

CS 360: Machine Learning

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Fall 2019



HVERFORD
COLLEGE

Outline for October 24

- Ensemble methods
 - Bagging
 - Random forests
 - Boosting
 - Weighted entropy

In-lab today:
Hand back exam

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Bagging

Train

- use bootstrap resampling to generate $X^{(t)}$
for $t = 1 \dots T$
- train classifier on $X^{(t)}$ to get $h^{(t)}$

Test

for x in test data:

voting

$$h(x) = \operatorname{argmax}_{y \in \{0,1\}} \sum_{t=1}^T \mathbb{1}(h^{(t)}(\bar{x})=y)$$

$h^{(t)}(\bar{x}) \in \{0,1\}$

r = prob base classifier is wrong.

R = # of votes for the wrong class.

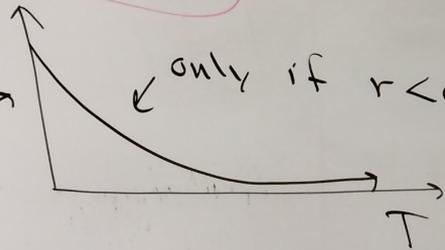
$$P(R=k) = \binom{T}{k} r^k (1-r)^{T-k}$$

$$\frac{T!}{k!(T-k)!}$$

$$P(R > \frac{T}{2}) = \sum_{k=\frac{T+1}{2}}^T \binom{T}{k} r^k (1-r)^{T-k}$$

Overall wrong!

only if $r < 0.5$



Example $T=3, r=\frac{1}{4}$

$\bar{X}_{test} = [\dots]$
 $Y_{test} = 1 \leftarrow true$

$h^{(1)}$	$h^{(2)}$	$h^{(3)}$	prob
0	0	0	$\frac{1}{4} \cdot \frac{1}{4} \cdot \frac{1}{4} = (\frac{1}{4})^3$
0	0	1	$(\frac{1}{4})^2 (\frac{3}{4})$
0	1	0	$(\frac{1}{4})^2 (\frac{3}{4})$
0	1	1	
1	0	0	$(\frac{1}{4})^2 (\frac{3}{4})$
1	0	1	
1	1	0	
1	1	1	

$$P(R > \frac{T}{2})$$

$$= (\frac{1}{4})^3$$

$$+ 3(\frac{1}{4})^2 (\frac{3}{4})$$

$$\approx \boxed{0.16}$$

lower error!

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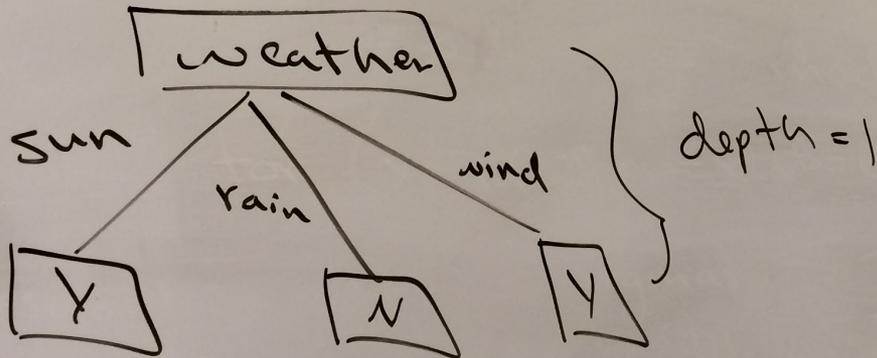
Random Forests

- Idea: choose a different subset of features for every classifier t
- Typically use *decision stumps* (depth 1)
- Goal: decorrelate models
- In practice: choose \sqrt{p} features
 - Without replacement for each model
 - Every model: data points and features chosen independently

Random Forests

* idea: need models to be decorrelated!

* choose ^{"weak"} base classifier
a decision stump



choose subset of features!

* in practice: \sqrt{p}
features
w/o replacement.

⇒ decorrelate stumps!

* generate random subset of features independently across models.

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Boosting Idea

→ $\frac{1}{n}$

Train:

- equal weights on all examples to start.

- for T iterations:

- learn classifier using weighted examples.
- change example weights based on train errors.

Test:

- get predictions from T classifiers
- vote based on how well that classifier did during training.

AdaBoost (Adaptive Boosting)

$$y \in \{-1, 1\}$$

set $w_i^{(1)} = \frac{1}{n}$ for $i = 1 \dots n$

for $t = 1 \dots T$: # of classifiers.

