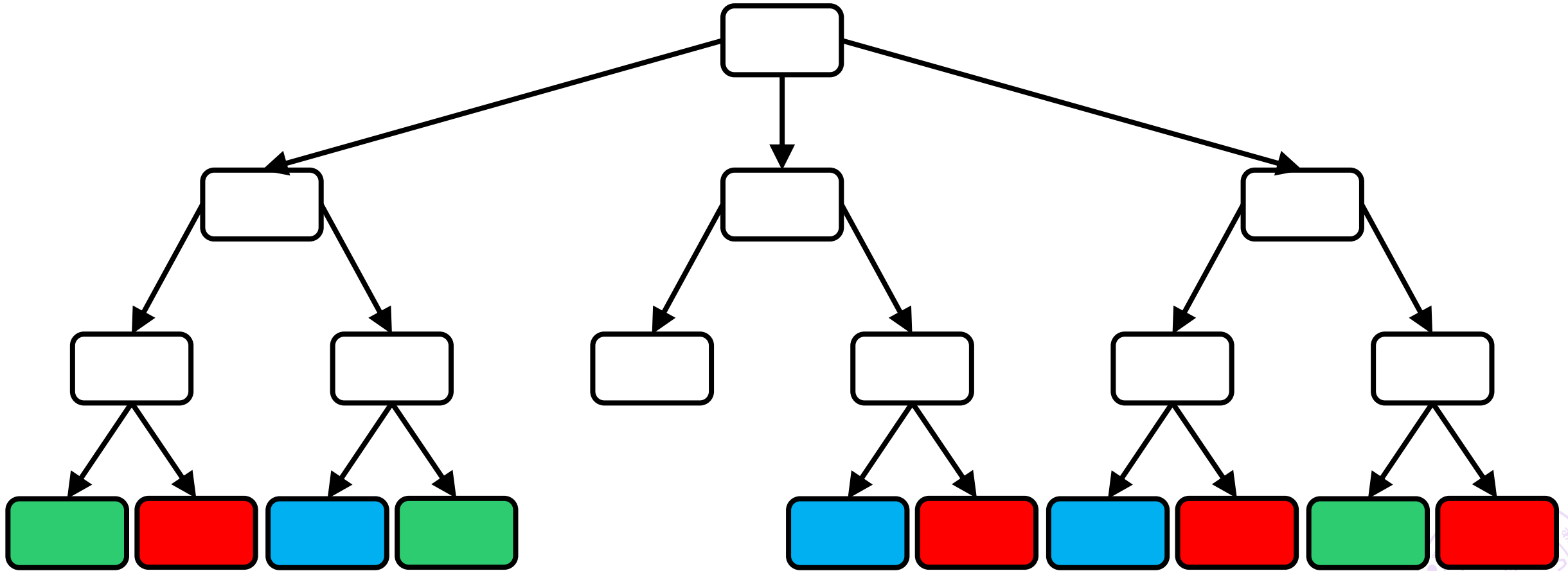
Classification with Decision Trees



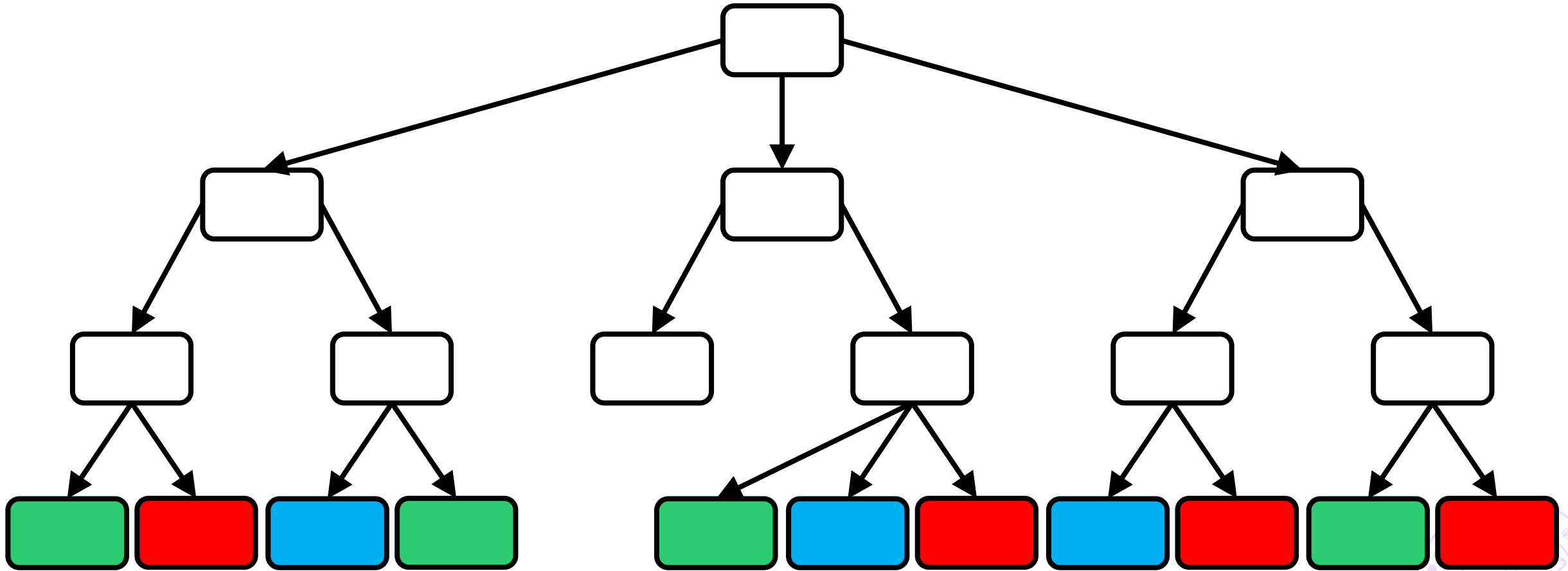
Decision Trees – all shapes and sizes



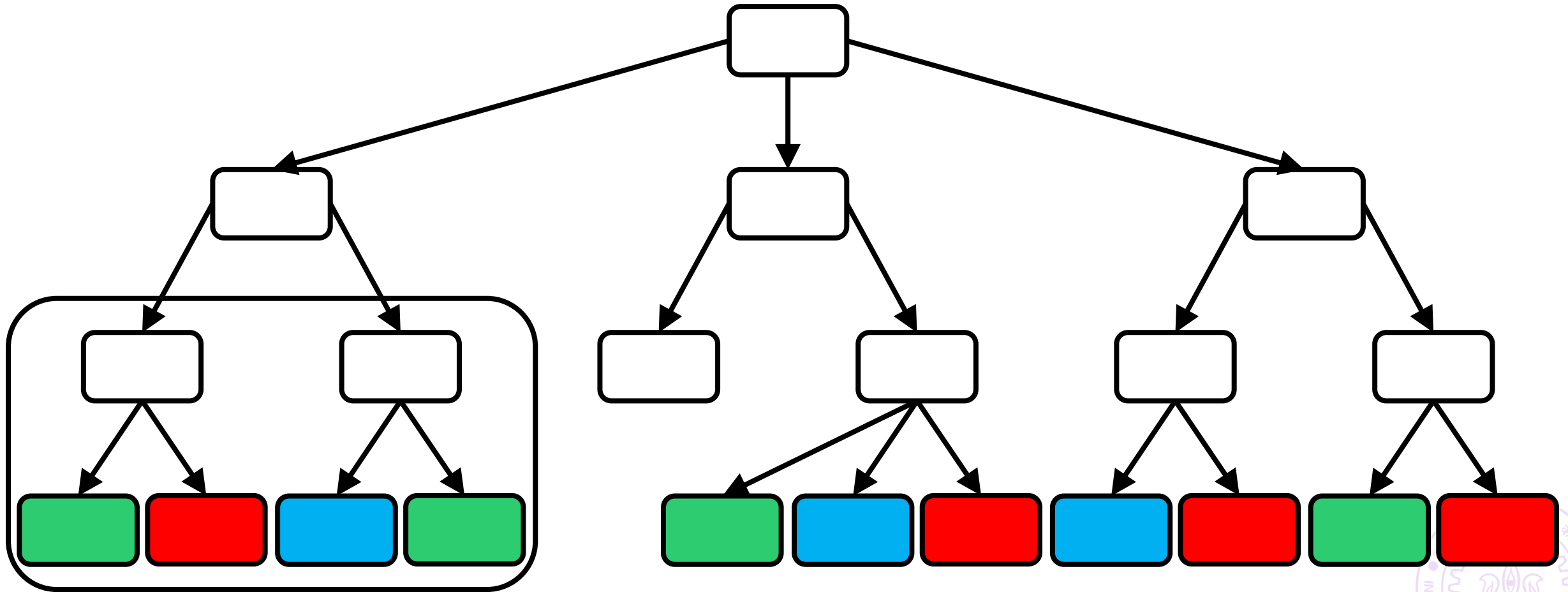
Decision Trees – all shapes and sizes



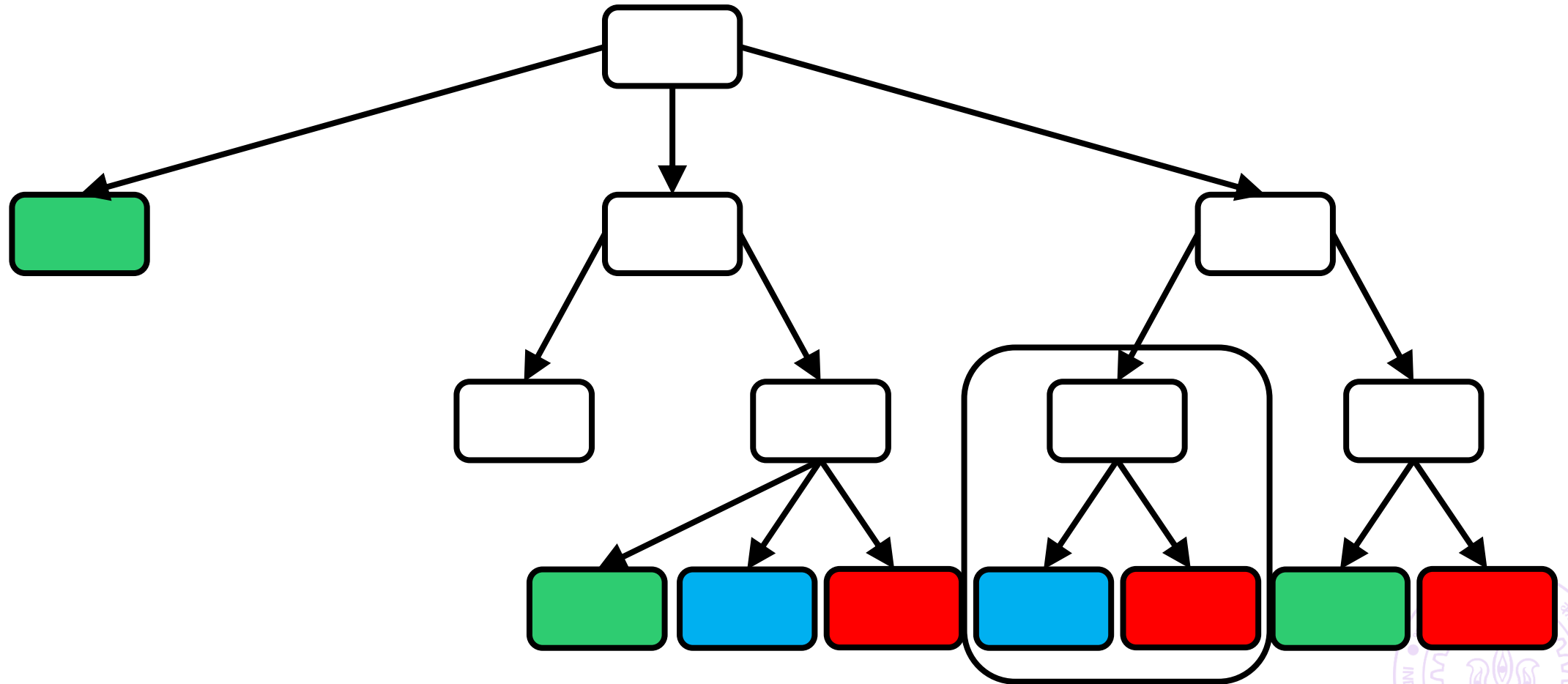
Decision Trees – all shapes and sizes



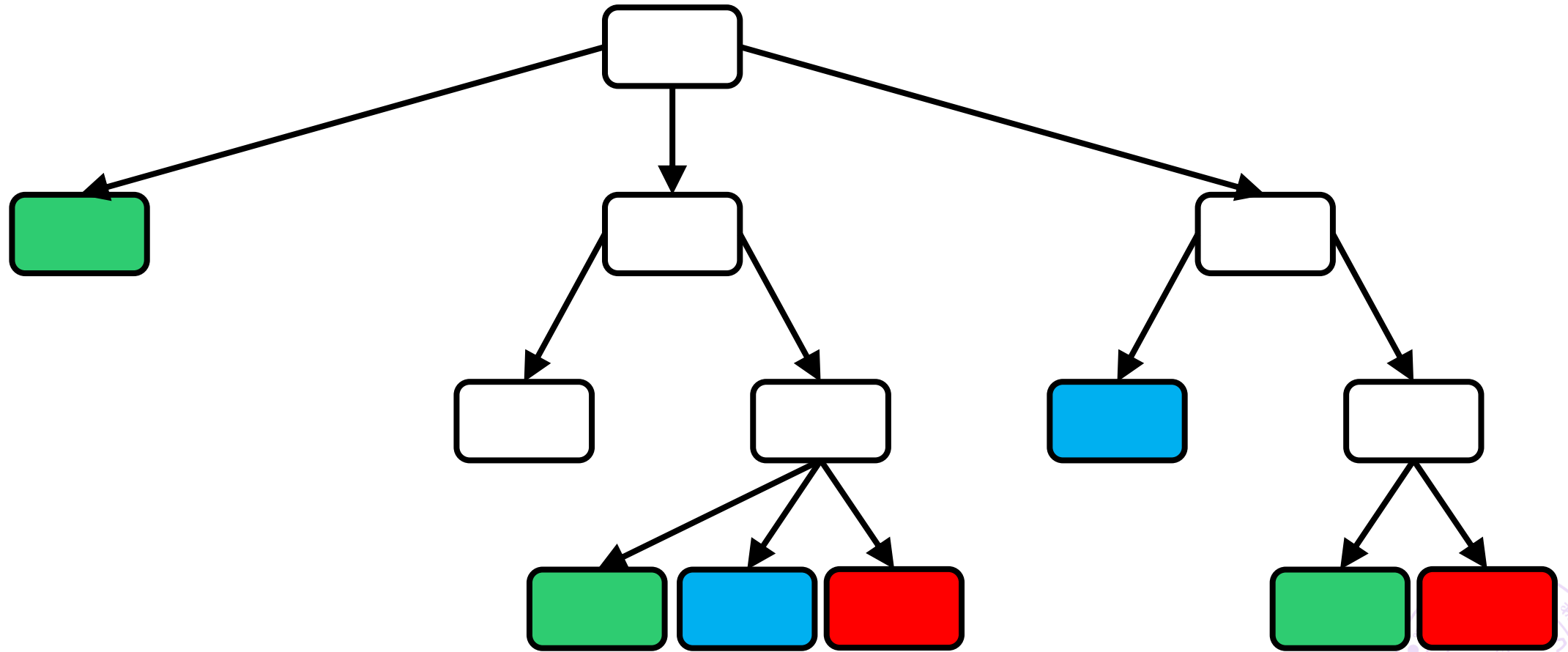
Decision Trees – all shapes and sizes



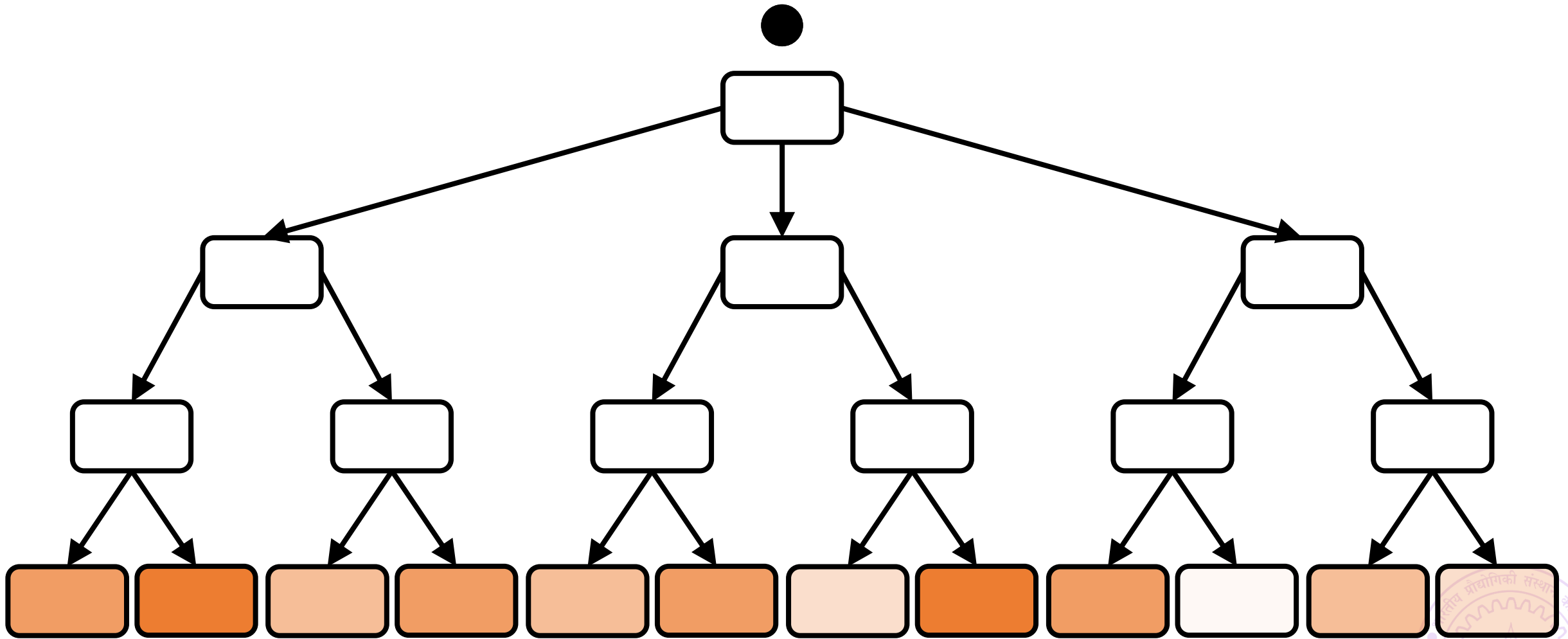
Decision Trees – all shapes and sizes



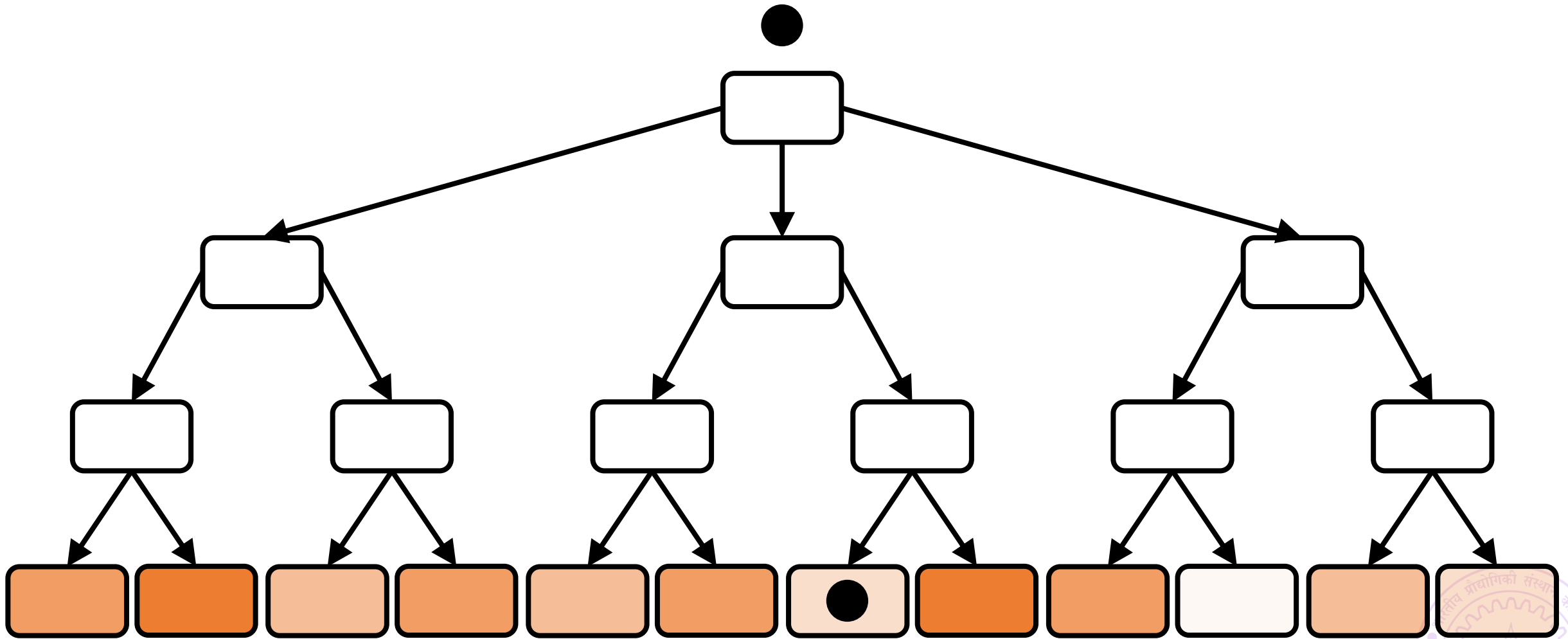
