

< Decision Tree - Continuous valued features >

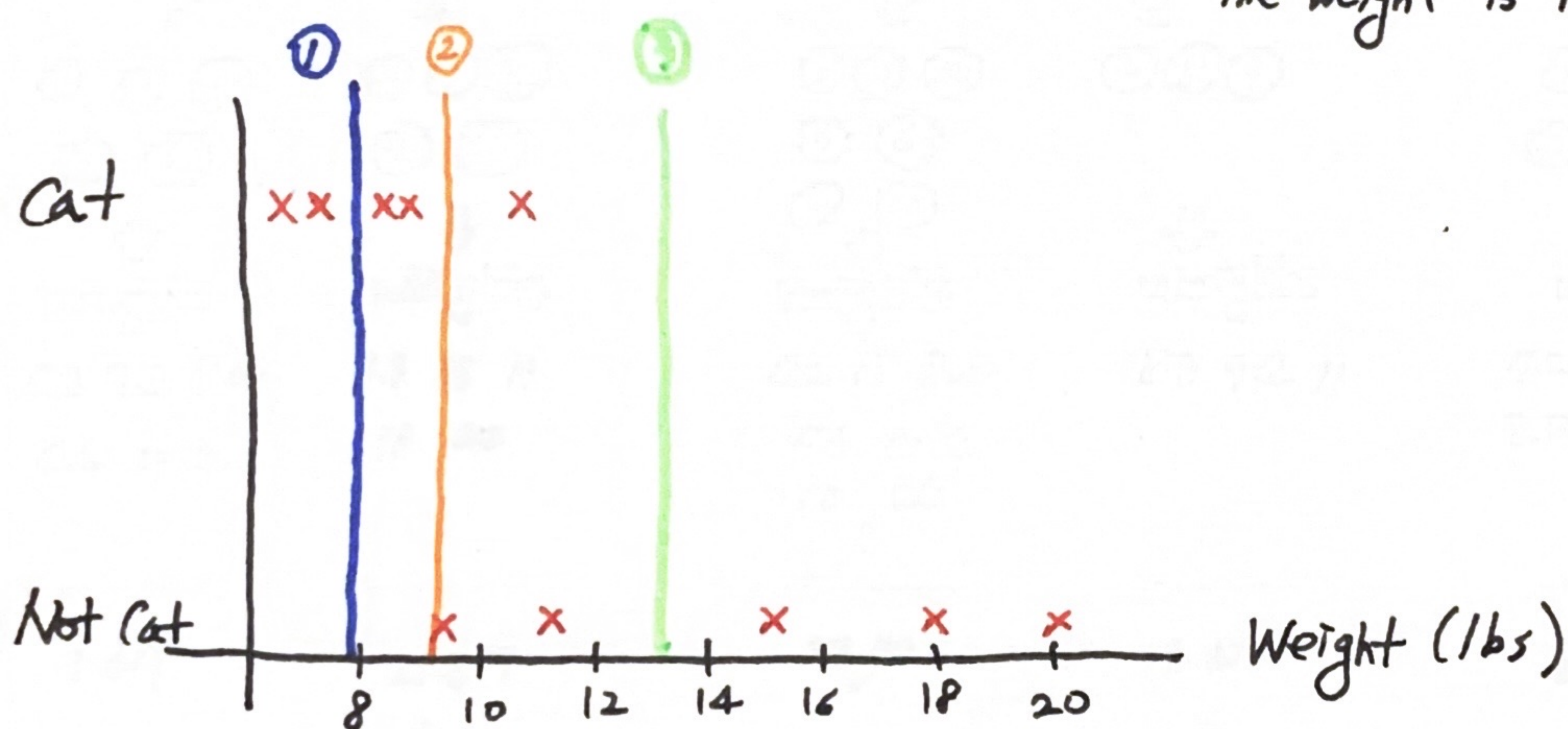
- not just discrete number but continuous value that is features can be any number

ex)

	Ear shape	Face shape	Whiskers	Weight (lbs)	Cat
①	Pointy	Round	Present	11.2	1
②	Flappy	Not round	Present	11.8	1
③	Flappy	Round	Absent	15	0
⋮	⋮	⋮	⋮	⋮	⋮

