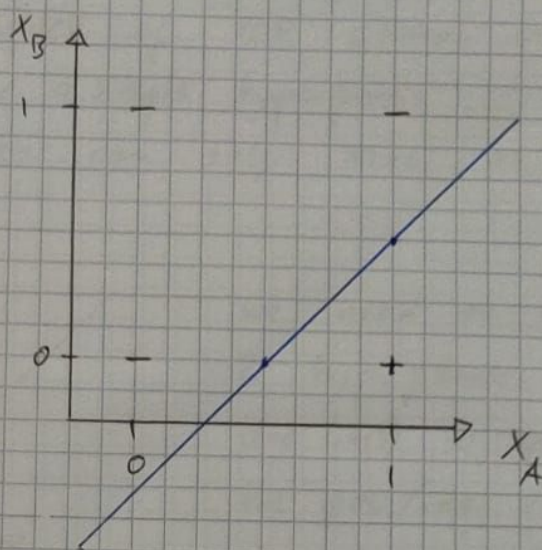


## Ex. Sheet 4

(E1)

(a)

$X_A$	$X_B$	$c(x)$
0	0	0
0	1	0
1	0	1
1	1	0



$$w_0 + X_A w_1 + X_B w_2 = 0$$

Let  $w_1 := 1$  Plug in  $X_A = 0.5, X_B = 0$

$$w_0 + 0.5 = 0$$

$$\Rightarrow w_0 = -0.5$$

Plug in  $X_A = 1, X_B = 0.5$

$$-0.5 + 1 + 0.5 w_2 = 0$$

$$\Rightarrow w_2 = -1$$

$$\Rightarrow \underline{w = (-0.5, 1, -1)}$$

$w_1 > 0$

because

$w_1 X_A$  should be

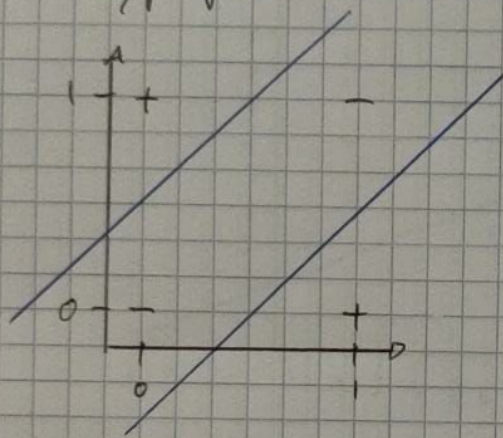
increasing if

$X_A$  is increasing



(b) Because in the feature space of  $[0,1]^2$ , positive and negative examples of XOR cannot be linearly separated by a single hyperplane.

A single perceptron can only serve as one hyperplane.

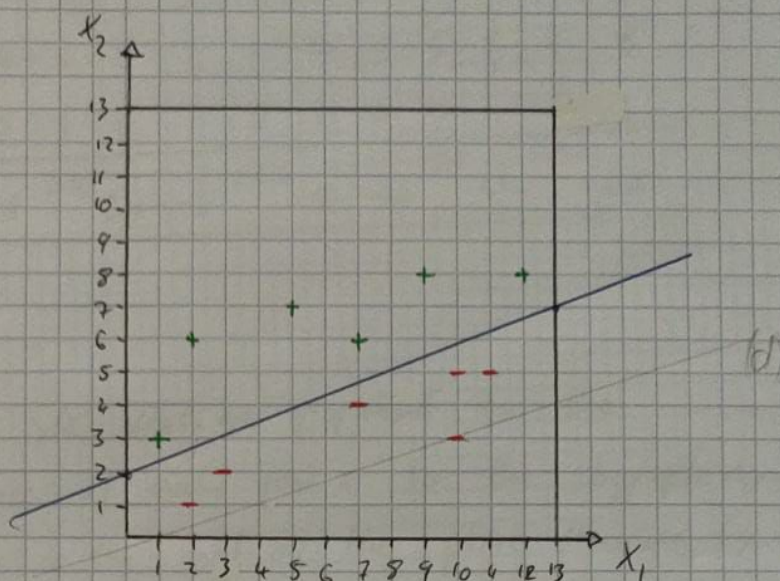


we need at least 2 hyperplanes (here blue lines) to separate the pos./neg. examples.