Decision Trees – all shapes and sizes



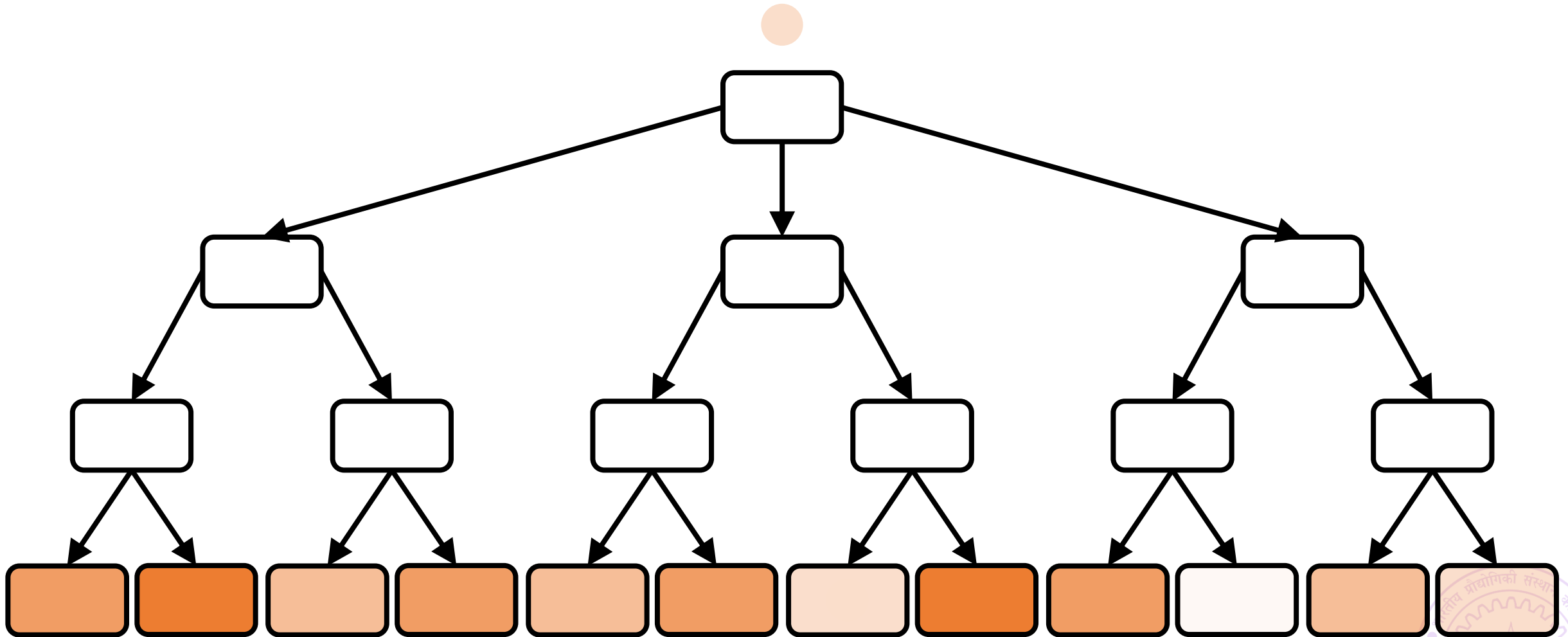
Regression with Decision Trees



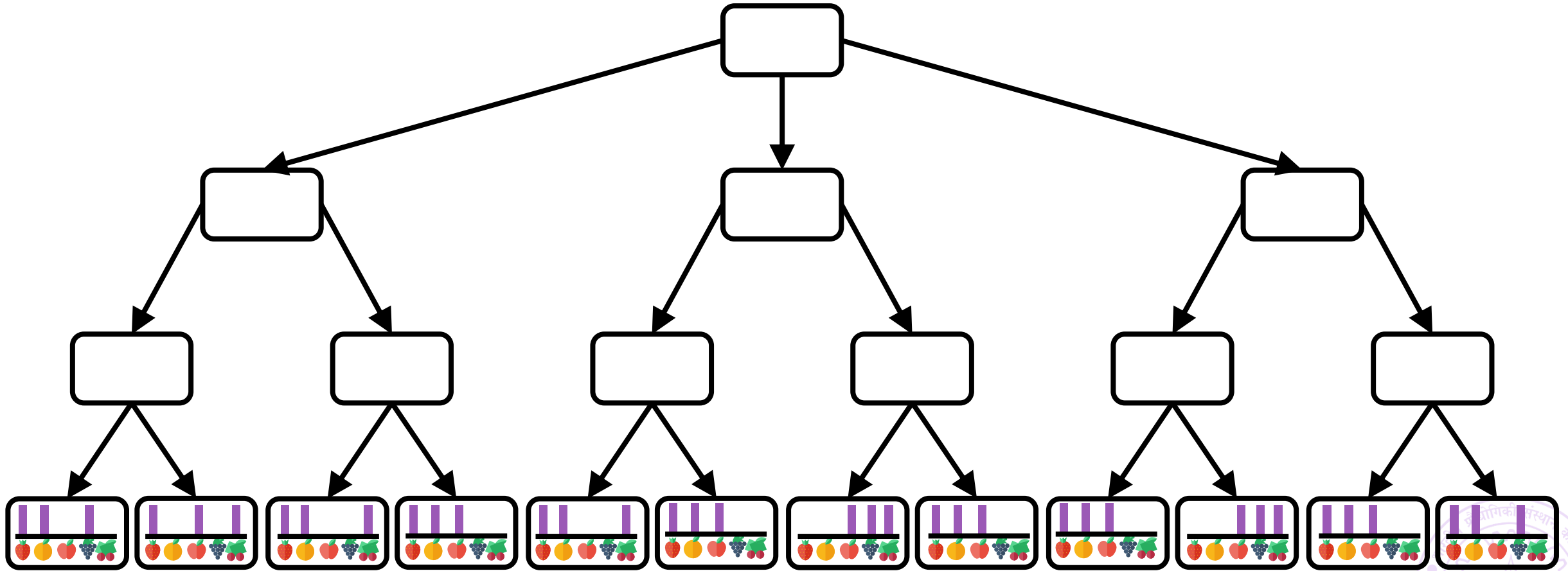
Regression with Decision Trees



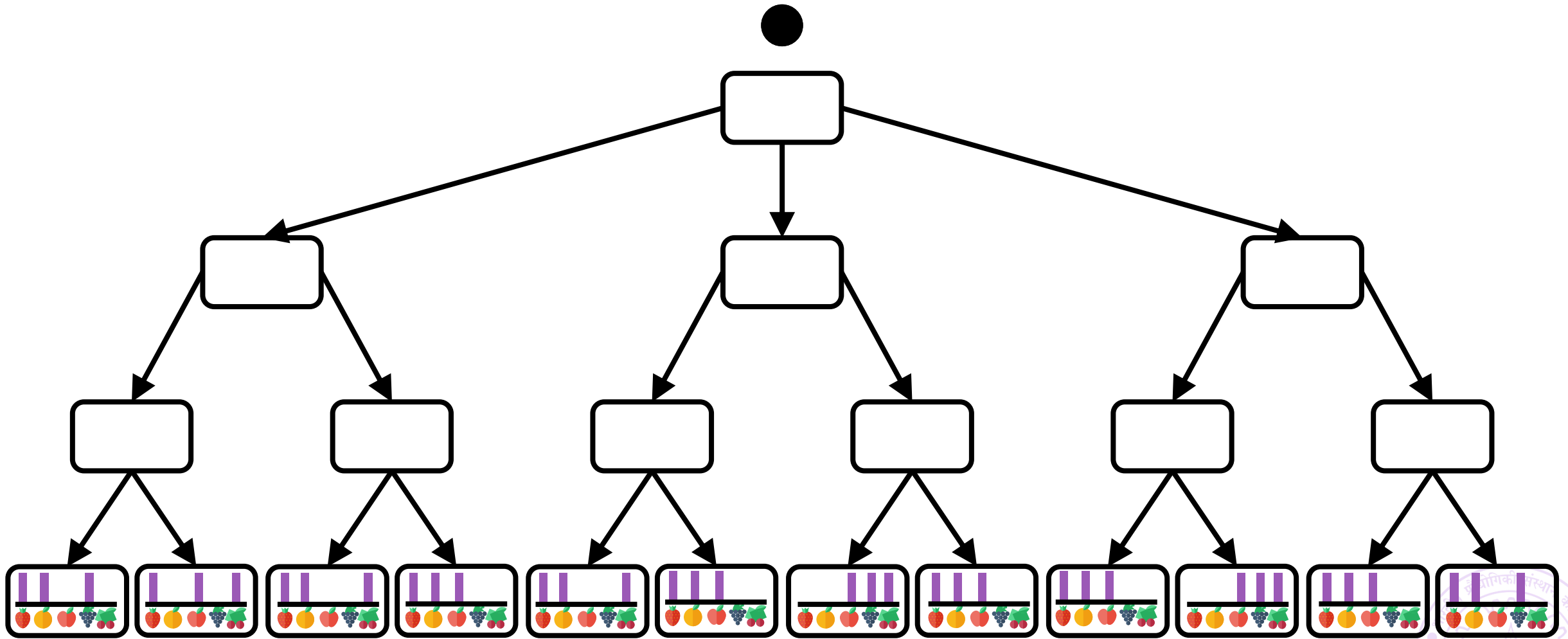
Regression with Decision Trees



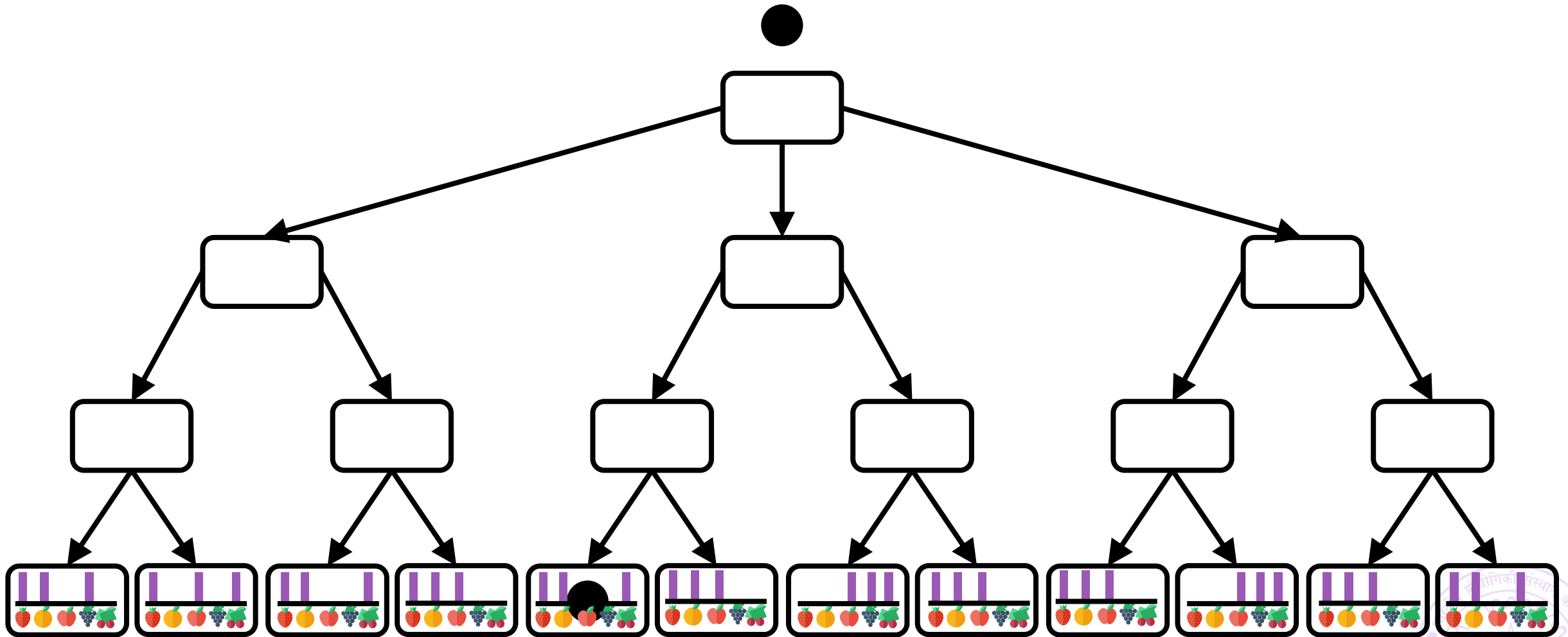
Multi-label Classification with Decision Trees



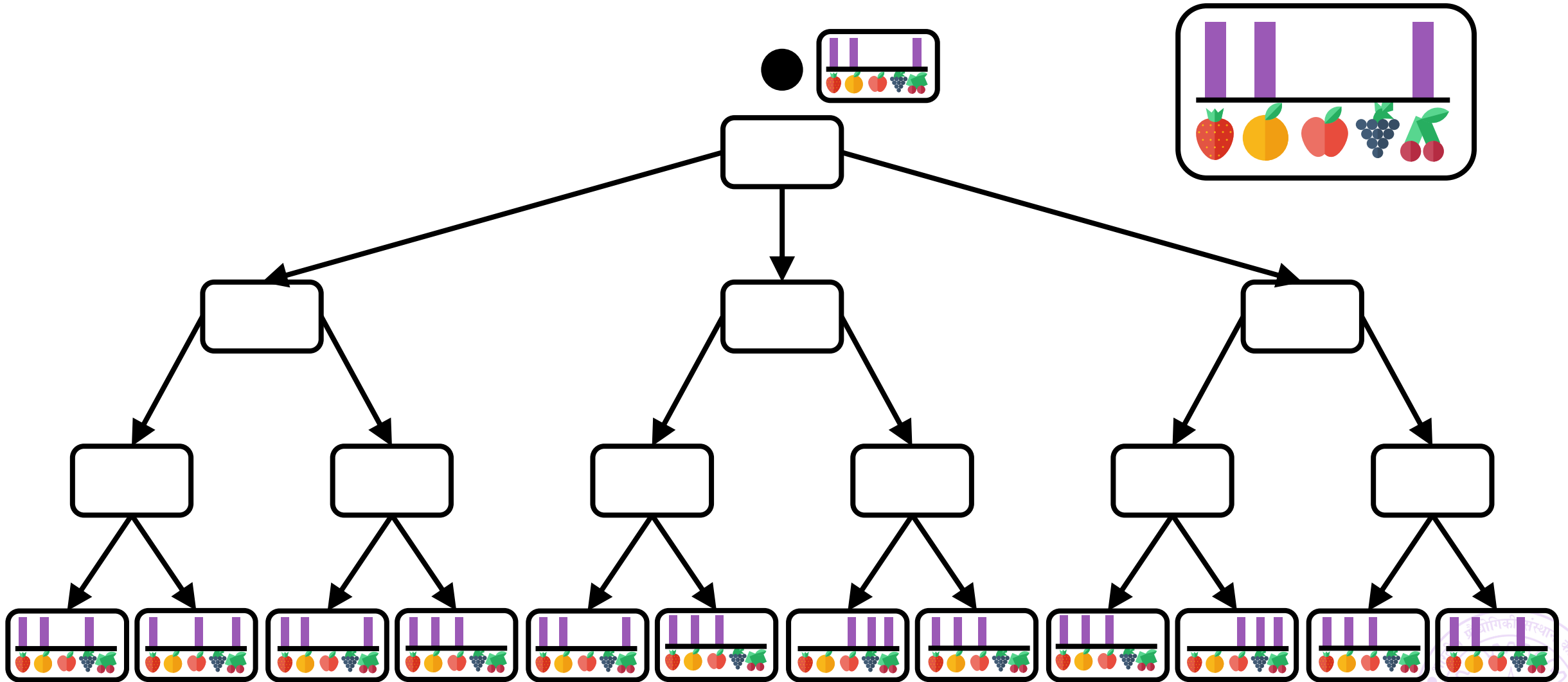
Multi-label Classification with Decision Trees



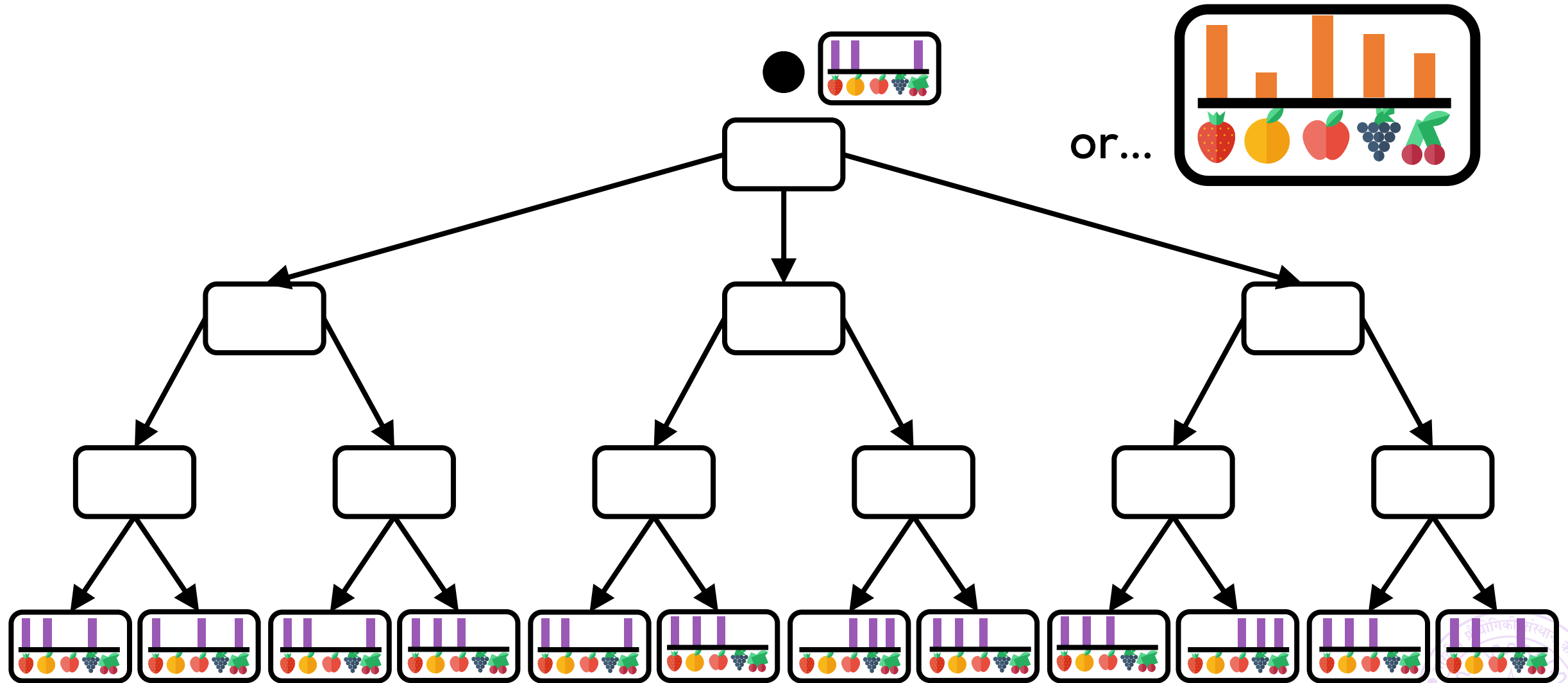
Multi-label Classification with Decision Trees