* Splitting on a continuous variable

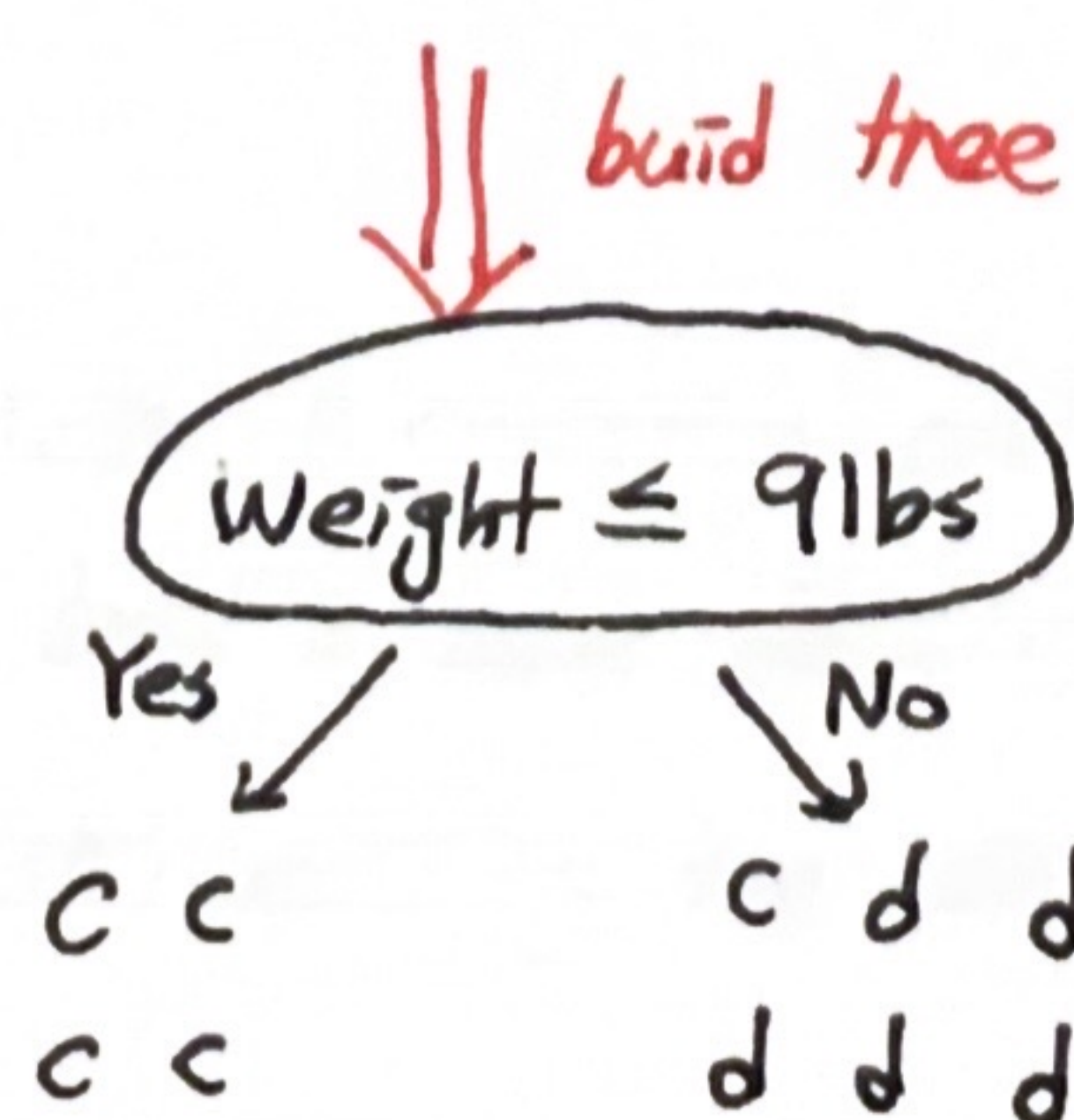
- the way to split on the weight feature : compute entropy based on whether or not the weight is less than or equal to some values threshold → for picking the best information gain



① $\text{Weight} \leq 8 \text{ lbs} \Rightarrow H(0.5) - \left(\frac{2}{10} H\left(\frac{2}{2}\right) + \frac{2}{10} H\left(\frac{2}{2}\right) \right) = 0.24$

② $\text{Weight} \leq 9 \text{ lbs} \Rightarrow H(0.5) - \left(\frac{4}{10} H\left(\frac{4}{4}\right) + \frac{6}{10} H\left(\frac{1}{6}\right) \right) = 0.61$

③ $\text{Weight} \leq 13 \text{ lbs} \Rightarrow H(0.5) - \left(\frac{7}{10} H\left(\frac{7}{7}\right) + \frac{3}{10} H\left(\frac{0}{3}\right) \right) = 0.40$



* Summarize *

decision tree with continuous value

① try different threshold

② do information gain calculation

③ split on the continuous value feature with the selected threshold which gives the best information gain