E2

(a)



$$(b) \quad w \cdot \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} = 0, \quad w \cdot \begin{bmatrix} 1 \\ 13 \\ 7 \end{bmatrix} = 0$$

$$w_0 + 0 \cdot w_1 + 2 \cdot w_2 = 0$$

$$\text{let } w_1 = -1$$

$$w_0 - 13 + 7w_2 = 0$$

$$\Rightarrow w_0 = 13 - 7w_2$$

$$13 - 7w_2 + 2w_2 = 0 = 13 - 5w_2$$

$$w_2 = \frac{13}{5}$$

$$= w_0 = 13 - 7 \cdot \frac{13}{5} = -\frac{26}{5}$$

$$\Rightarrow \underline{\underline{w = \left[ -\frac{26}{5}, -1, \frac{13}{5} \right]}}$$

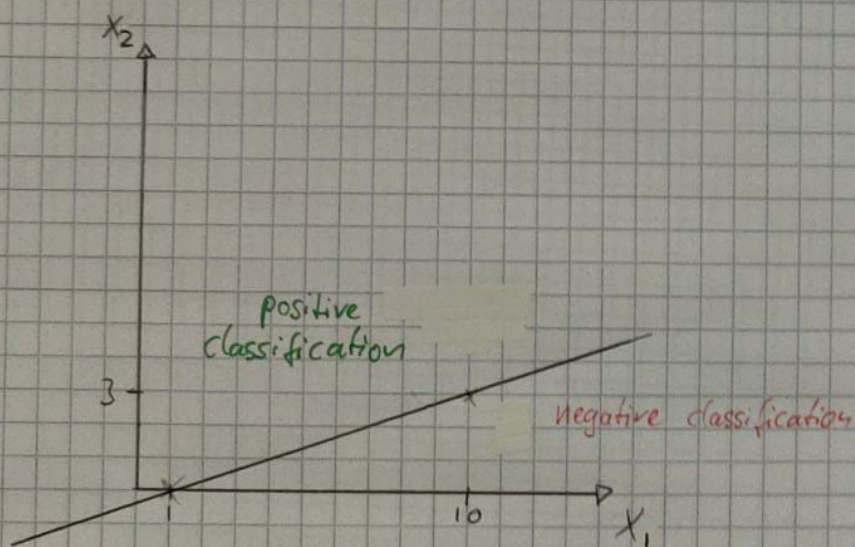


(c)

No	1	2	3	4	5	6	7	8	9	10	11	12
$x_1$	1	7	7	2	3	2	5	10	11	12	10	9
$x_2$	3	4	6	6	2	1	7	3	5	8	5	8
$c(x)$	1	0	1	1	1	0	1	0	0	1	0	1
$y(x) = w_0 + w_1 x_1 + w_2 x_2$	1.6	-1.8	3.4	8.4	-3	-4.6	8	-7.4	-3.2	3.6	-2.2	6.6

pattern: the  $y(x)$  values are proportional to the distance of each point  $x$  to the hyperplane, which is the root of the plane described by the  $w$  vector. Thus, the sign of an example's  $y(x)$  corresponds to the classification of the example  $x$ .

(d)



$$\text{let } x_1 = 1 \Rightarrow [1, -1, 3] \cdot \begin{bmatrix} 1 \\ 1 \\ x_2 \end{bmatrix} = 0$$

$$1 - 1 + 3x_2 = 0$$

$$x_2 = 0 \rightarrow p(1, 0)$$

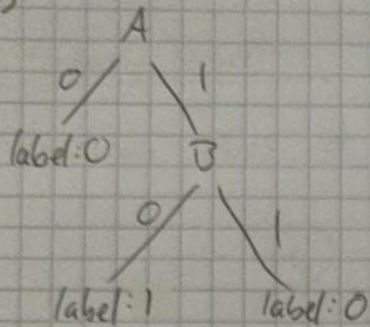
$$\text{let } x_1 = 10 \Rightarrow 1 - 10 + 3x_2 = 0$$