Multi-label Classification with Decision Trees



Multi-label Classification with Decision Trees



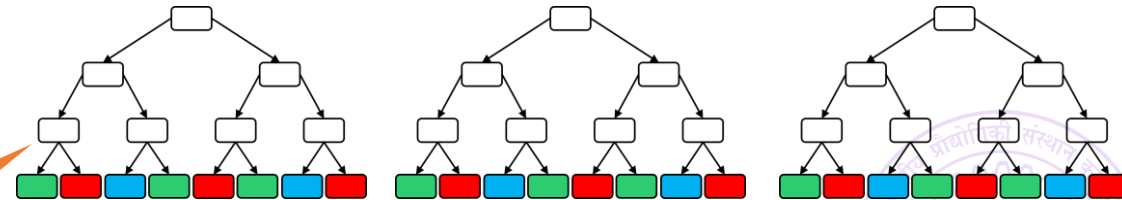
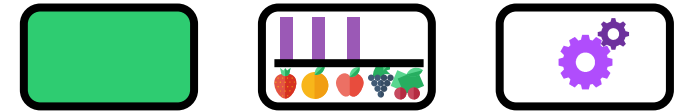
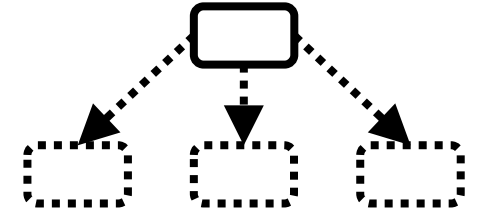
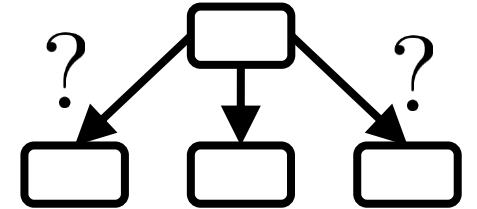
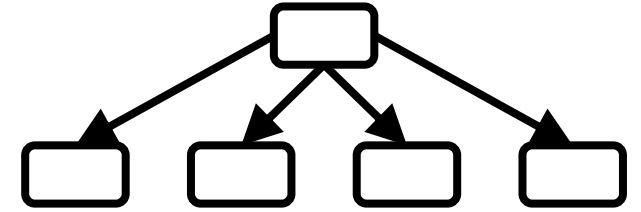
DT Learning Algorithms

August 11, 2017



The Questions that Matter

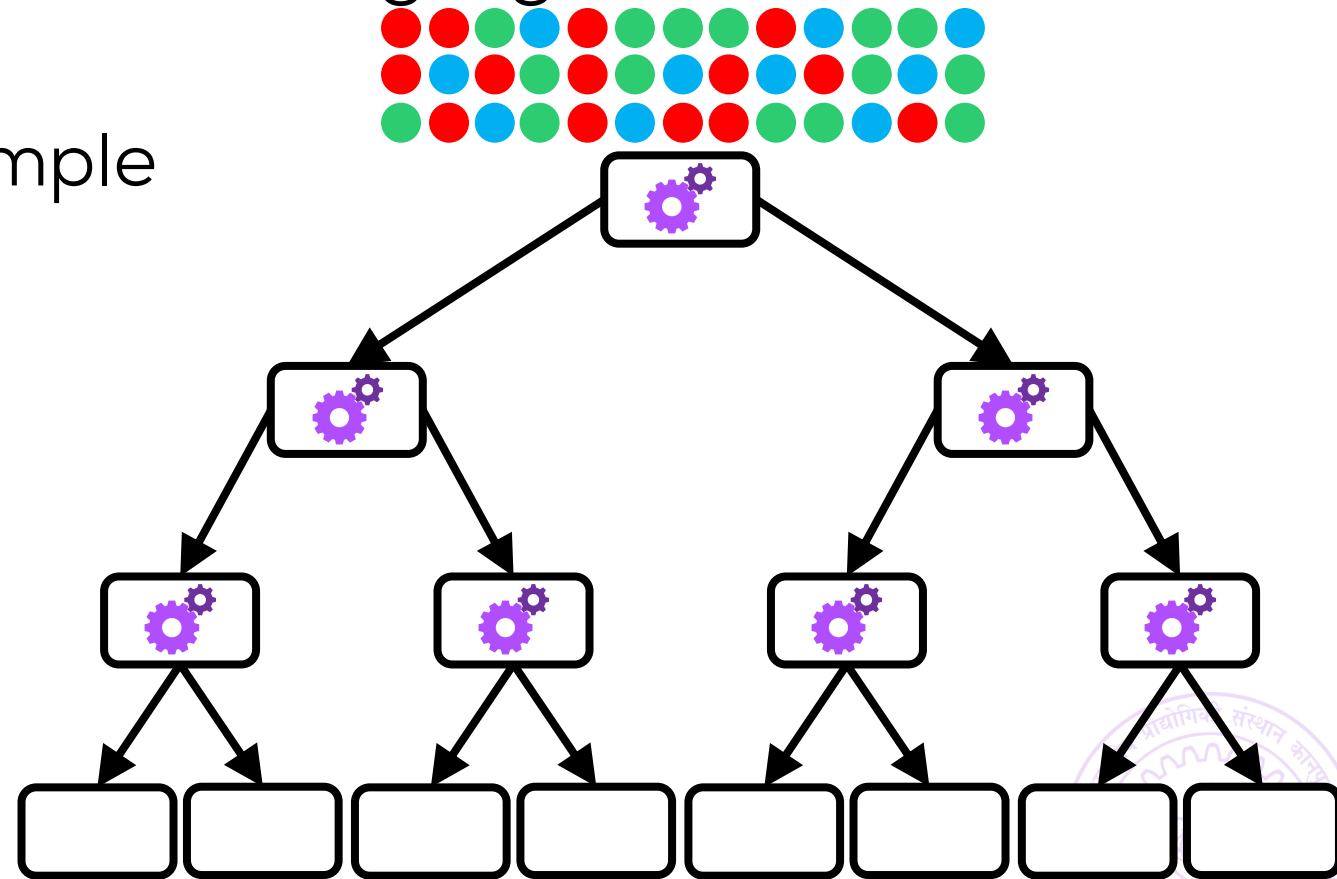
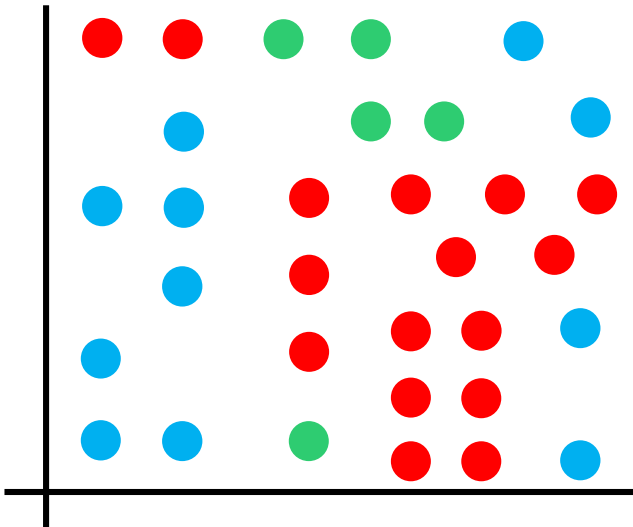
- How many children should a node have?
- How to send data points to children?
- When to stop splitting and make the node a leaf?
- What to do at a leaf?
- How many trees to train?



Model selection,
memory constraints

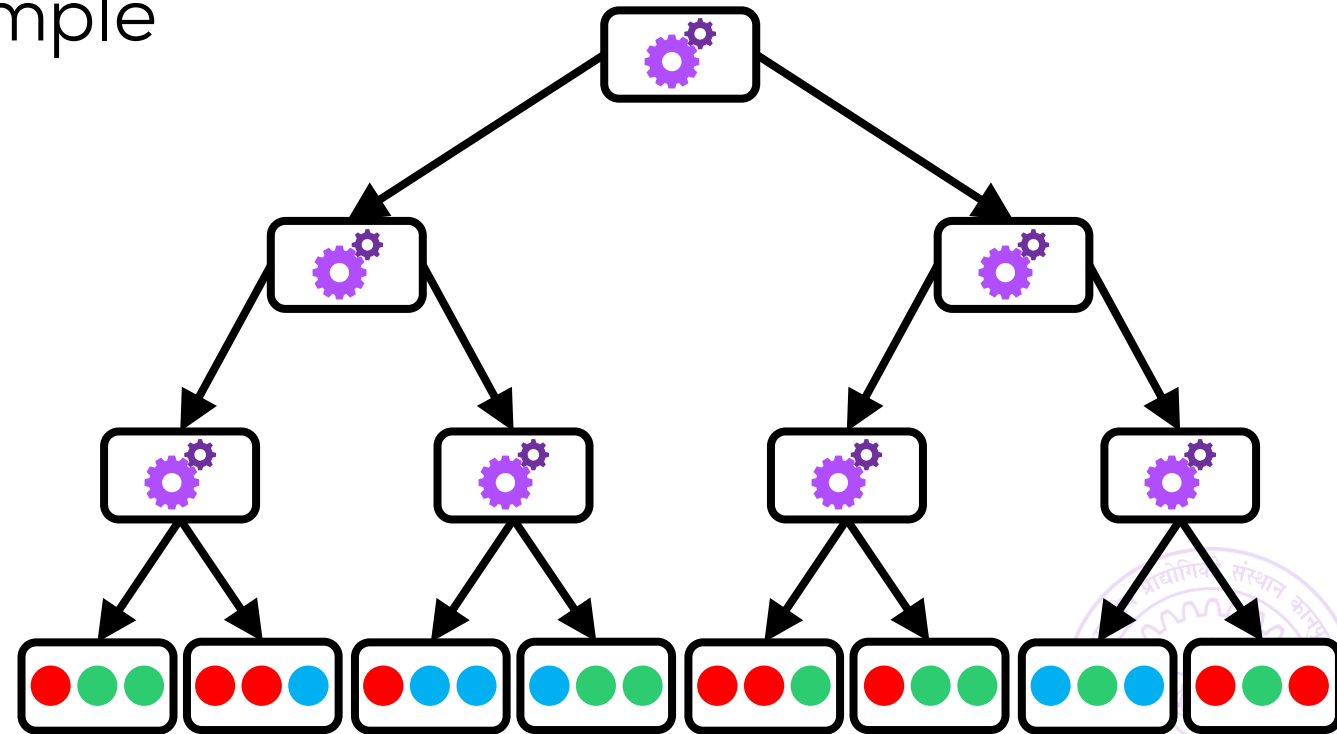
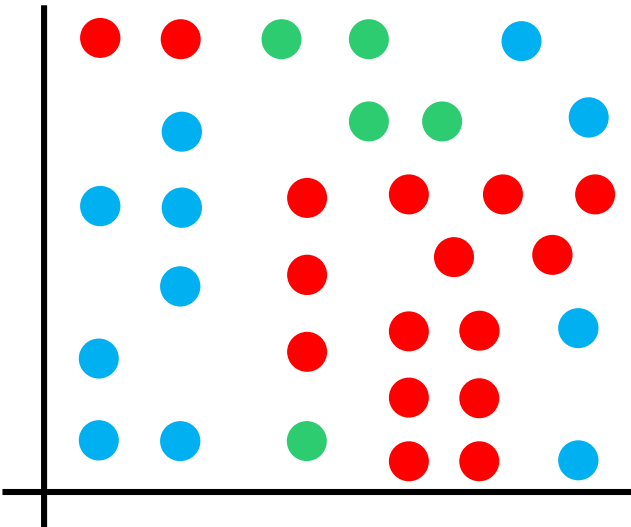
Leaf Action

- Can take any (complicated) action at a leaf
 - Why not call another machine learning algorithm?
 - Useful trick to keep in mind
- For speed, keep leaf action simple
 - Simplest action – constant



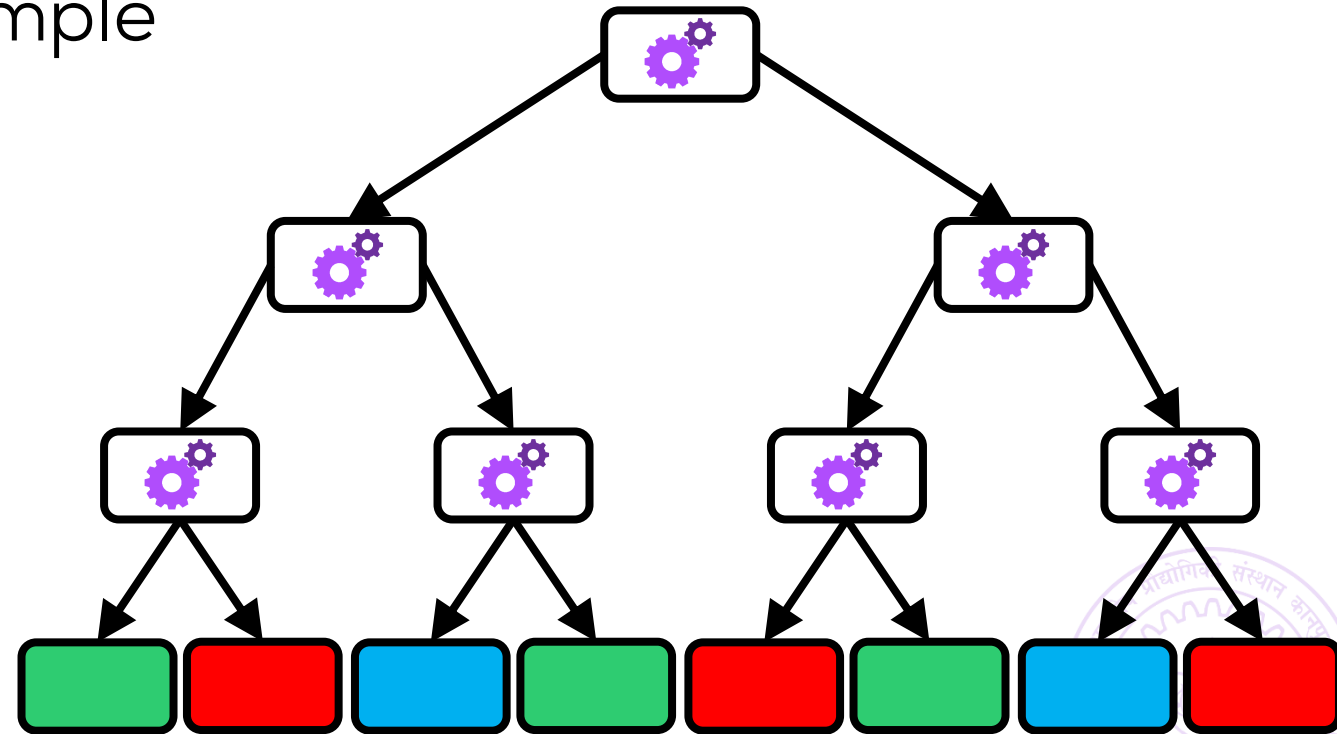
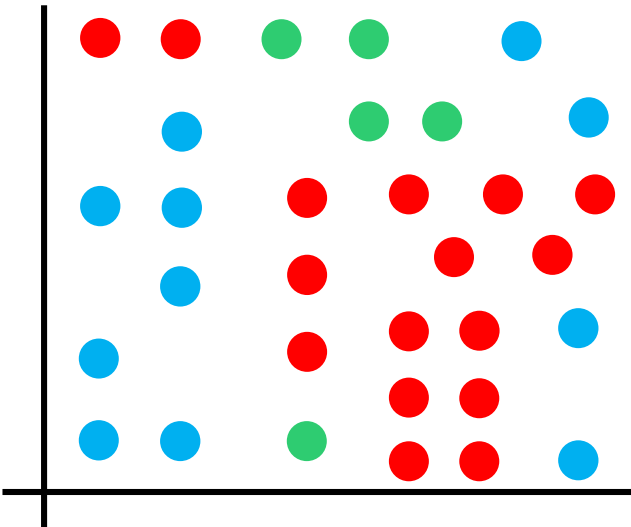
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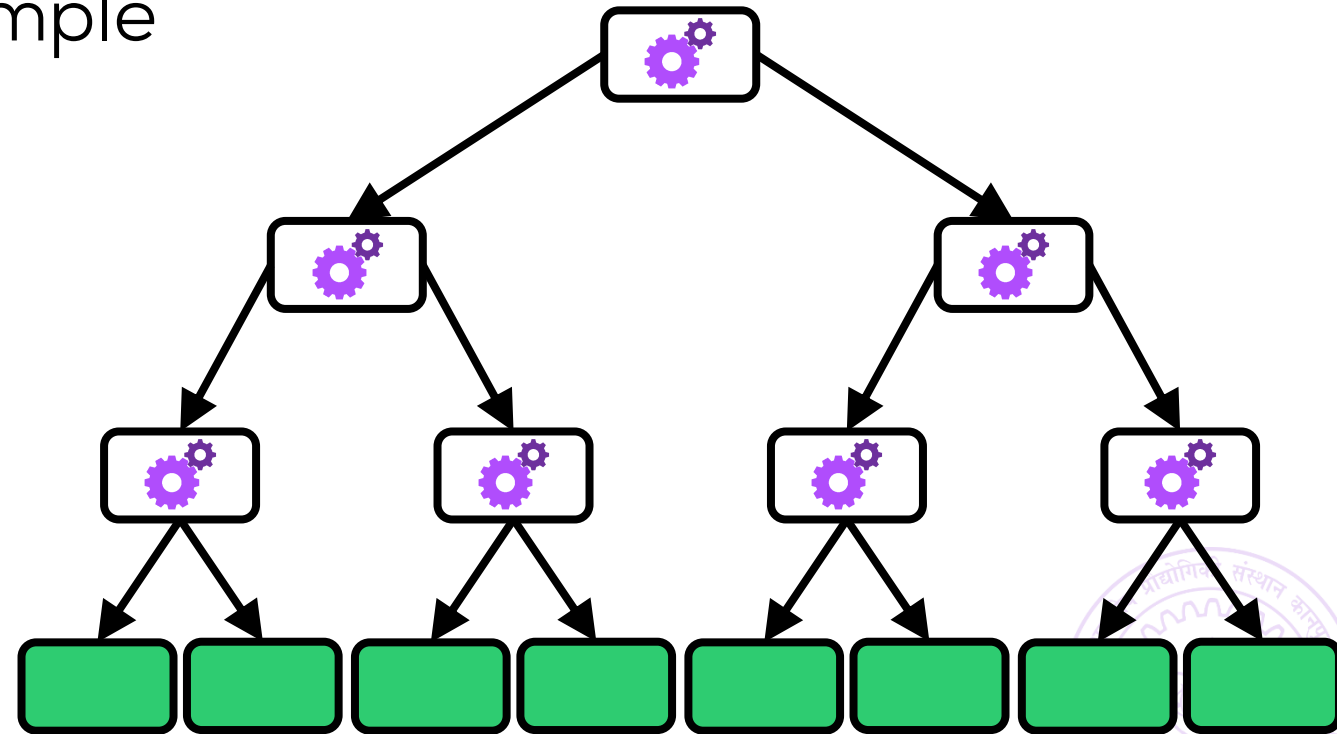
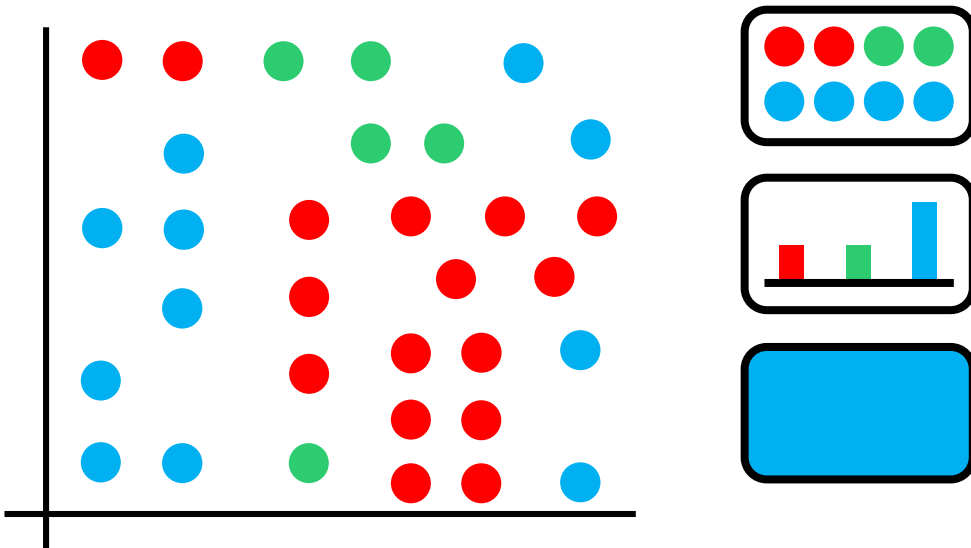
Leaf Action