① fit classifier to weighted training set
 $\Rightarrow h^{(t)}(\vec{x})$

classifier index

② compute weighted classification error.

$$\epsilon_t = \sum_{i=1}^n w_i^{(t)} \mathbb{1}(y_i \neq h^{(t)}(\vec{x}_i))$$

③ compute model score.

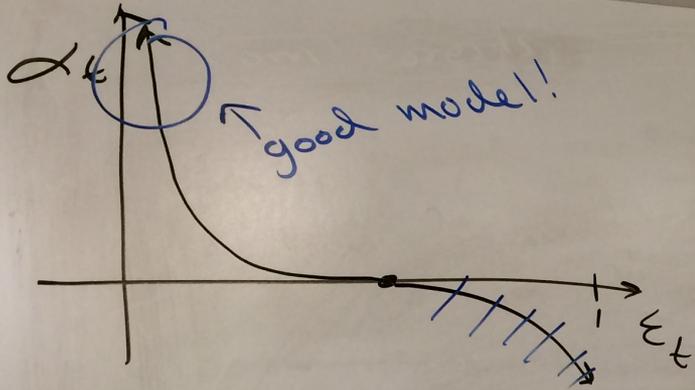
$$\alpha_t = \frac{1}{2} \log \left(\frac{1 - \epsilon_t}{\epsilon_t} \right)$$

④ update weights

$$w_i^{(t+1)} = c_t w_i^{(t)} \exp(-\alpha_t y_i h^{(t)}(\vec{x}_i))$$

normalizer
want weights
to sum to
1

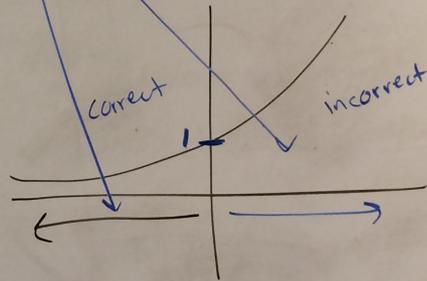
less than
1 if
correct
else > 1



$$\log\left(\frac{1 - \frac{1}{2}}{\frac{1}{2}}\right) = \log(1) = 0$$

$$\begin{cases} e^{-\alpha_t} & \text{if } y_i = h^{(t)}(\vec{x}_i) \\ e^{\alpha_t} & \text{if } y_i \neq h^{(t)}(\vec{x}_i) \end{cases}$$

$$\text{Sign}(x) = \begin{cases} +1 & \text{if } x > 0 \\ -1 & \text{if } x \leq 0 \end{cases}$$

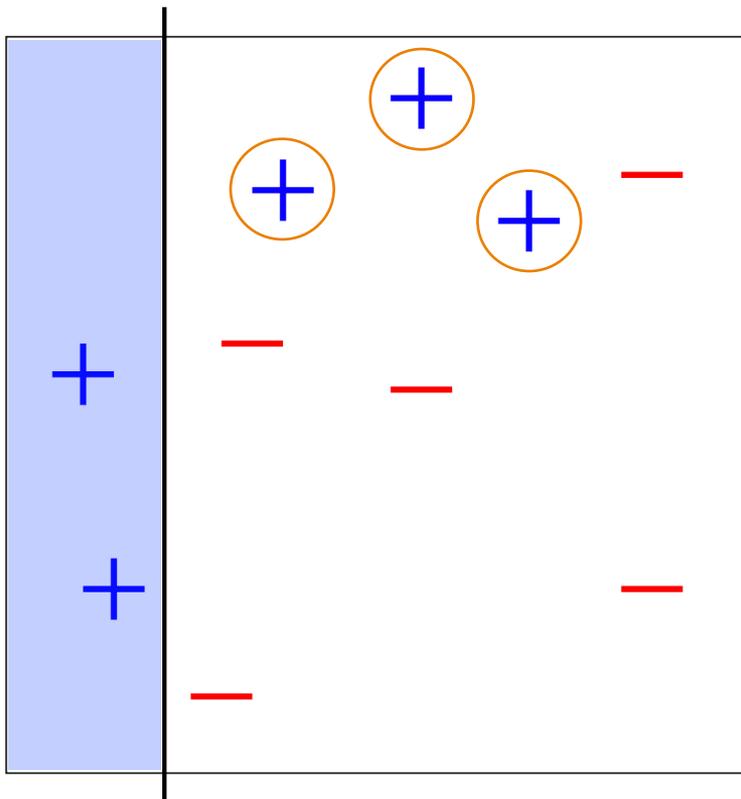


$$\underbrace{h(\vec{x})}_{\text{overall model}} = \text{sign}\left(\sum_{t=1}^T \alpha_t h^{(t)}(\vec{x})\right)$$