$$x_2 = 3 \rightarrow p(10, 3)$$

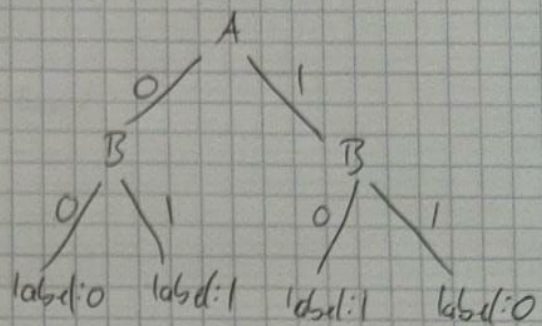


E3

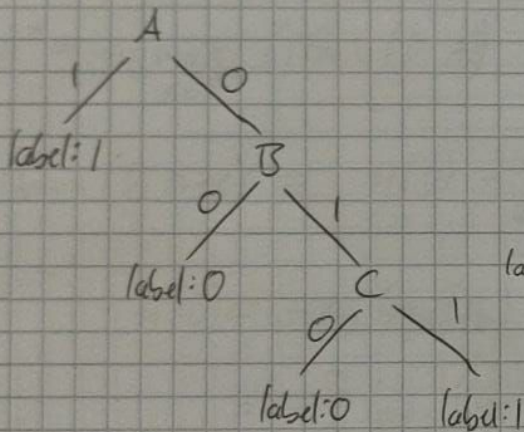
(a)  $A \wedge B$



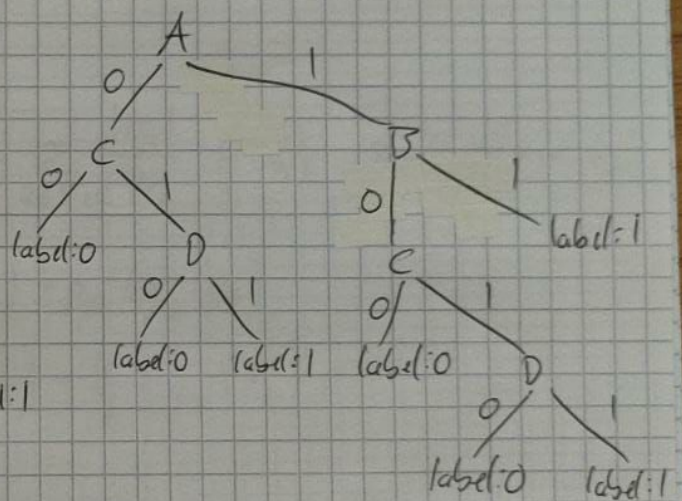
(b)  $A \text{ XOR } B$



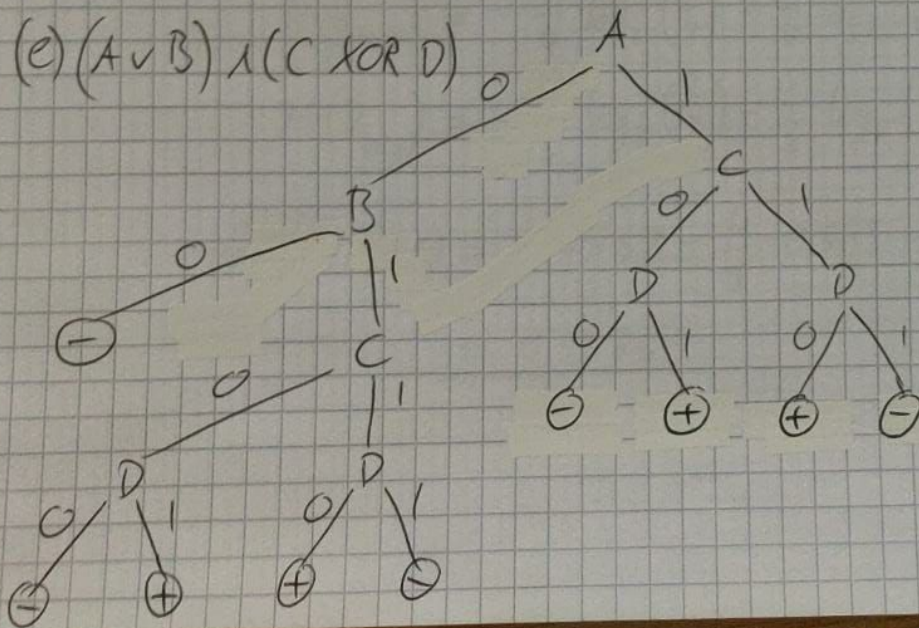
(c)  $A \vee (B \wedge C)$



(d)  $(A \wedge B) \vee (C \wedge D)$



(e)  $(A \vee B) \wedge (C \text{ XOR } D)$





E4

$\ominus$  = well-behaved,  $\oplus$  = dangerous

ID3(D, Features, Target):

most common class = well-behaved

$\odot$

label =  $\ominus$

impure  $\rightarrow$  yes, and Features  $\neq \emptyset$   
 $\hookrightarrow$  we split.

$$A^* = \operatorname{argmax}_{A \in F} (i\text{Gain}(D, A))$$

$$i\text{Gain}(D, \text{size}) \equiv H(D) - \sum_{a \in \text{size}} \frac{|D_a|}{|D|} \cdot H(D_a), \text{Dom}(\text{size}) = \{\text{small}, \text{big}\}$$

$$\begin{aligned} H(D) &= -p_{\ominus} \log_2(p_{\ominus}) - p_{\oplus} \log_2(p_{\oplus}) \\ &= -\frac{3}{7} \log_2\left(\frac{3}{7}\right) - \frac{4}{7} \log_2\left(\frac{4}{7}\right) \end{aligned}$$

$$H(D) \approx 0.9852$$

$$H(D_{\text{small}}) = -\frac{1}{4} \log_2\left(\frac{1}{4}\right) - \frac{3}{4} \log_2\left(\frac{3}{4}\right) \approx 0.8113$$

$$H(D_{\text{big}}) = -\frac{2}{3} \log_2\left(\frac{2}{3}\right) - \frac{1}{3} \log_2\left(\frac{1}{3}\right) \approx 0.9183$$

$$i\text{Gain}(D, \text{Size}) = 0.9852 - \frac{4}{7} \cdot 0.8113 - \frac{3}{7} \cdot 0.9183$$