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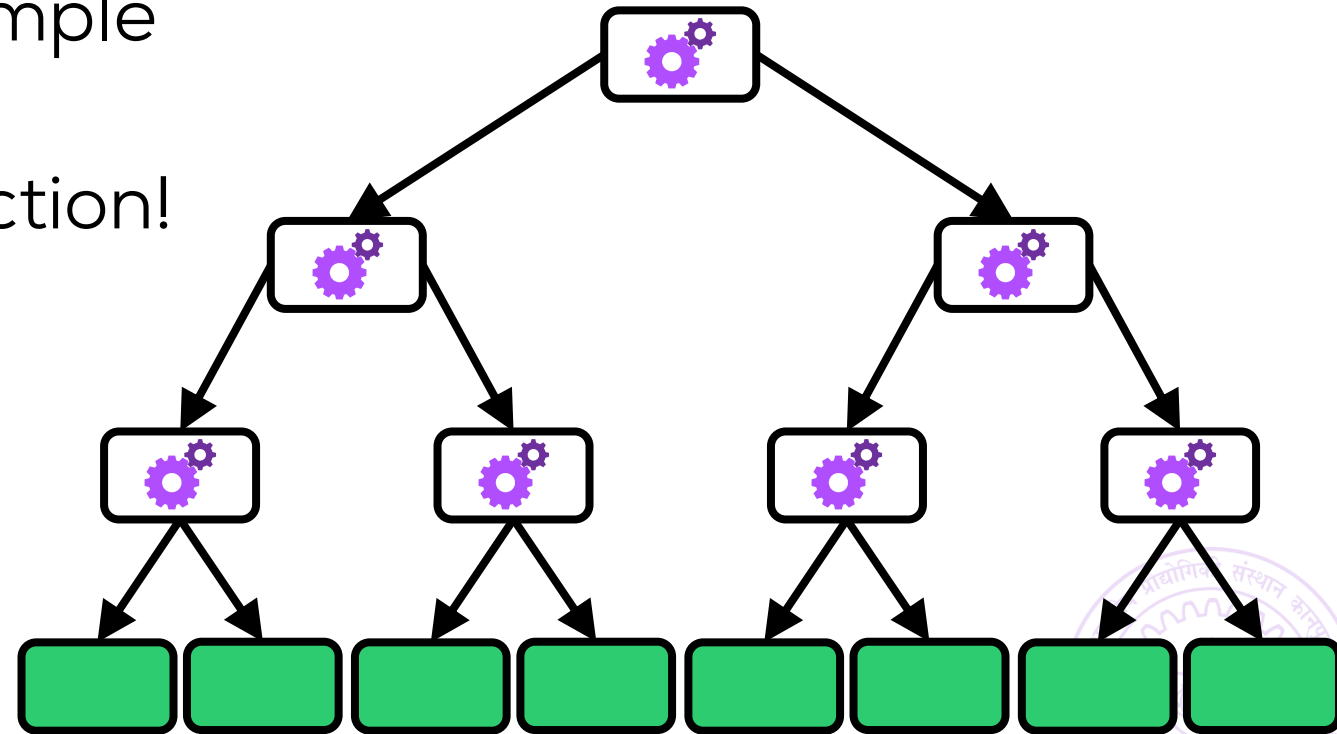
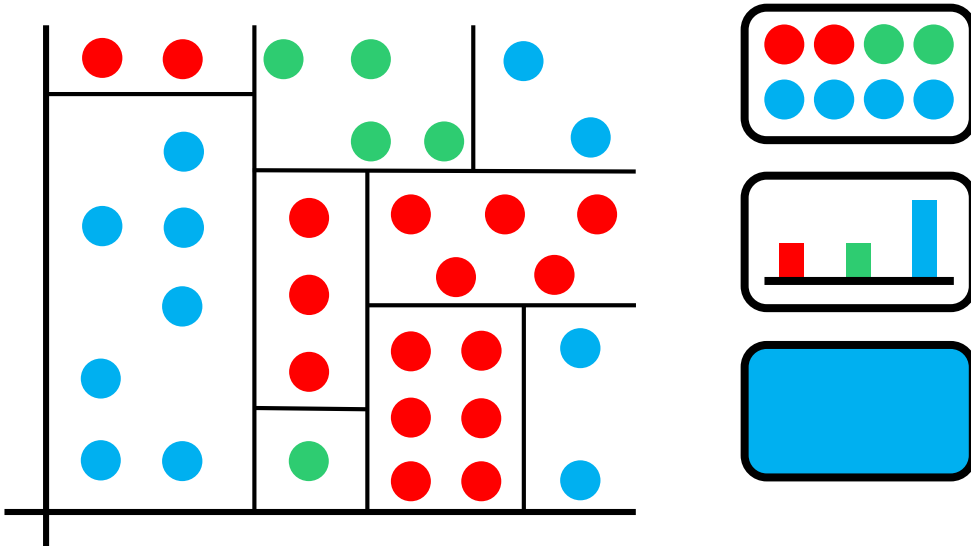
Leaf Action

- Can take any (complicated) action at a leaf
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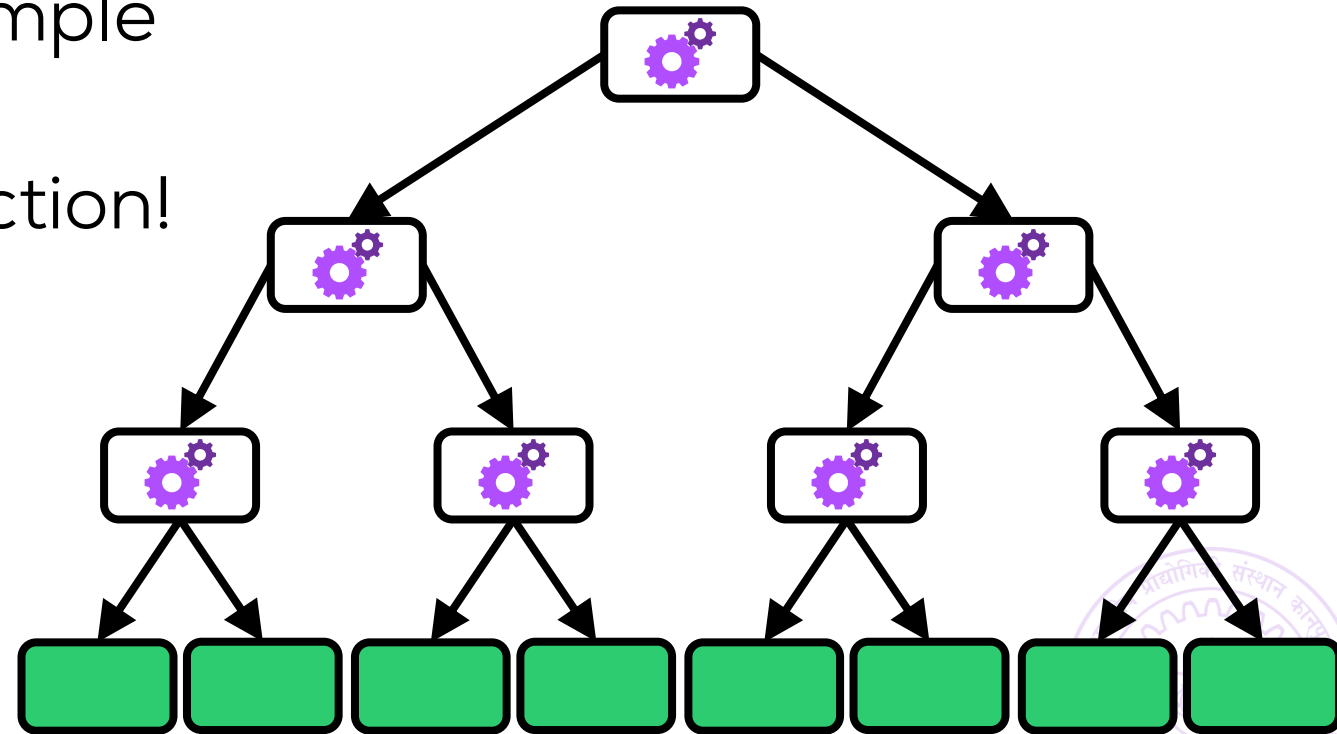
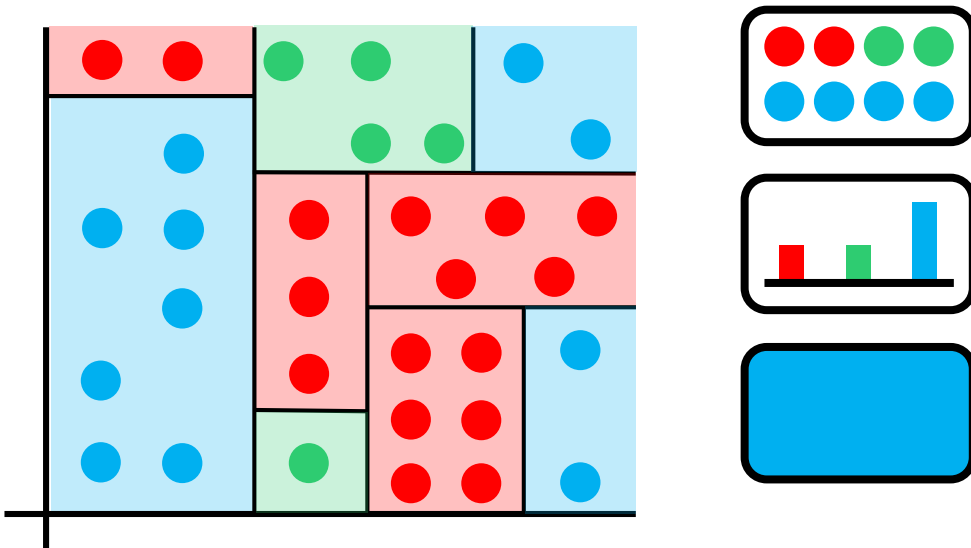
Leaf Action

- Can take any (complicated) action at a leaf
 - Why not call another machine learning algorithm?
 - Useful trick to keep in mind
- For speed, keep leaf action simple
 - Simplest action – constant
 - DT piecewise constant function!

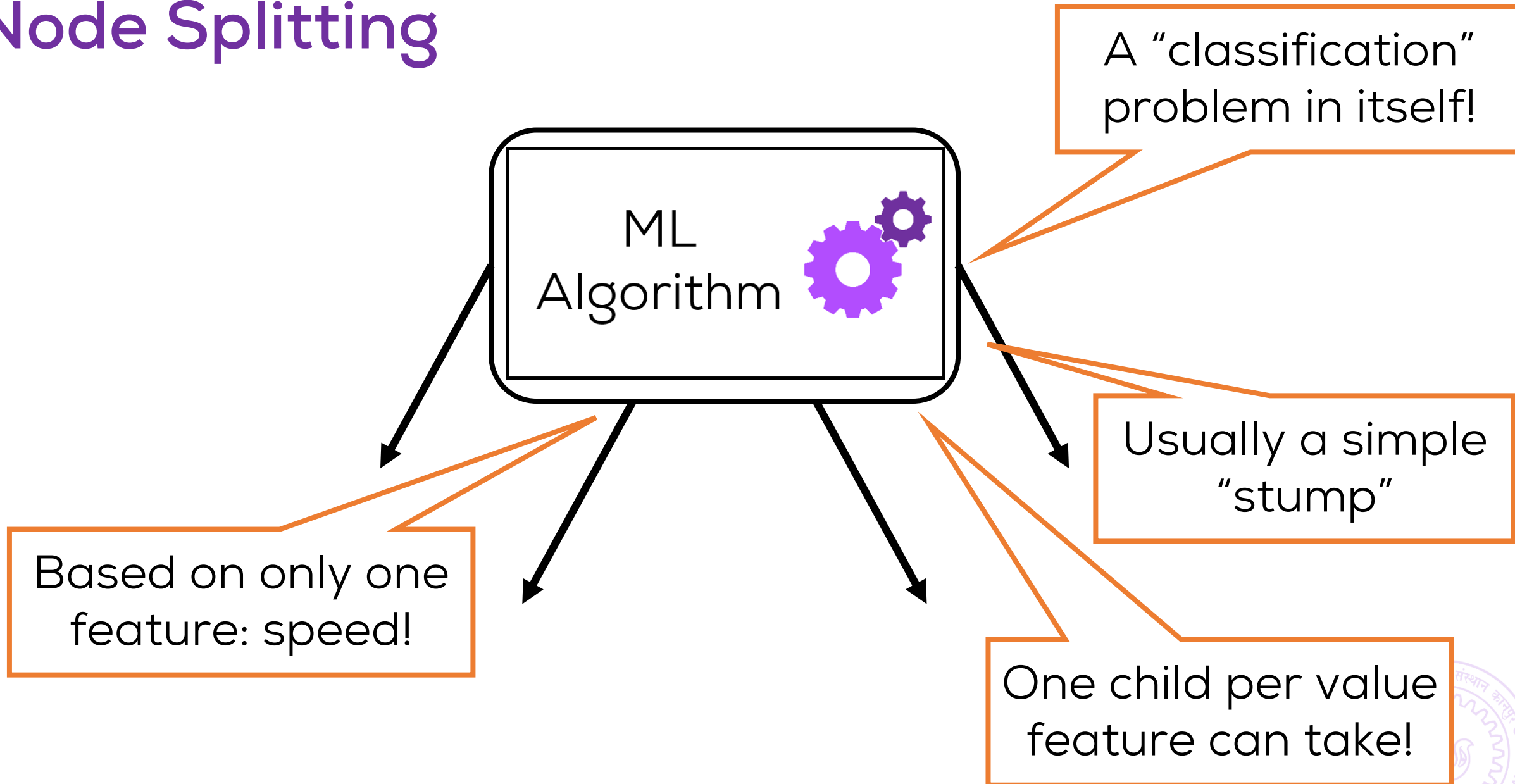


Leaf Action

- Can take any (complicated) action at a leaf
 - Why not call another machine learning algorithm?
 - Useful trick to keep in mind
- For speed, keep leaf action simple
 - Simplest action – constant
 - DT piecewise constant function!



Node Splitting



Node Splitting



A "classification" problem in itself!

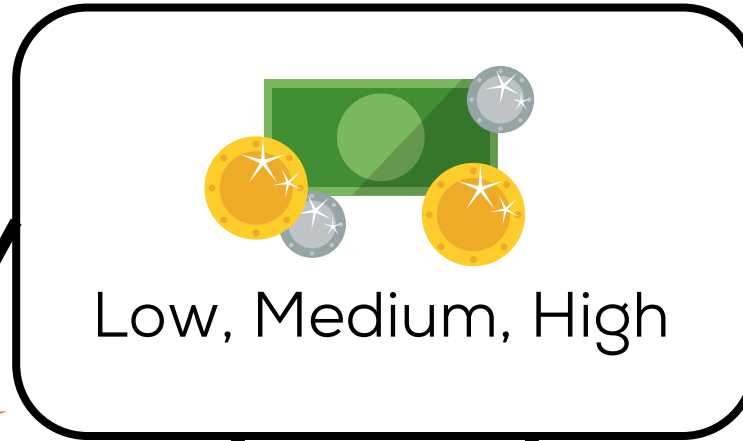
Discrete features easy to handle

Based on only one feature: speed!

Usually a simple "stump"

One child per value feature can take!

Node Splitting



A “classification”
problem in itself!

Discrete features
easy to handle

Usually a simple
“stump”

Based on only one
feature: speed!

How finely?

Continuous features
discretized

One child per value
feature can take!

Node Splitting

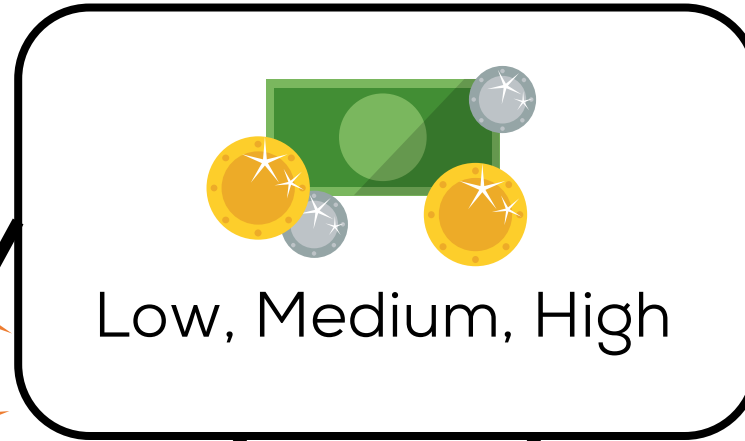
Do “multi-label” splits make sense?

Discrete features easy to handle

Based on only one feature: speed!

How finely?

Continuous features discretized

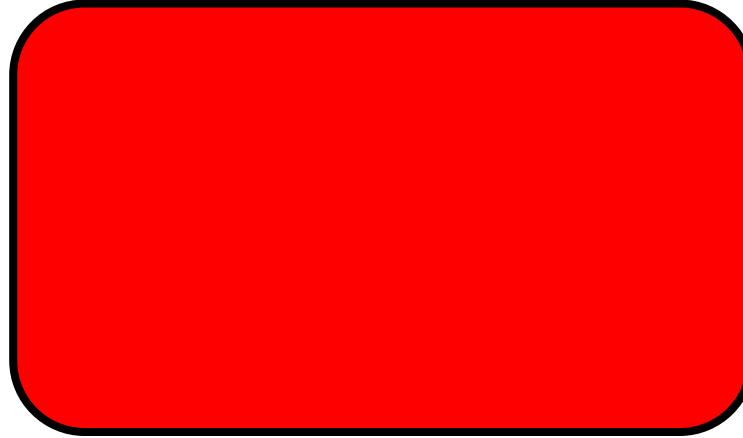


A “classification” problem in itself!

Usually a simple “stump”

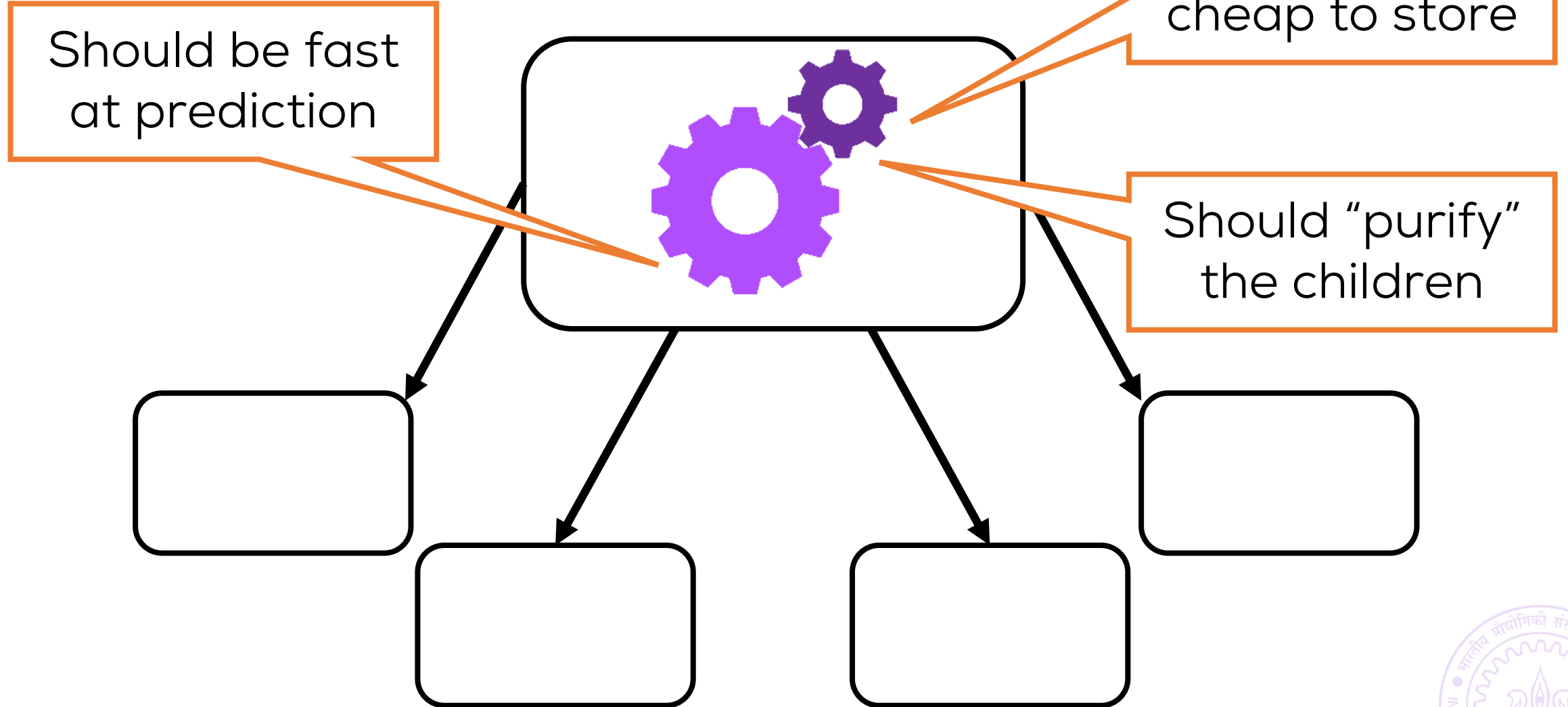
One child per value feature can take!