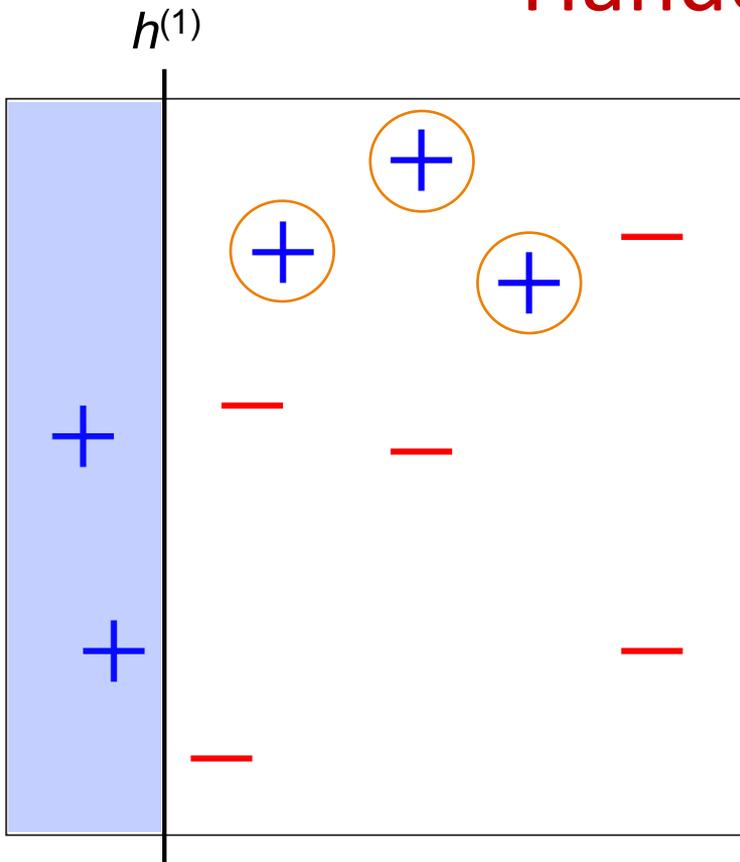
weighted
Vote.

Handout 14: Round 1

$h^{(1)}$



Handout 14: Round 1



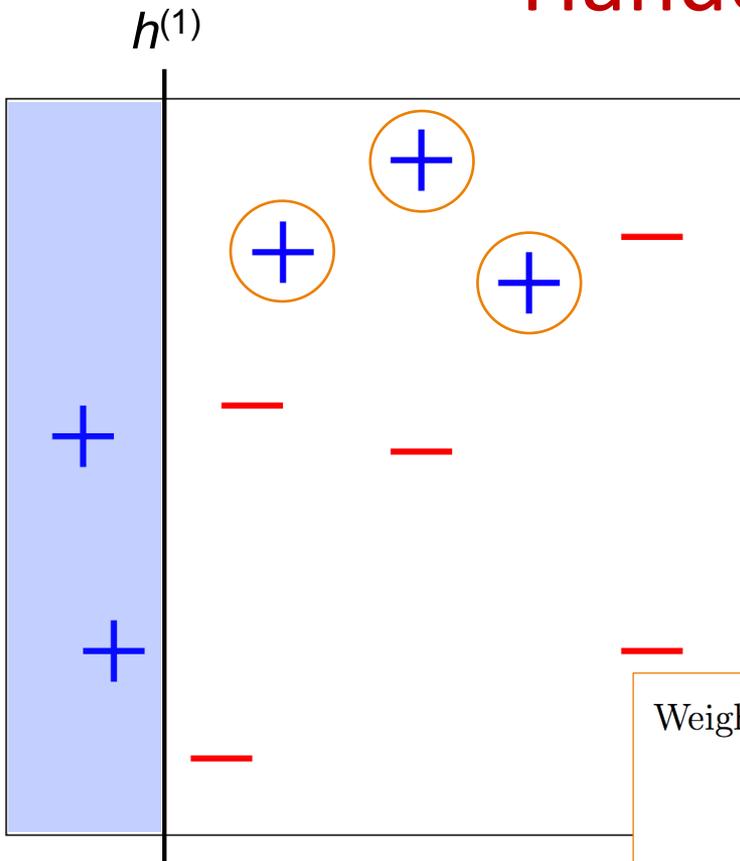
$$w_i^{(1)} = \frac{1}{10} \text{ for all } i = 1, 2, \dots, 10.$$