$$\underline{\underline{\approx 0.1280}}$$



$$iGain(D, Fur): \quad Dom(Fur) = \{ragged, smooth, curly\}$$

$$iGain(D, Fur) = H(D) - \frac{|D_{ragged}|}{|D|} \cdot H(D_{ragged}) - \frac{|D_{smooth}|}{|D|} \cdot H(D_{smooth}) - \frac{|D_{curly}|}{|D|} \cdot H(D_{curly})$$

$$H(D_{ragged}) = -\frac{1}{3} \cdot \log_2\left(\frac{1}{3}\right) - \frac{2}{3} \cdot \log_2\left(\frac{2}{3}\right) \approx 0.9183$$

$$H(D_{smooth}) = -\frac{2}{2} \log_2\left(\frac{2}{2}\right) - \frac{0}{2} \cdot \log_2\left(\frac{0}{2}\right) = 0$$

$$H(D_{curly}) = 0$$

$$iGain(D, Fur) = 0.9852 - \frac{3}{7} \cdot 0.9183 = \underline{\underline{0.5916}}$$

$$iGain(Color): \quad Dom(Color) = \{brown, black, white, red\}$$

$$H(D_{brown}) = 0 \quad \text{because just one example...}$$

$$H(D_{black}) = -\frac{1}{3} \cdot \log_2\left(\frac{1}{3}\right) - \frac{2}{3} \log_2\left(\frac{2}{3}\right) = 0.9183$$

$$H(D_{white}) = -\frac{1}{2} \log_2\left(\frac{1}{2}\right) - \frac{1}{2} \log_2\left(\frac{1}{2}\right) = 1$$

$$H(D_{red}) = 0$$

$$iGain(Color) = 0.9852 - \frac{3}{7} \cdot 0.9183 - \frac{2}{7}$$

$$= \underline{\underline{0.3059}}$$

$$\Rightarrow \underset{A \in F}{\operatorname{argmax}} (iGain(A)) = \underline{\underline{Fur}}$$

The Attribute "Fur" is the feature that best classifies the examples in the Dataset.  $A^* = Fur$ .