Node Splitting

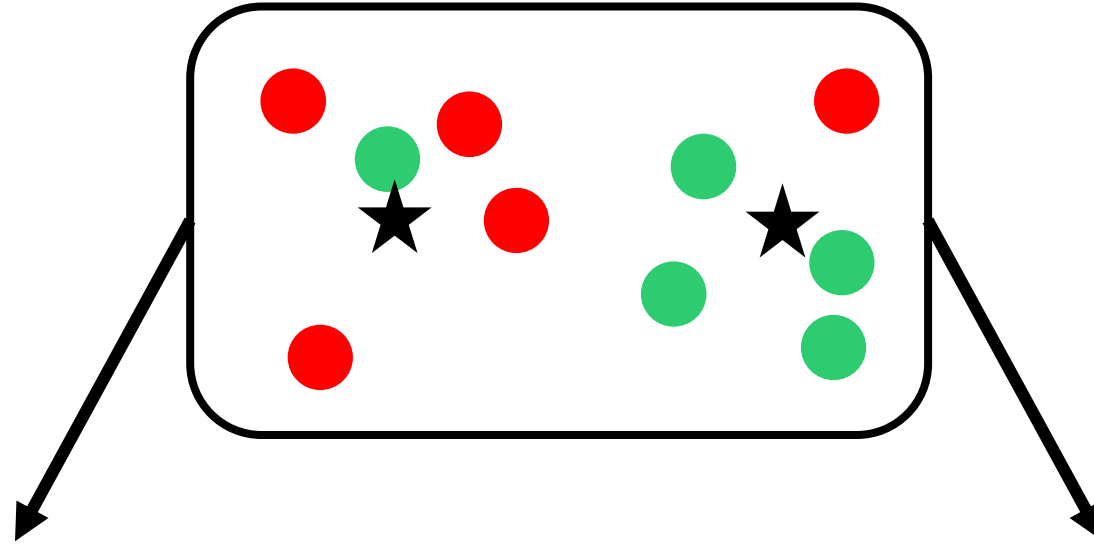


A pure node needs
no children

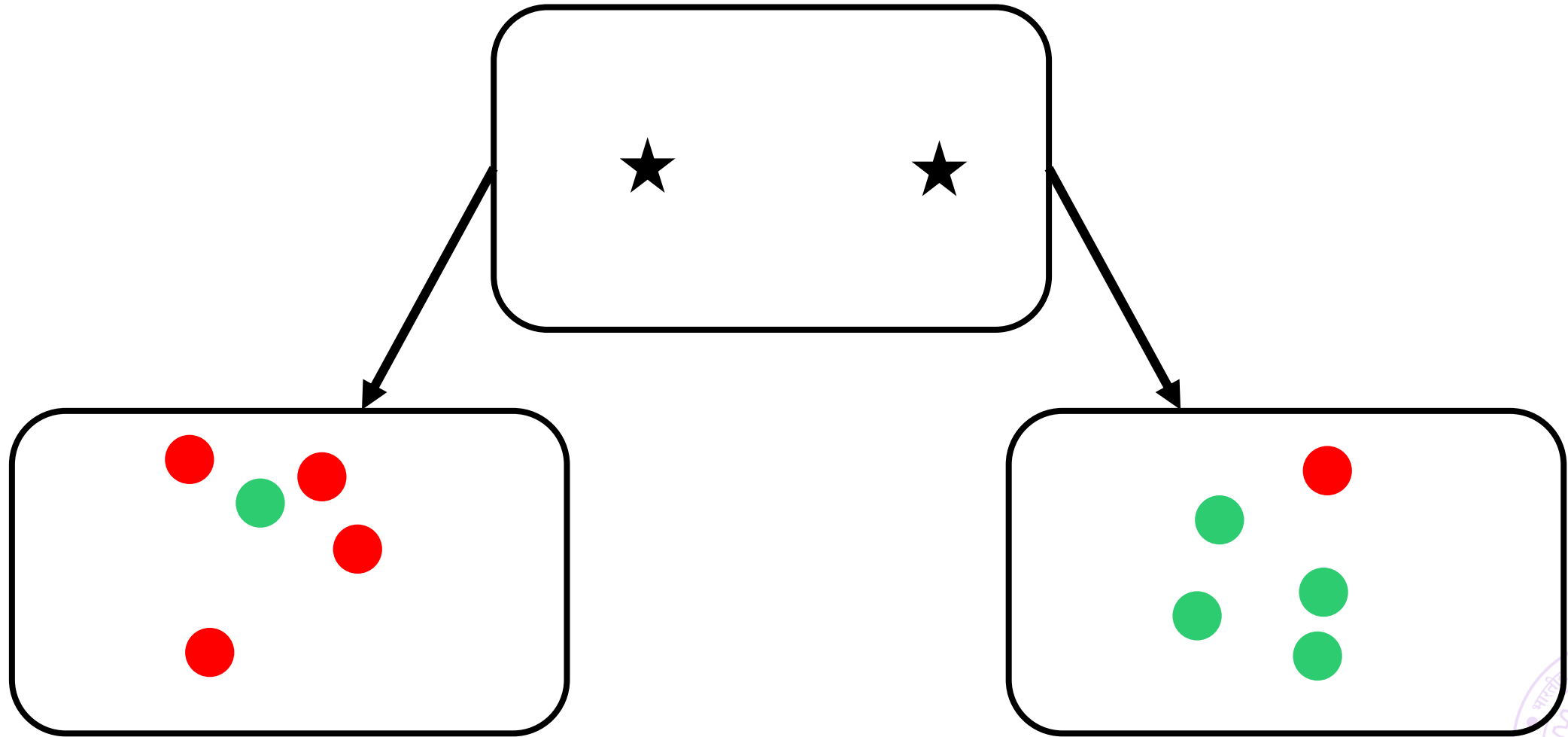
Node Splitting



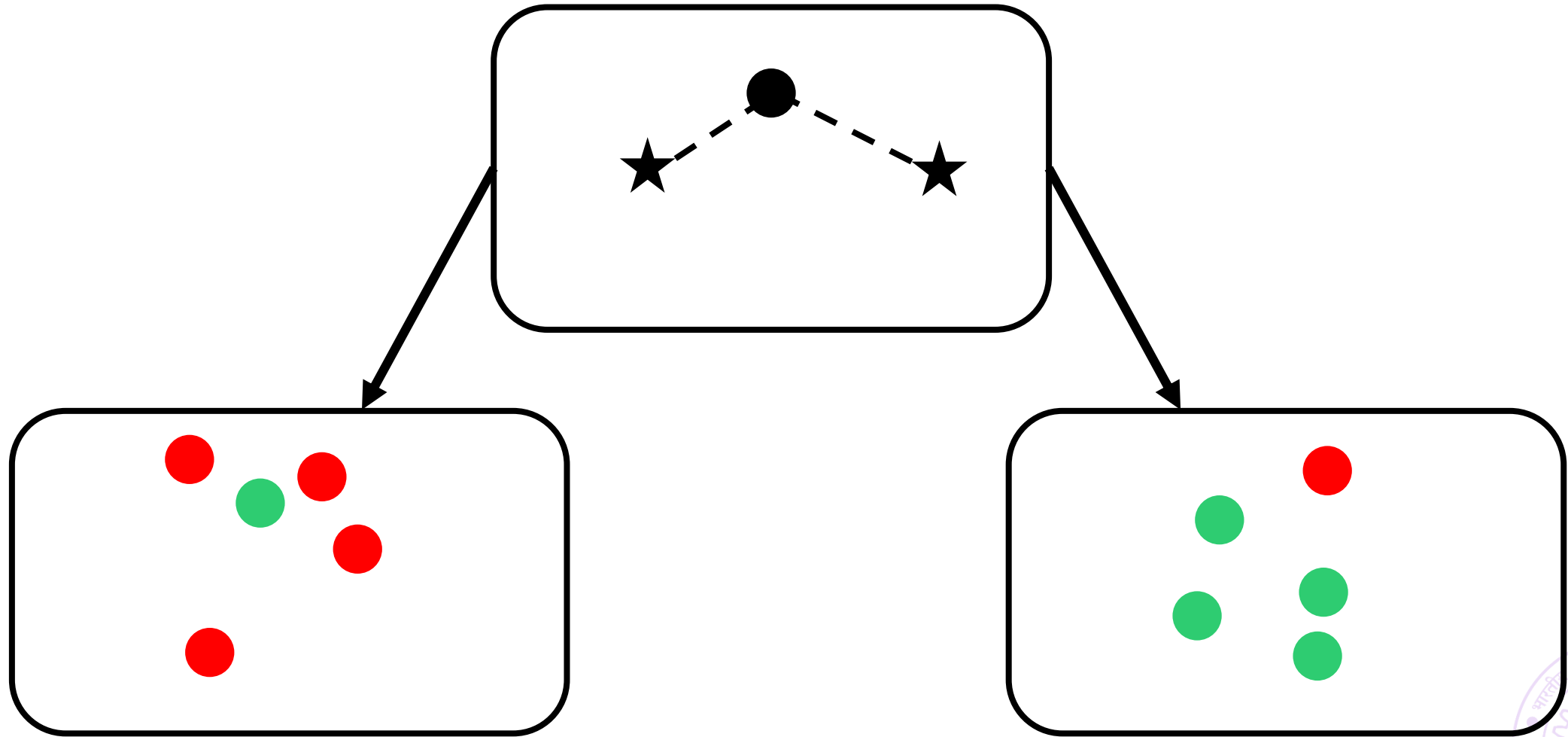
Node Splitting via prototypes (clustering)



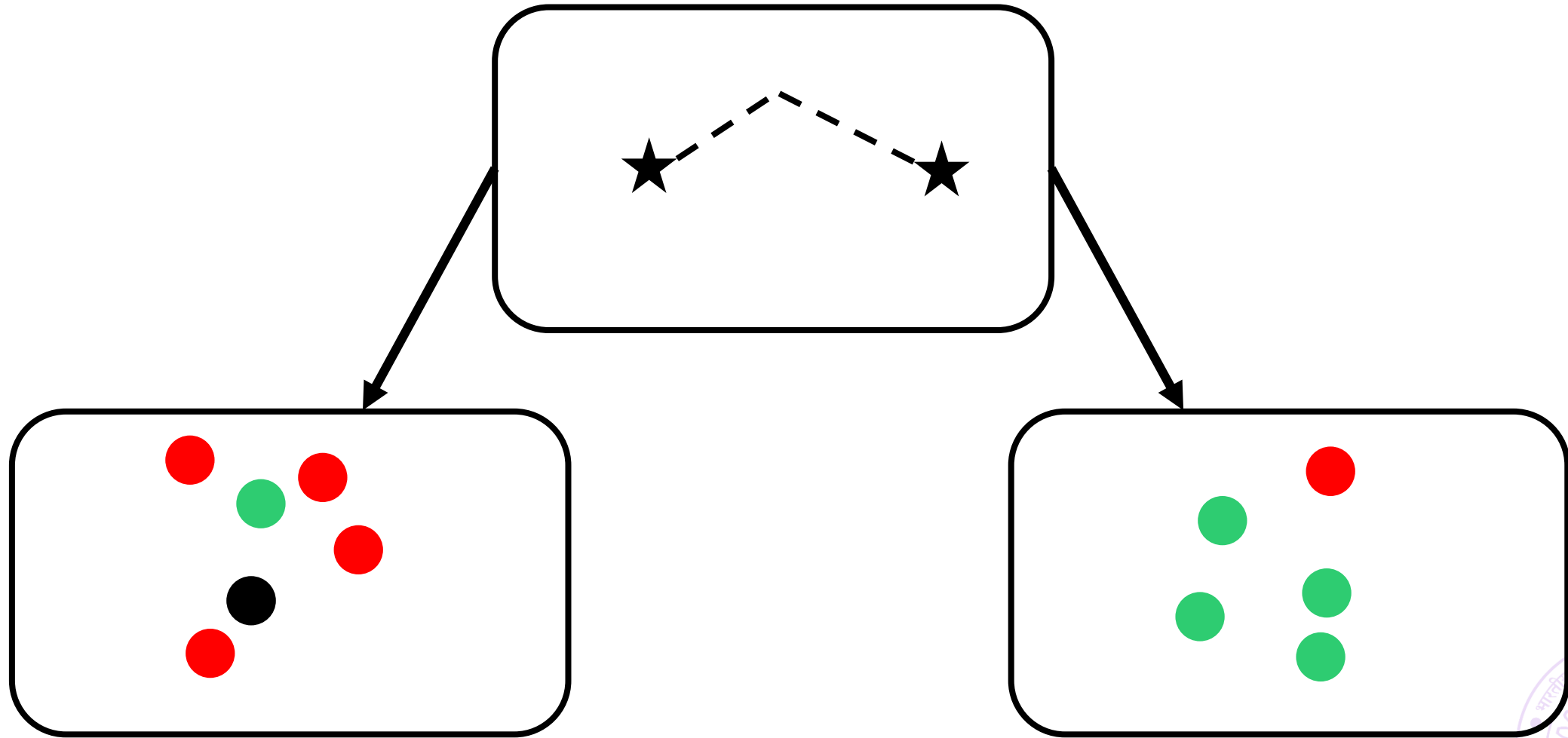
Node Splitting via prototypes (clustering)



Node Splitting via prototypes (clustering)

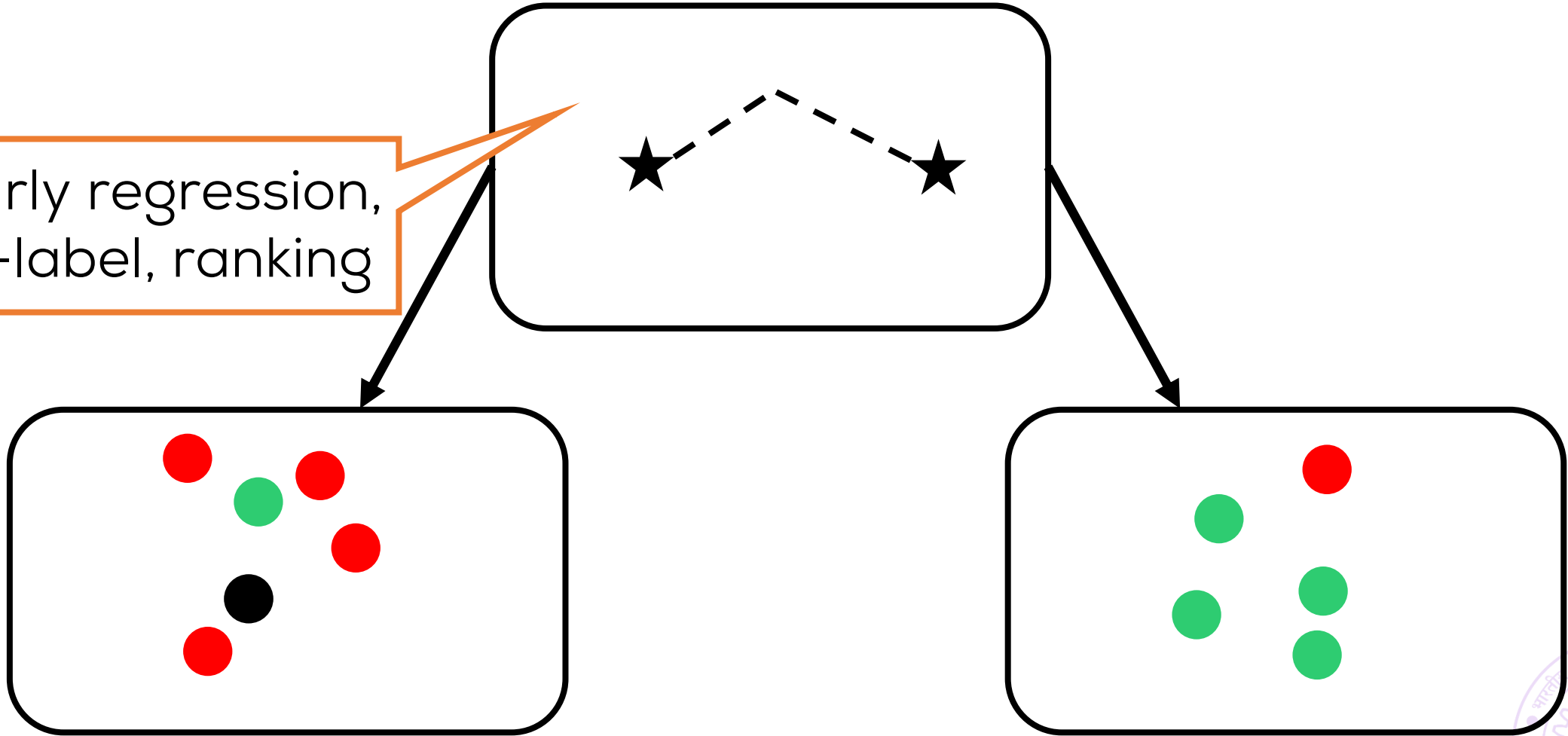


Node Splitting via prototypes (clustering)

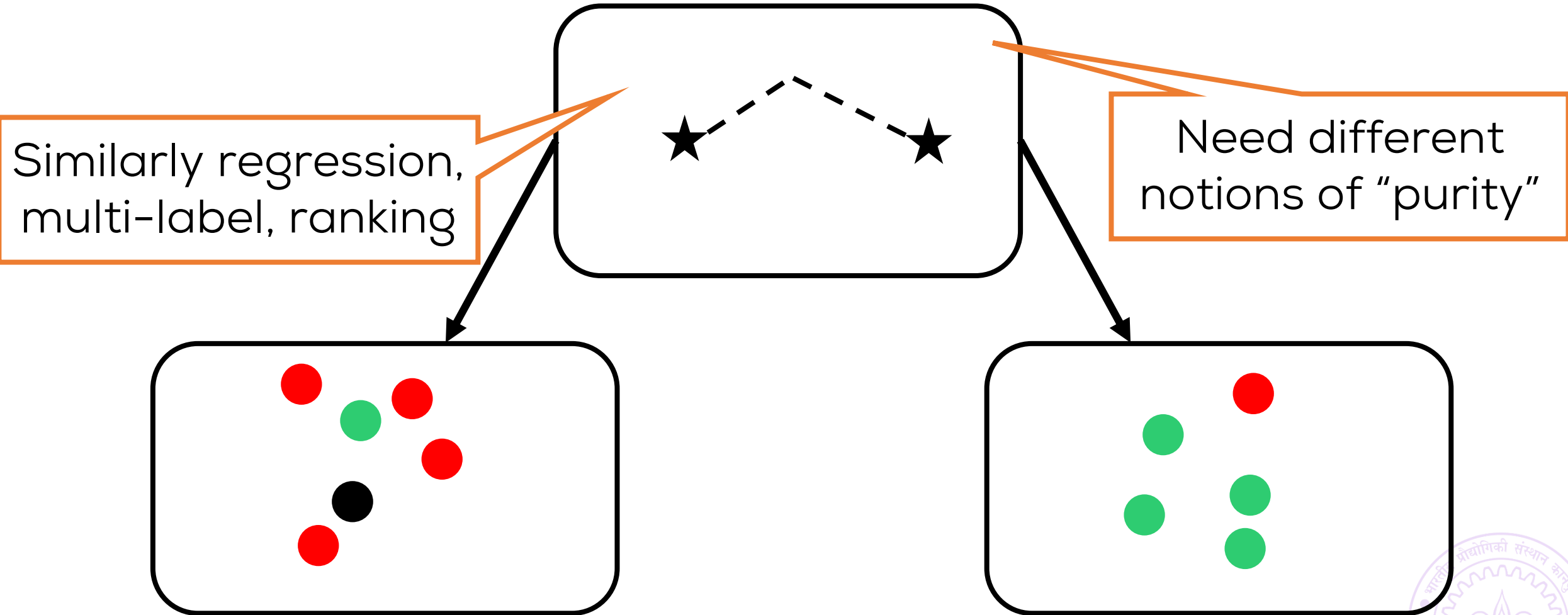


Node Splitting via prototypes (clustering)