$$\epsilon_1 = \frac{3}{10} \text{ (three points incorrectly classified, all with weight } \frac{1}{10}\text{)}$$

$$\alpha_1 = \frac{1}{2} \ln \left(\frac{1 - \frac{3}{10}}{\frac{3}{10}} \right) = \ln \sqrt{\frac{7}{3}} \approx 0.42$$

- correctly classified: $w_i^{(2)} = c_1 \cdot \frac{1}{10} \exp \left(-\ln \sqrt{\frac{7}{3}} \right)$
- incorrectly classified: $w_i^{(2)} = c_1 \cdot \frac{1}{10} \exp \left(\ln \sqrt{\frac{7}{3}} \right)$

Handout 14: Round 1



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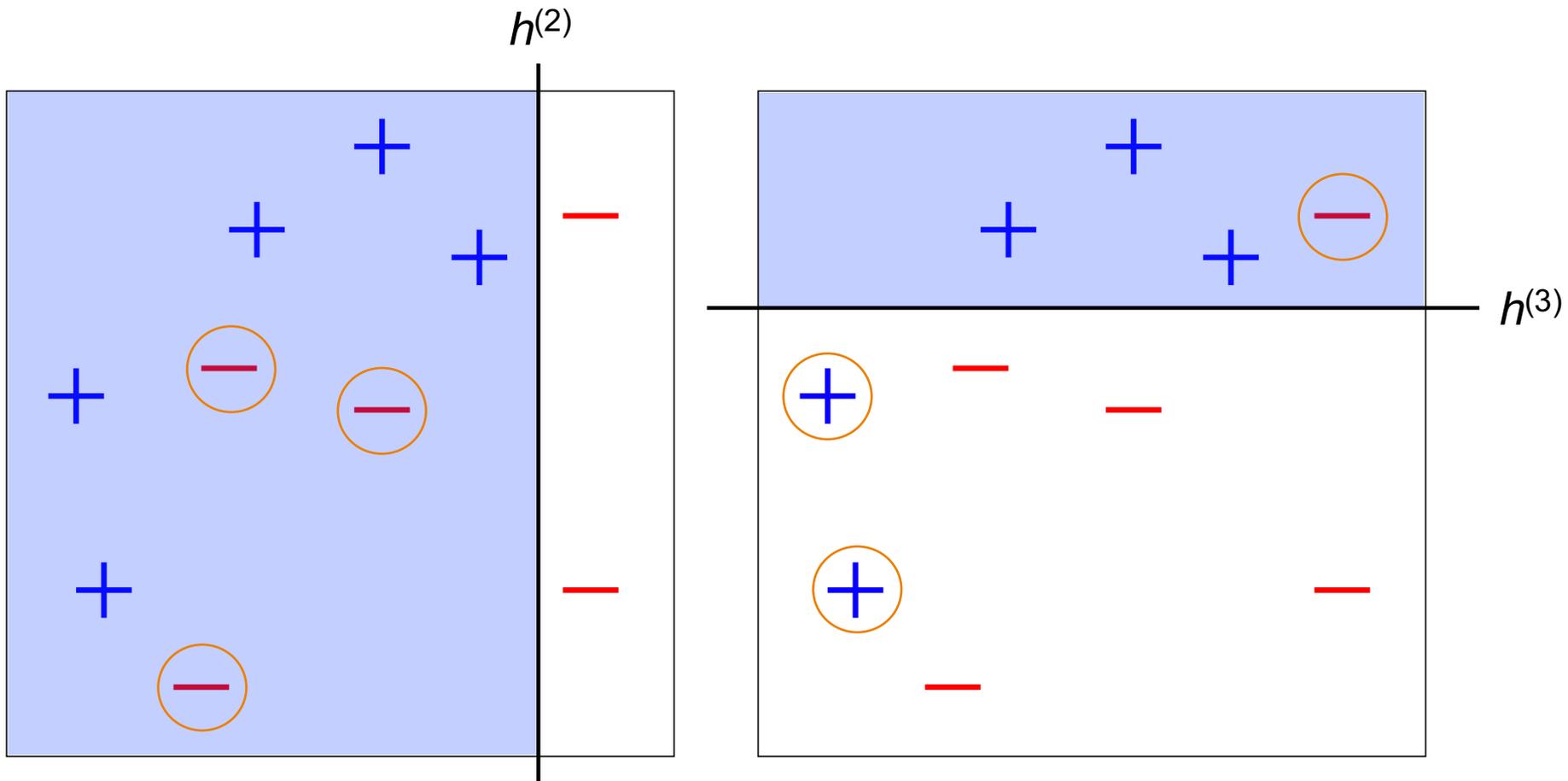
Weights must sum to 1, \Rightarrow