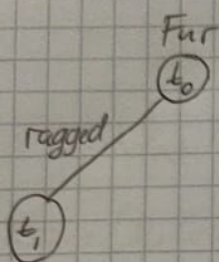


Now iterate values in  $A^*$  and do recursive calls.

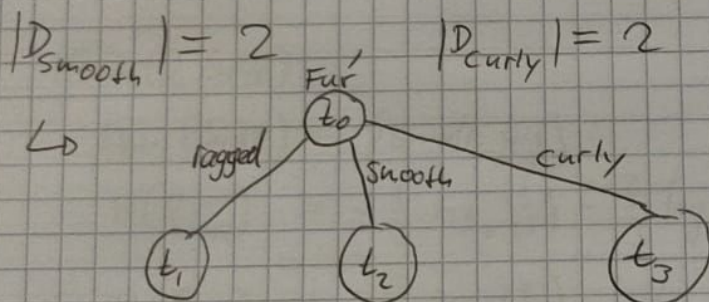
$$D_{\text{ragged}} = \{(X, c(X)) \in D : X|_{\text{Fur}} = \text{ragged}\}$$

$$|D_{\text{ragged}}| = 3 \Rightarrow D_{\text{ragged}} \neq \emptyset$$

$\Rightarrow$  create Edge (for ragged,  $\text{ID3}(D_{\text{ragged}}, F \setminus \{\text{Fur}\}, \text{Target})$ )



likewise for smooth, curly:



remaining calls before return:

$$\text{ID3}(D_{\text{ragged}}, F \setminus \{\text{Fur}\}, \text{Target}),$$

$$\text{ID3}(D_{\text{smooth}}, F \setminus \{\text{Fur}\}, \text{Target}),$$

$$\text{ID3}(D_{\text{curly}}, F \setminus \{\text{Fur}\}, \text{Target})$$

$$\text{ID3}(D_{\text{ragged}}, F \setminus \{\text{Fur}\}, \text{Target}):$$

$$\text{label}(t_1) = \text{mostCommonClass}(D_{\text{ragged}}, \text{Target}) = \text{well-behaved}$$

$$D := D_{\text{ragged}}, F := F \setminus \{\text{Fur}\}$$

impure  $\rightarrow$  yes.,  $F \neq \emptyset$

↳ Split.

$$A^* = \underset{A \in F}{\text{argmax}} (i\text{Gain}(D, A))$$

D:

Color	Size	Class
brown	small	$\ominus$
black	big	$\oplus$
red	big	$\oplus$



$$F = \{\text{Color, Size}\}$$

$$i\text{Gain}(D, \text{Color}): \quad \text{Dom}(\text{Color}) = \{\text{brown, black, white, red}\}$$

$$H(D) = -\frac{1}{3} \cdot \log_2\left(\frac{1}{3}\right) - \frac{2}{3} \log_2\left(\frac{2}{3}\right) = 0.9183$$

$$H(D_{\text{brown}}) = 0, \quad |D_{\text{brown}}| = 1$$

$$H(D_{\text{black}}) = 0, \quad |D_{\text{black}}| = 1$$

$$? D_{\text{white}} = \emptyset \quad H(D_{\text{white}}) = \frac{0}{-2}, \quad |D_{\text{white}}| = 0$$

$$H(D_{\text{red}}) = 0, \quad |D_{\text{red}}| = 1$$

$$\begin{aligned} i\text{Gain}(D, \text{Color}) &= H(D) - \frac{1}{3} \cdot 0 - \frac{1}{3} \cdot 0 - 0 \cdot 0 - \frac{1}{3} \cdot 0 \\ &= \underline{\underline{0.9183}} \end{aligned}$$