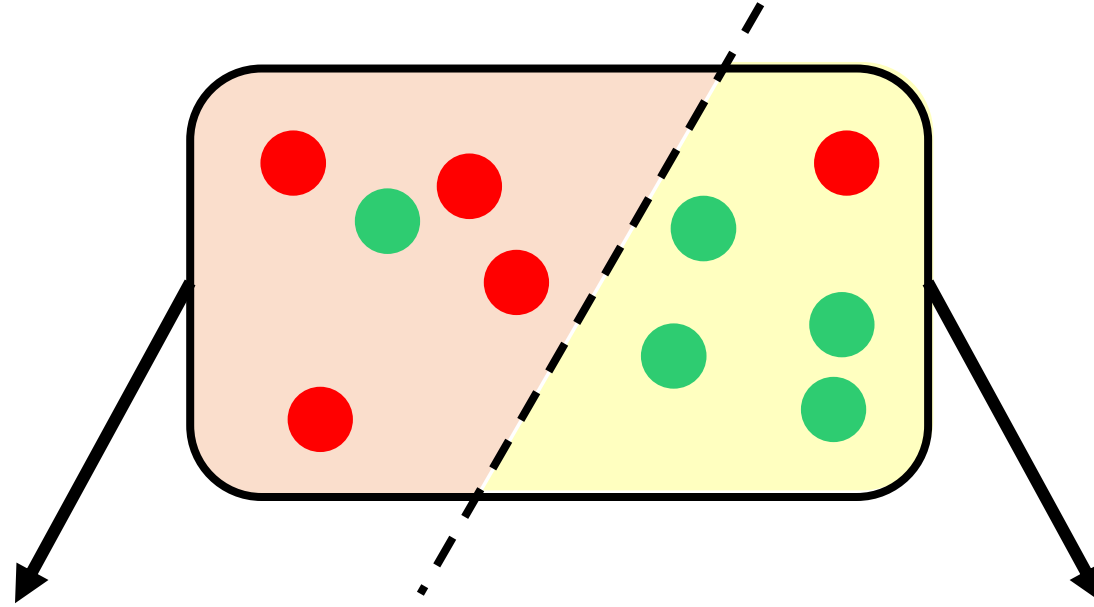
Similarly regression,
multi-label, ranking



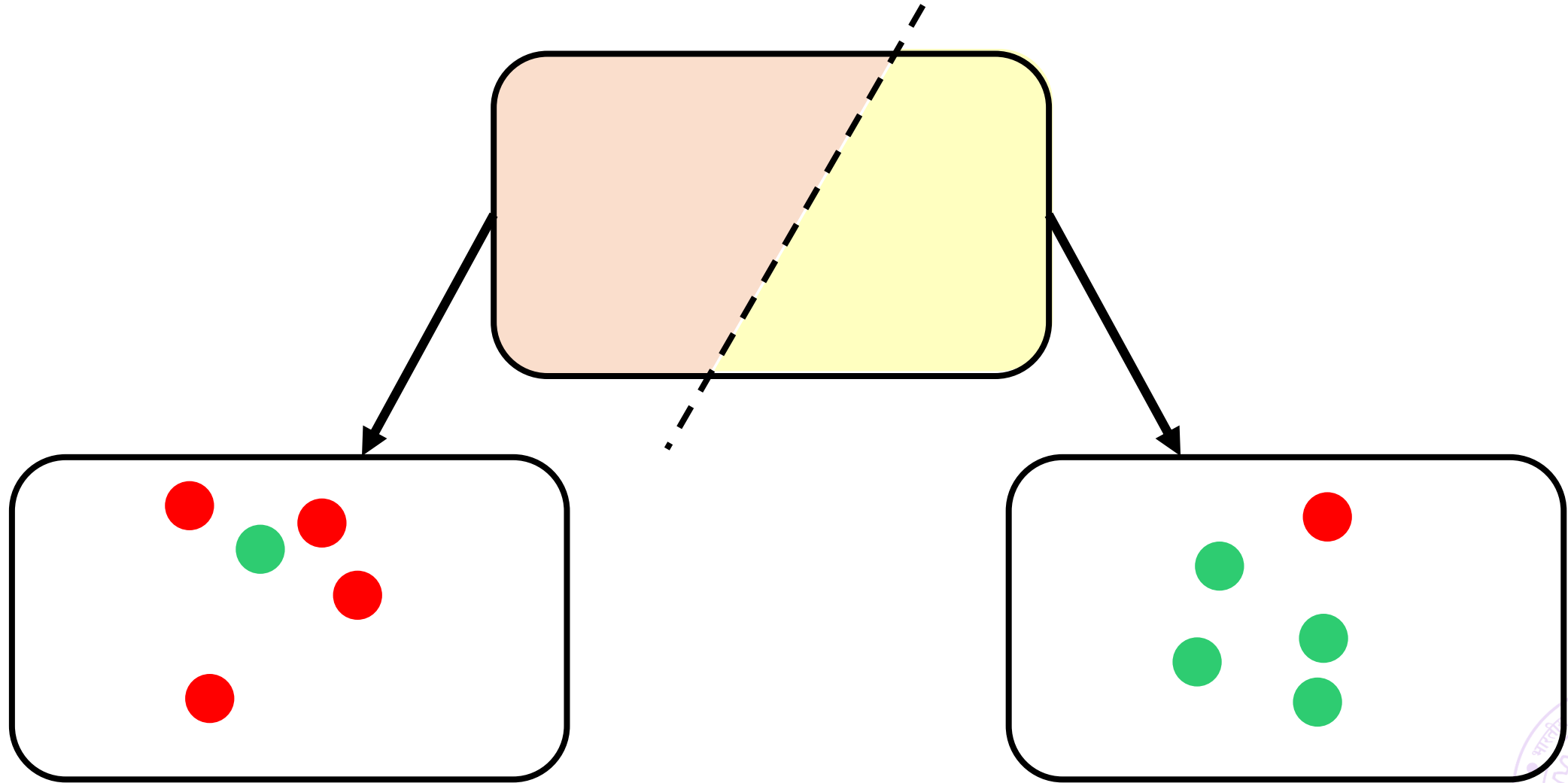
Node Splitting via prototypes (clustering)



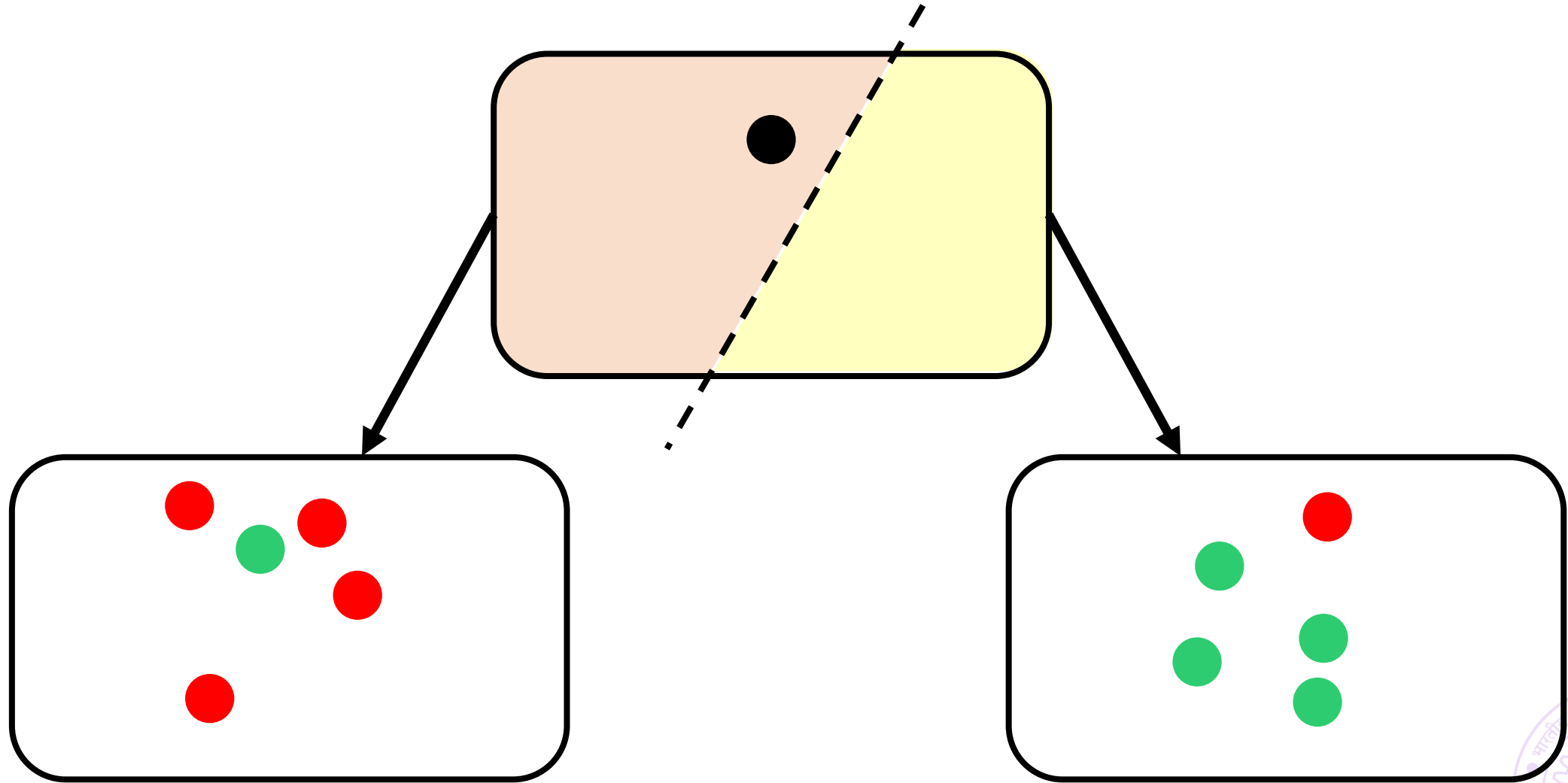
Node Splitting via linear classifiers



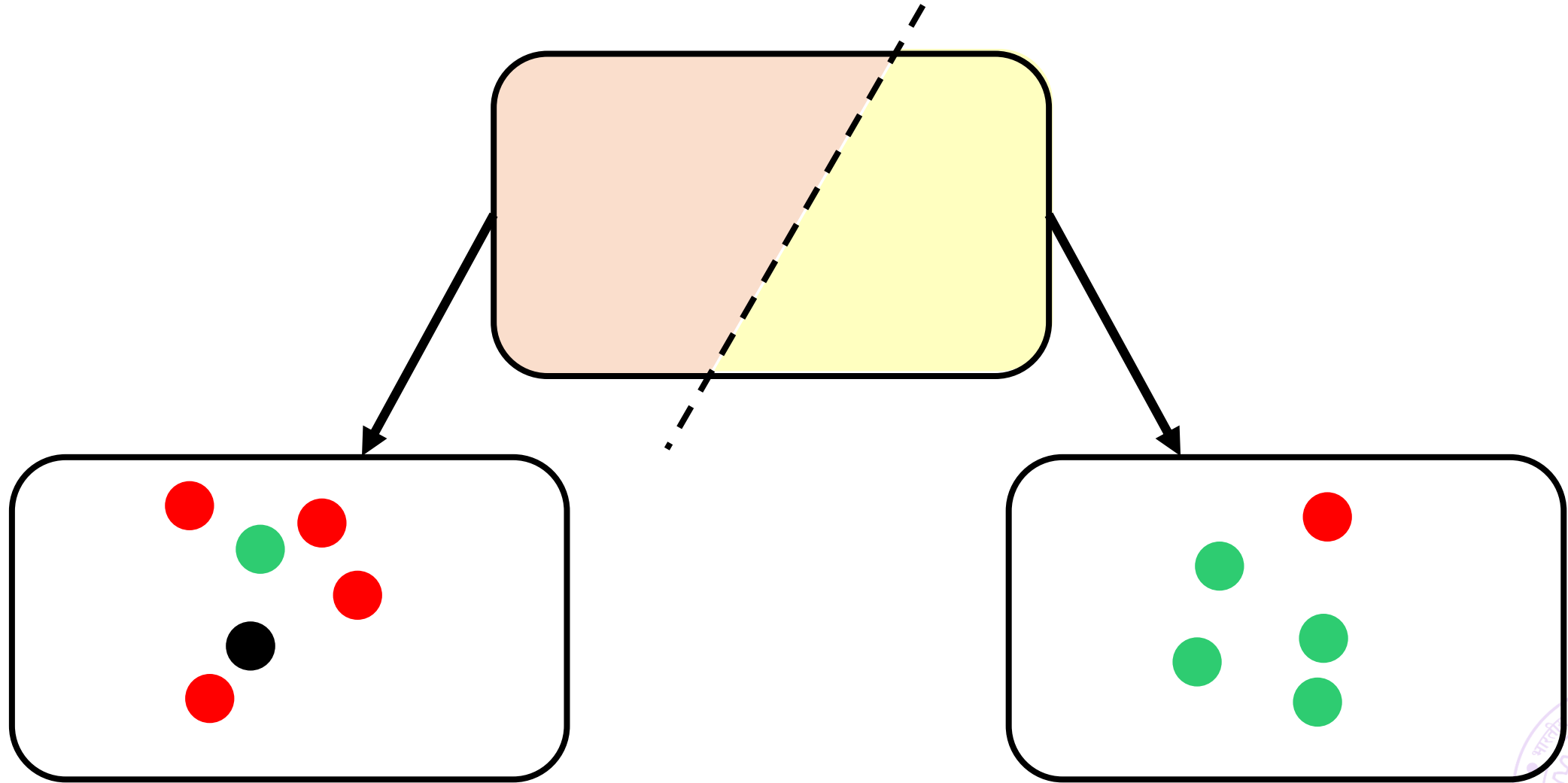
Node Splitting via linear classifiers



Node Splitting via linear classifiers

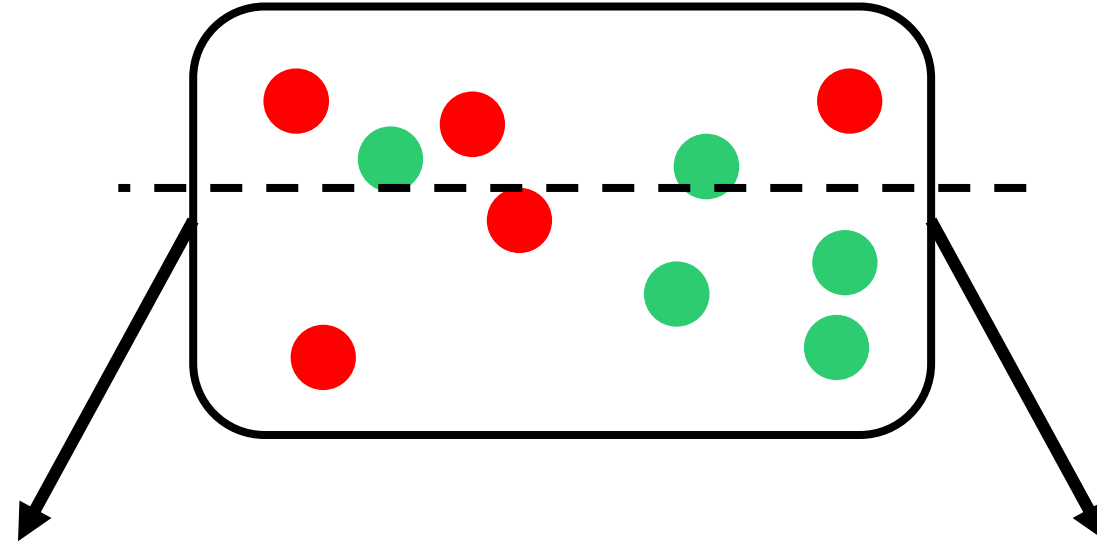
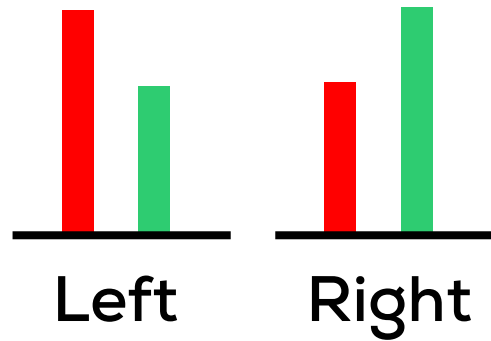


Node Splitting via linear classifiers



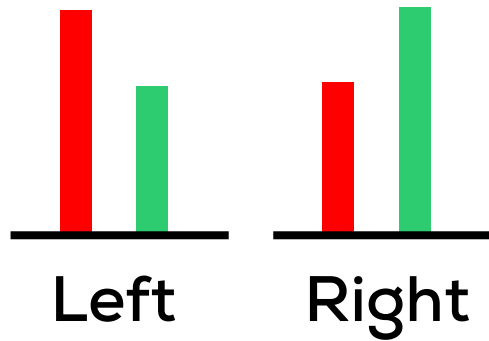
Node Splitting via feature stumps

Horizontal Split

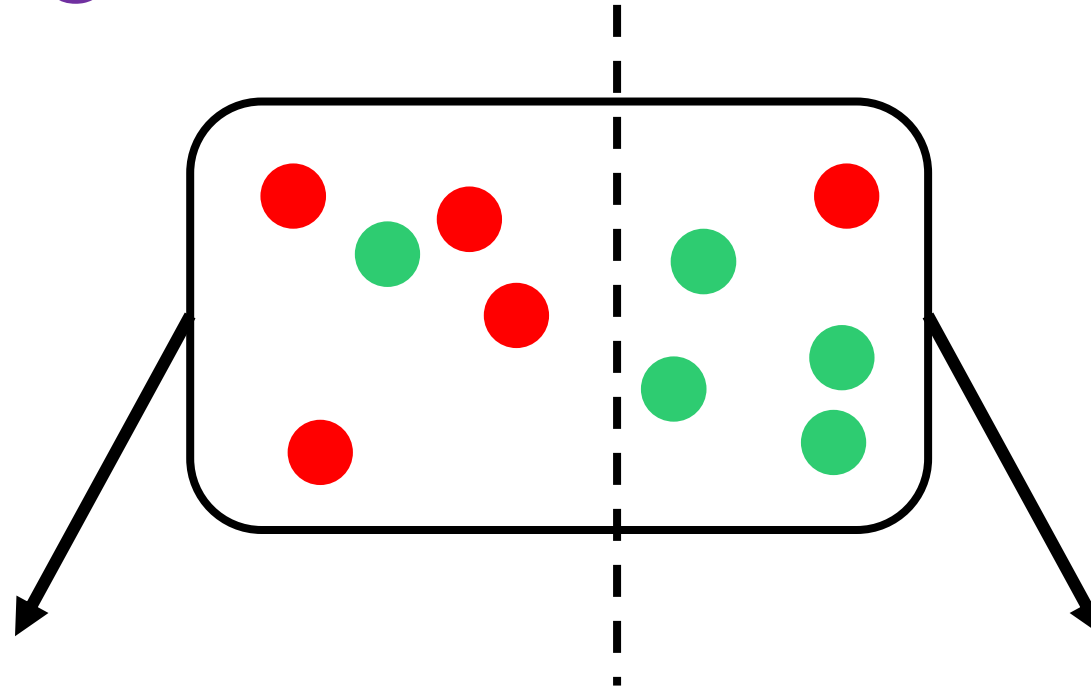
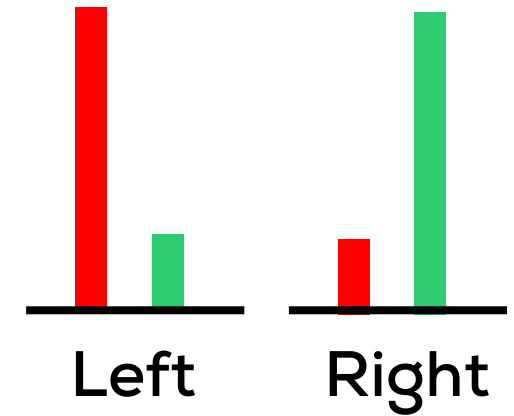


Node Splitting via feature stumps

Horizontal Split

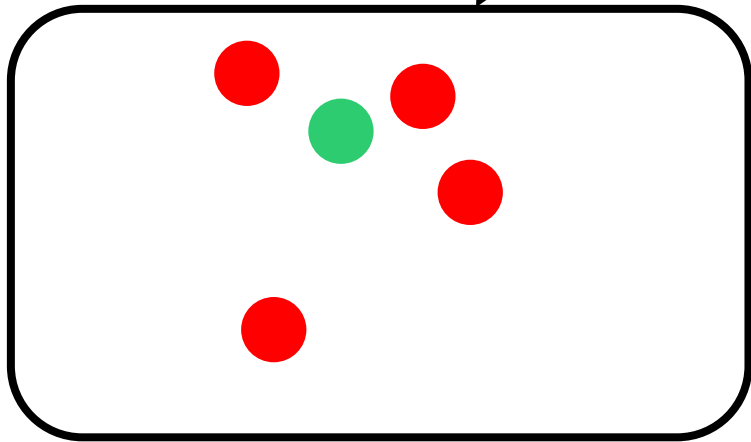
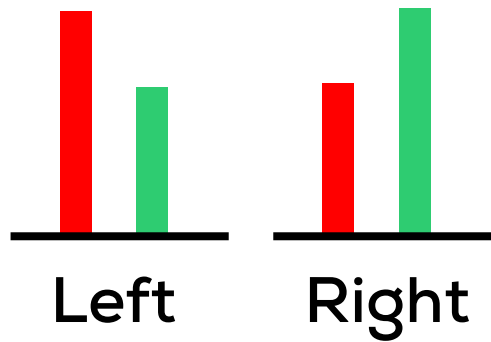


Vertical Split

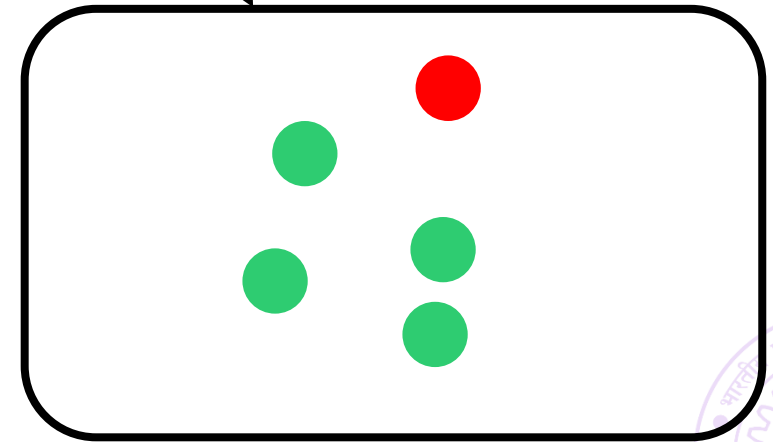
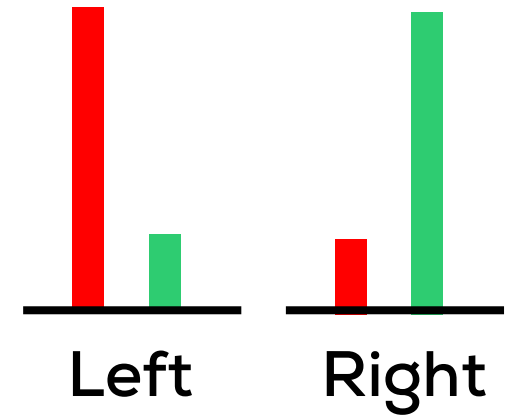


Node Splitting via feature stumps

Horizontal Split

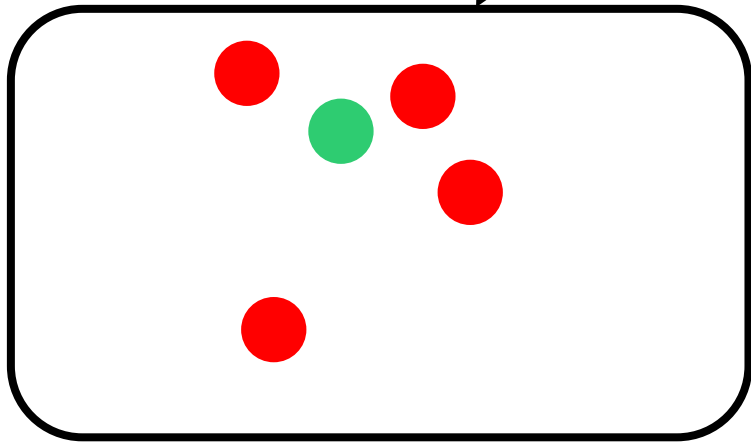
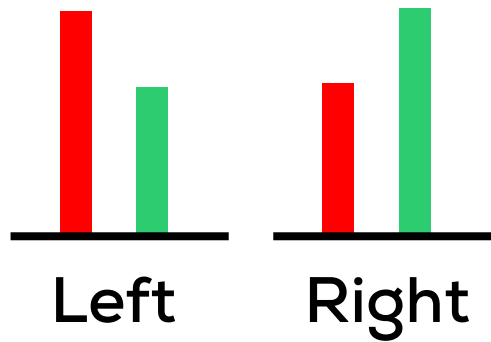


Vertical Split

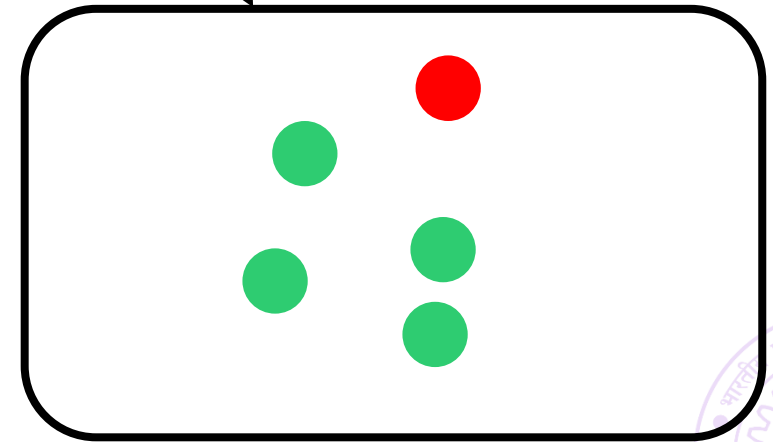
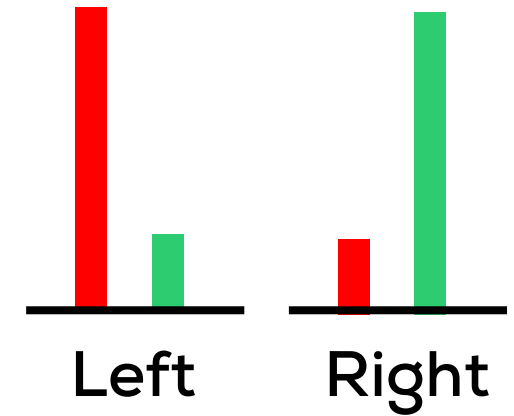


Node Splitting via feature stumps

Horizontal Split

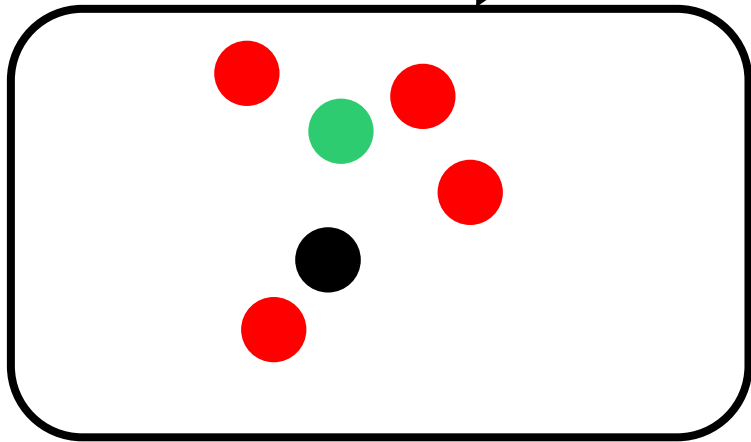
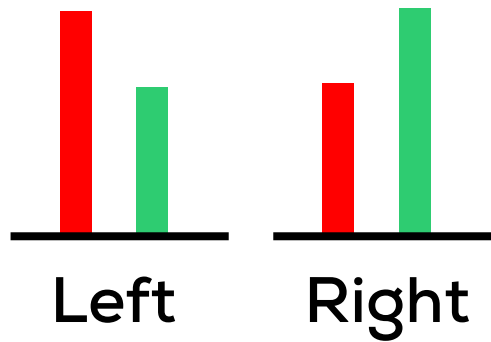


Vertical Split

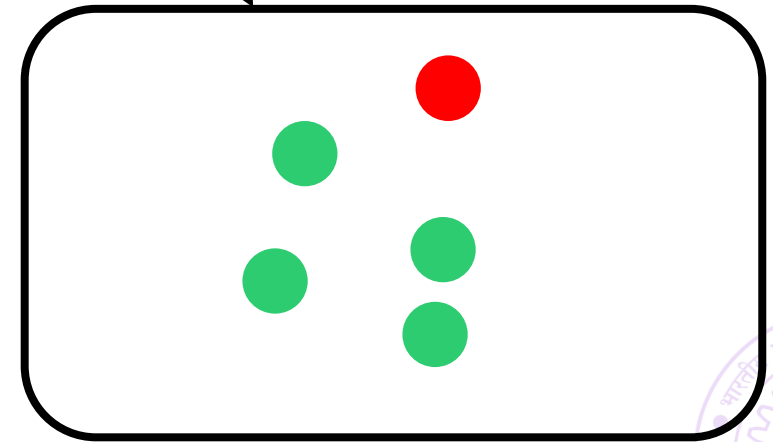
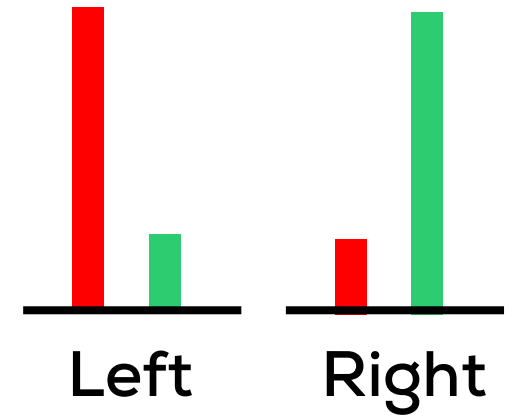