$$i\text{Gain}(D, \text{Size}):$$

$$\text{Dom}(\text{Size}) = \{\text{small, big}\}$$

$$H(D_{\text{small}}) = -\frac{1}{3} \cdot \log_2\left(\frac{1}{3}\right) - \frac{2}{3} \log_2\left(\frac{2}{3}\right) = 0.9183$$

$$H(D_{\text{big}}) = -\frac{2}{3} \log_2\left(\frac{2}{3}\right) - \frac{1}{3} \log_2\left(\frac{1}{3}\right) = 0.9183$$

$$\begin{aligned} i\text{Gain}(D, \text{Size}) &= 0.9183 - \frac{1}{3} \cdot 0.9183 - \frac{2}{3} \cdot 0.9183 \\ &= \underline{\underline{0}} \end{aligned}$$

$$\Rightarrow \underset{A \in F}{\text{argmax}} (i\text{Gain}(D, A)) = \text{Color}$$

The feature "Color" is the most discriminative feature, in the remaining dataset.

$$A^* = \text{Color.}$$



Iterate  $a \in A^*$ ,  $A^* = \{\text{brown, black, white, red}\}$

$$|D_{\text{brown}}| = 1$$

$\Rightarrow \text{createEdge}(t_1, \text{brown}, \text{ID3}(D_{\text{brown}}, F \setminus \{\text{Color}\}, \text{Target}))$

$$|D_{\text{black}}| = 1$$

$\Rightarrow \text{createEdge}(t_1, \text{black}, \text{ID3}(D_{\text{black}}, F \setminus \{\text{Color}\}, \text{Target}))$

$$|D_{\text{white}}| = 0$$

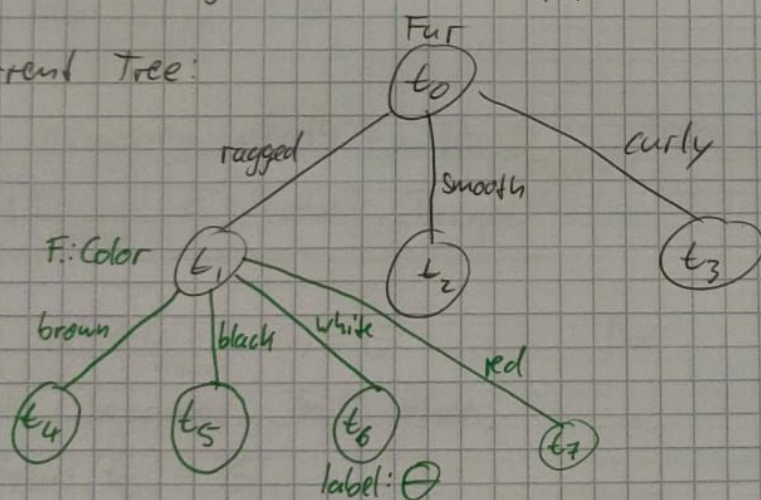
$\Rightarrow t_1$  is leaf Node,  $\text{label}(t_1) = \text{most Com. class in } D$

$\text{createEdge}(t_1, \text{white}, t_1) = \emptyset$

$$|D_{\text{red}}| = 1$$

$\Rightarrow \text{createEdge}(t_1, \text{red}, \text{ID3}(D_{\text{red}}, F \setminus \{\text{Color}\}, \text{Target}))$

Current Tree:



recursion depth	remaining function calls
1	$\text{ID3}(D_{\text{smooth}}, F \setminus \{\text{Fur}\}, \text{Target})$ $\text{ID3}(D_{\text{curly}}, F \setminus \{\text{Fur}\}, \text{Target})$
2	$\text{ID3}(D_{\text{brown}}, F \setminus \{\text{Color}\}, \text{Target})$ $\text{ID3}(D_{\text{black}}, F \setminus \{\text{Color}\}, \text{Target})$ $\text{ID3}(D_{\text{red}}, F \setminus \{\text{Color}\}, \text{Target})$

where  $F = F \setminus \{\text{Fur}\}$ ,  
 $D = D_{\text{ragged}}$



ID3( $D_{smooth}$ ,  $F \setminus \{Fur\}$ , Target)