Node Splitting via feature stumps

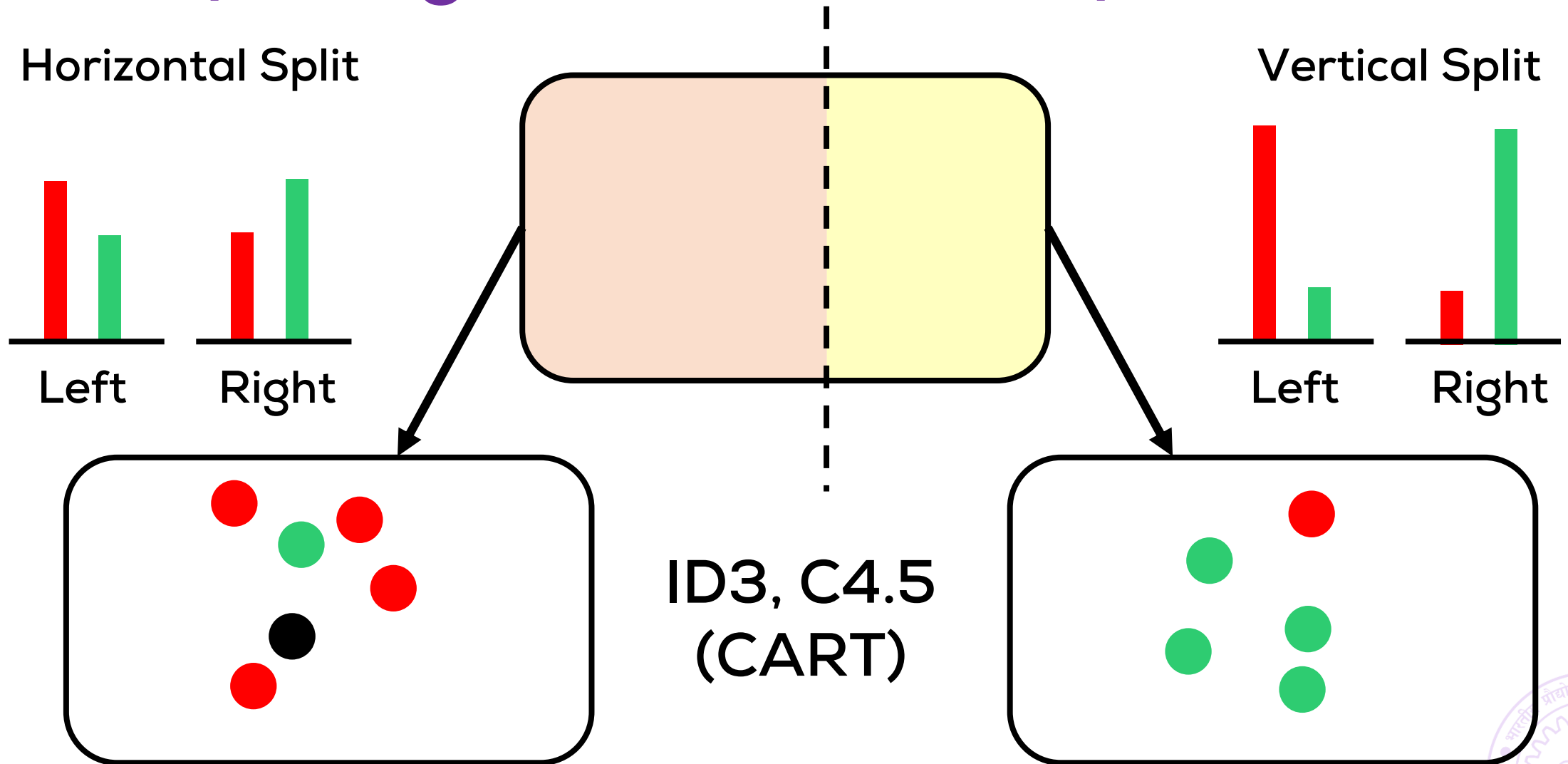
Horizontal Split



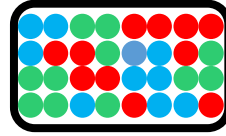
Vertical Split



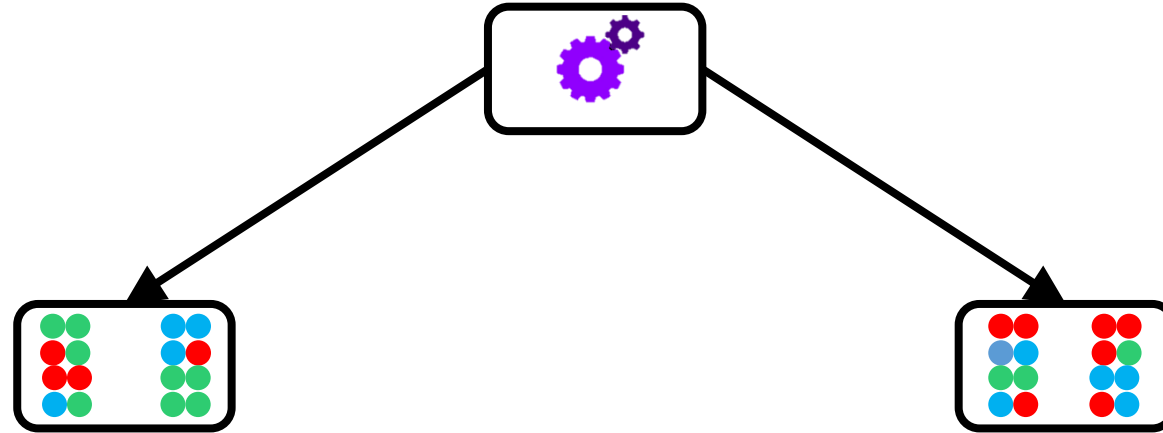
Node Splitting via feature stumps



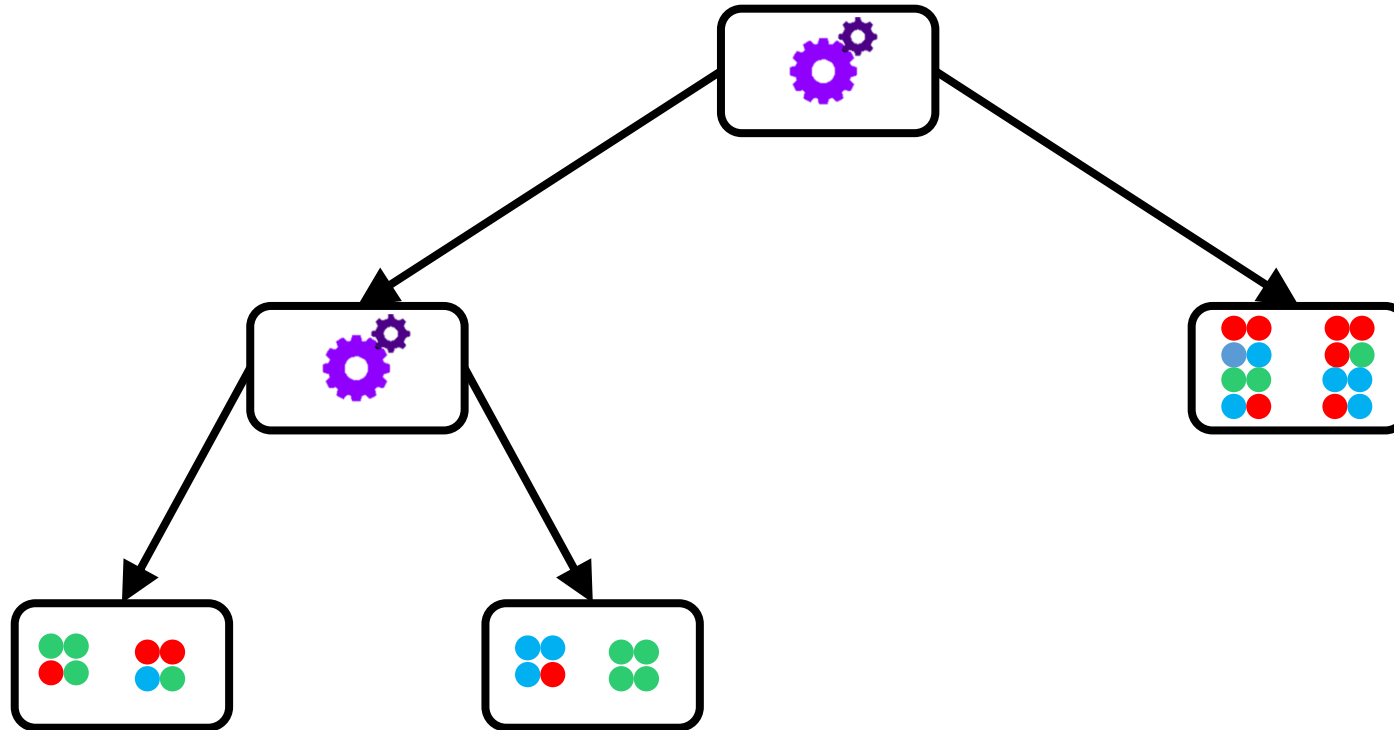
Growing the Tree



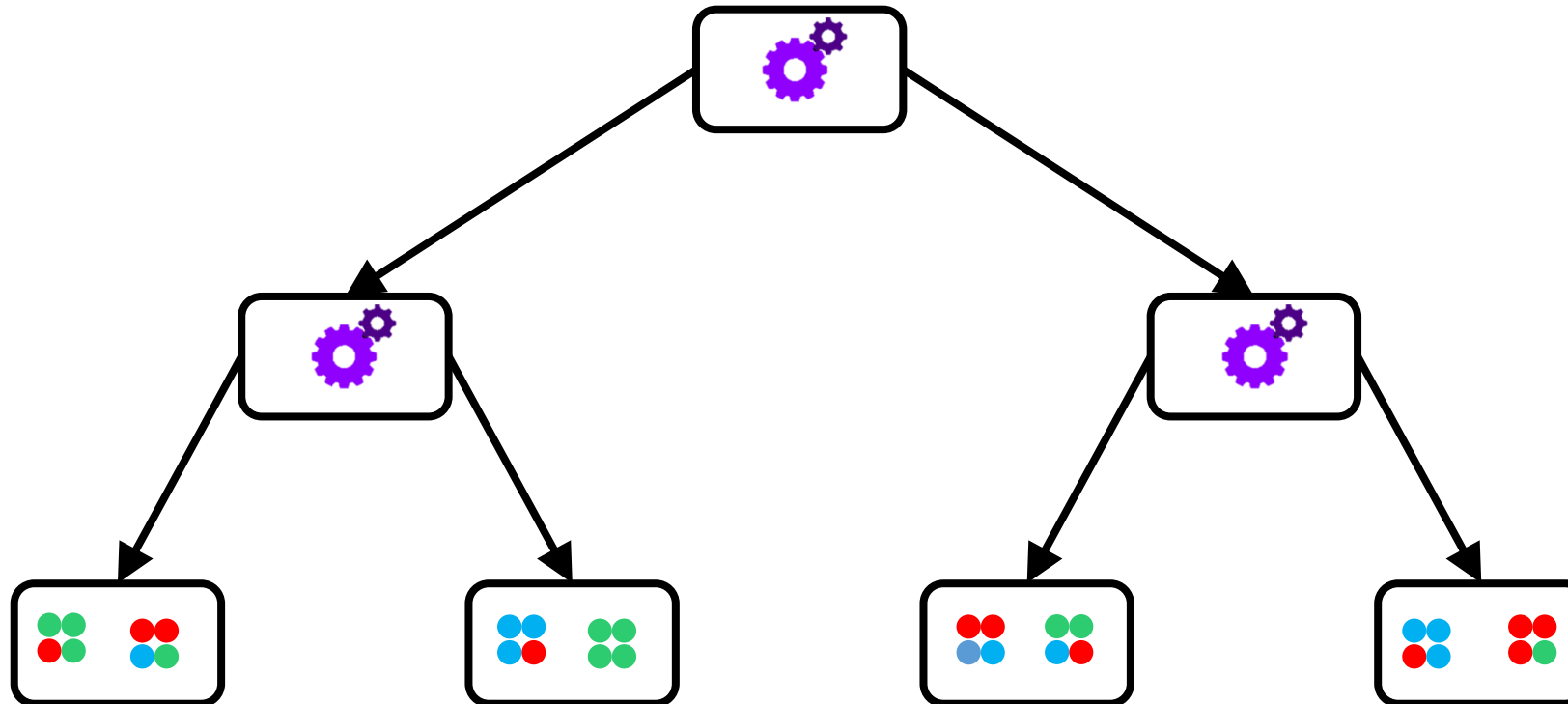
Growing the Tree



Growing the Tree

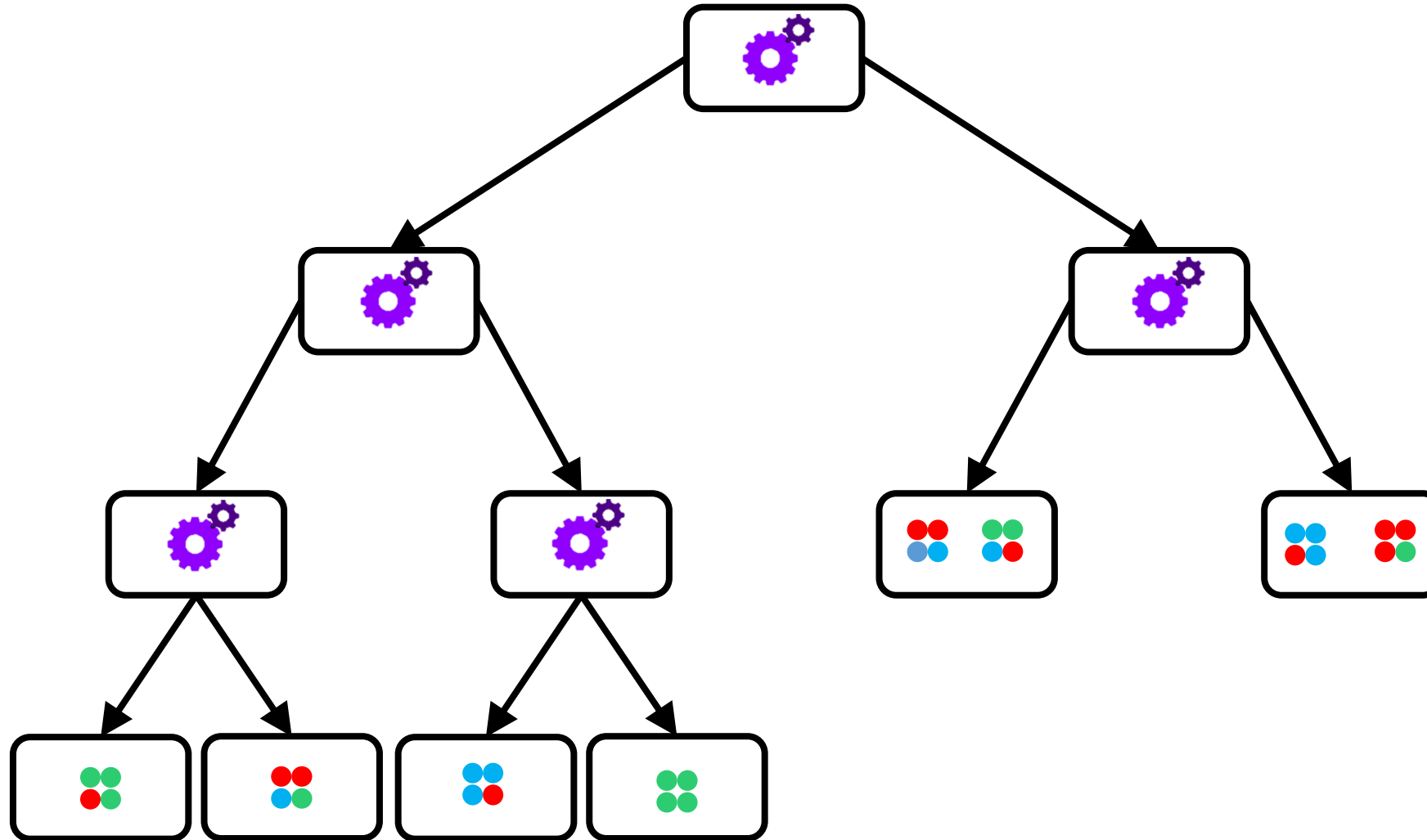


Growing the Tree

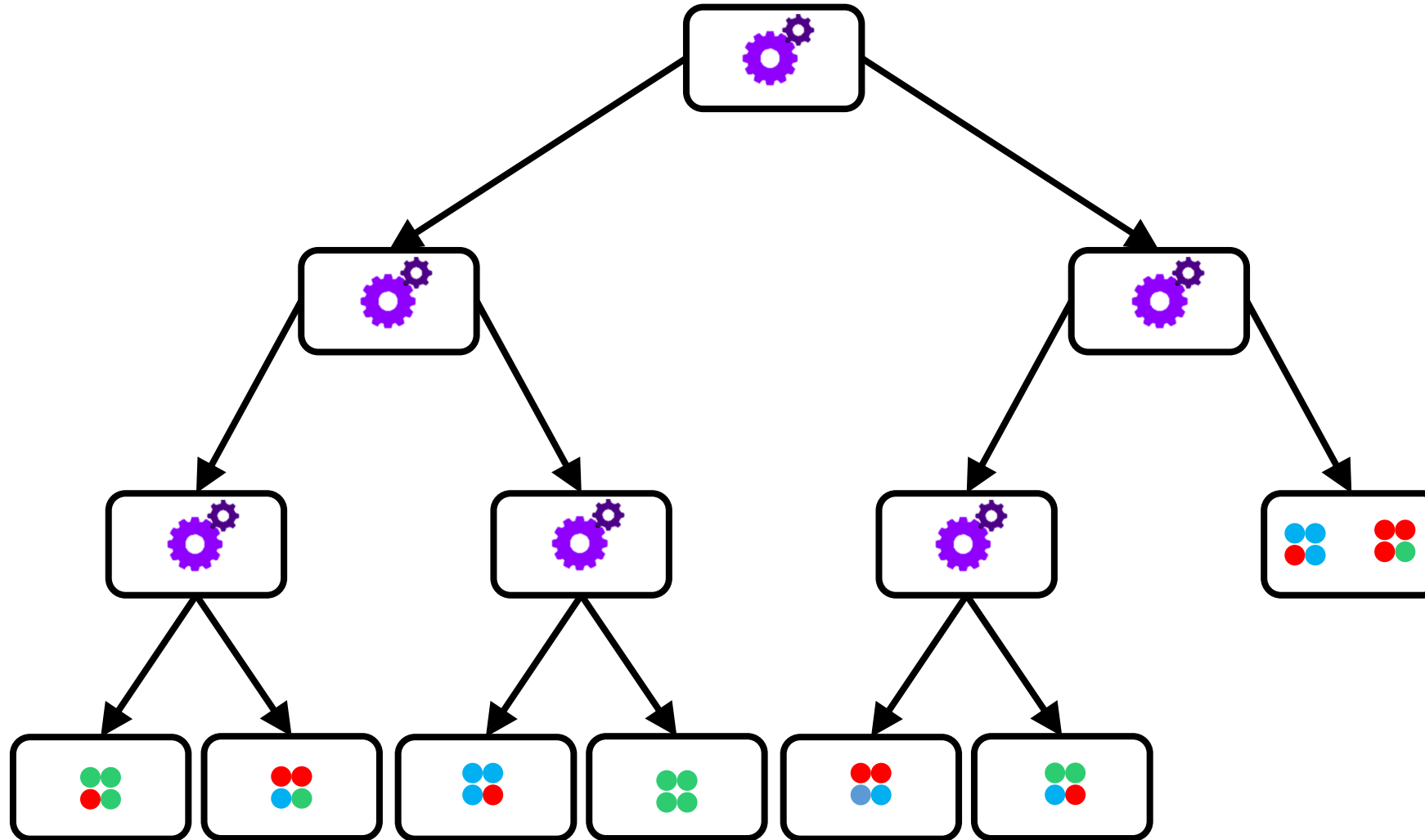


August 11, 2017

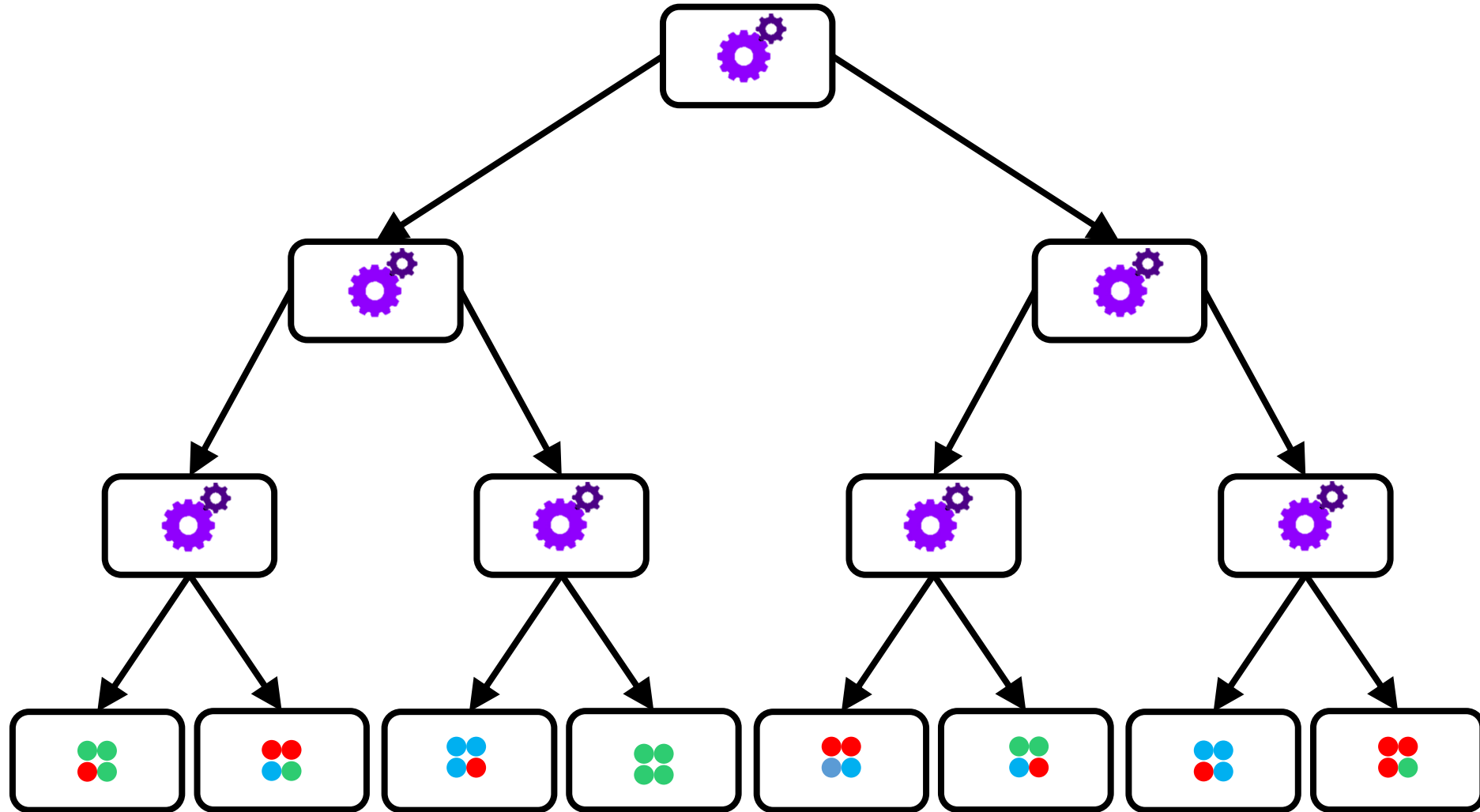
Growing the Tree



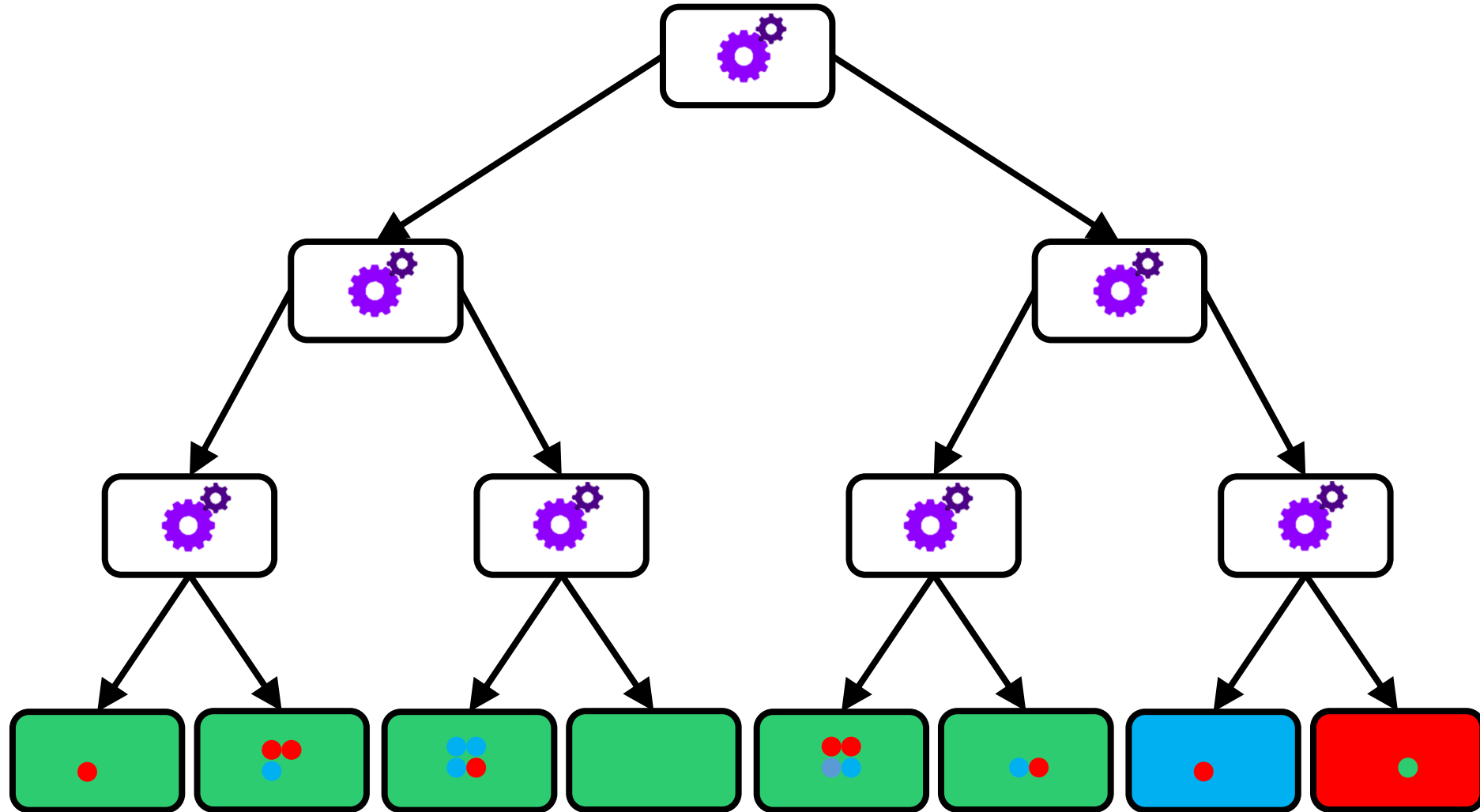
Growing the Tree



Growing the Tree

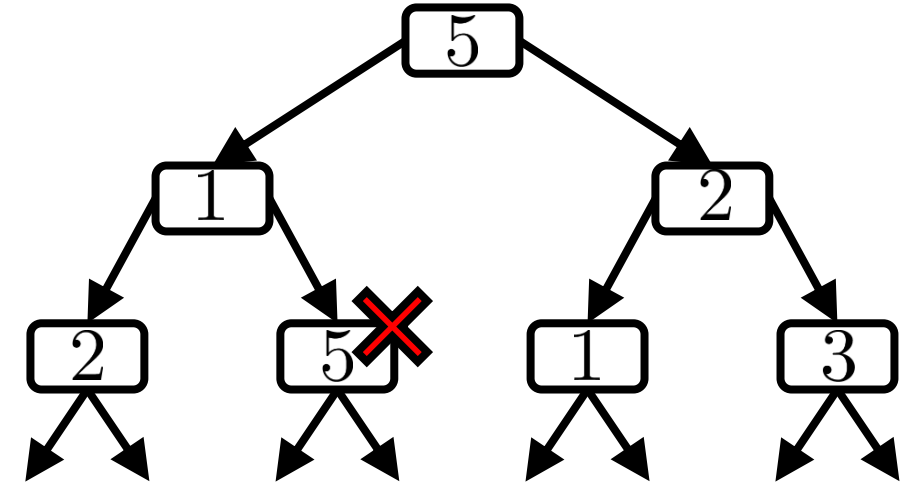


Growing the Tree



Pruning Strategies

- Stop if node is pure or almost pure
- Stop if all features exhausted
 - Don't use a feature twice on a path
 - Limits depth of tree to d
- Can stop if a node is ill-populated
- Can grow a tree and then merge nodes
 - Merge two leaves and see if it worsens or not - repeat
 - Use a validation set to make these decisions



A few Thoughts

- Why can't I use more features in a node decision stump?
 - Decision tree learning is an NP-hard problem in itself
 - Speed is an issue if using many features at each node
- Why cant I reuse features along a path?
 - Might mean you did not utilize it fully before
 - May lead to very deep trees
- Do I *have* to do things this way?
 - No, can experiment with well-founded, even crazy, ideas
 - But, give techniques that stood the test of time, a chance
- Some of the state-of-the-art multi-label classifiers are DTs

Please give your Feedback

<http://tinyurl.com/ml17-18afb>