$D := D_{smooth}$ ,  $F := F \setminus \{Fur\}$

$$D =$$

Color	Size	Class
black	big	dangerous
white	small	dangerous

$label(t_2) = \oplus = \text{dangerous.}$

$\forall (x, c(x)) \in D: c(x) = \oplus \Rightarrow \text{return } t_2$

$\Rightarrow t_2$  becomes leaf Node,  $label = \oplus$

ID3( $D_{curly}$ ,  $F \setminus \{Fur\}$ , Target)

$\forall (x, c(x)) \in D: c(x) = \ominus$

$\Rightarrow \text{return } t_3$

$\Rightarrow t_3$  becomes leaf Node,

$label(t_3) = \ominus$

$$D =$$

Color	Size	Class
black	small	$\ominus$
white	small	$\ominus$

Now recursion depth 2,  $D = D_{tagged}$ ,  $F = F \setminus \{Fur\}$

ID3( $D_{brown}$ ,  $F \setminus \{Color\}$ , Target):

$\forall (x, c(x)) \in D_{brown}: c(x) = \ominus$

$\Rightarrow t_4$  becomes leaf Node,

$label(t_4) = \ominus$

$$D =$$

Size	Class
small	$\ominus$

ID3( $D_{black}$ ,  $F \setminus \{Color\}$ , Target):

$\forall (x, c(x)) \in D_{black}: c(x) = \oplus$

$\Rightarrow t_5$  leaf Node,  $label(t_5) = \oplus$

$$D =$$

size	Class
big	$\oplus$

ID3( $D_{red}$ ,  $F \setminus \{Color\}$ , Target):

$\Rightarrow t_7$  leaf Node,  $label(t_7) = \ominus$

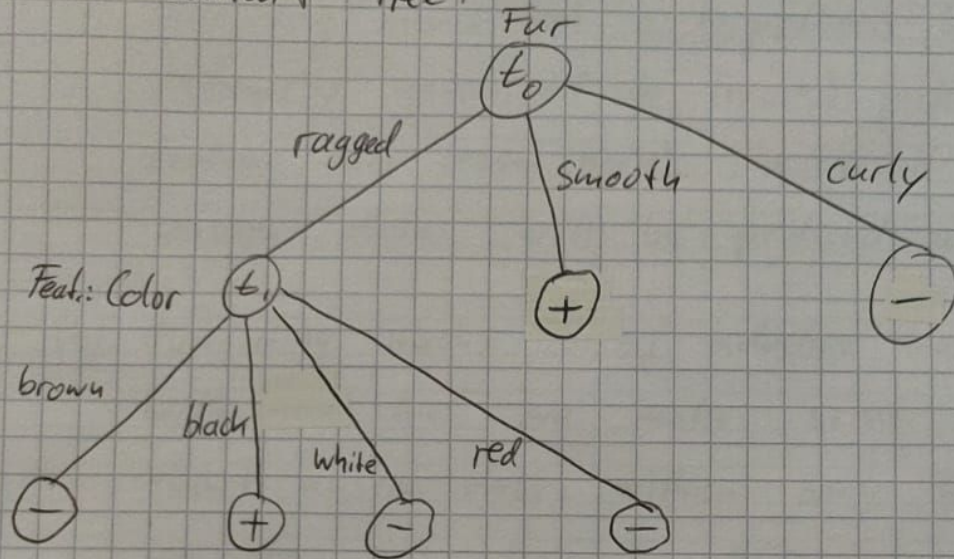
$$D =$$

Size	Class
big	$\ominus$



All recursive calls have returned.

The current Tree:



Finally, the initial call of ID3 will return  $t_0$ .

(b) dangerous, by rule: IF (Fur = ragged)  $\wedge$  (Color = black)  
THEN label =  $\oplus$ .