$$7 \cdot \frac{c_1}{10} \exp \left(-\ln \sqrt{\frac{7}{3}} \right) + 3 \cdot c_1 \cdot \frac{1}{10} \exp \left(\ln \sqrt{\frac{7}{3}} \right) = 1$$

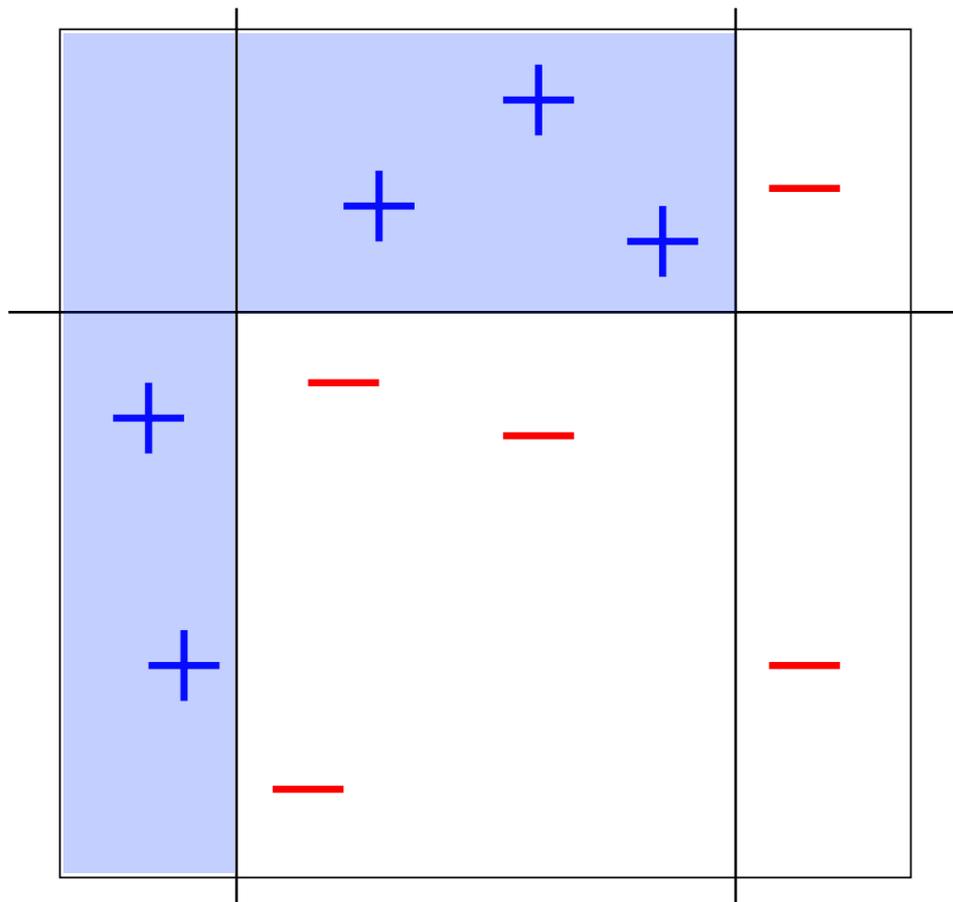
$$\Rightarrow c_1 = \frac{5}{\sqrt{21}}$$

- correctly classified: $w_i^{(2)} = \frac{5}{\sqrt{21}} \cdot \frac{1}{10} \sqrt{\frac{3}{7}} = \frac{1}{14}$ decrease!
- incorrectly classified: $w_i^{(2)} = \frac{5}{\sqrt{21}} \cdot \frac{1}{10} \sqrt{\frac{7}{3}} = \frac{1}{6}$ increase!

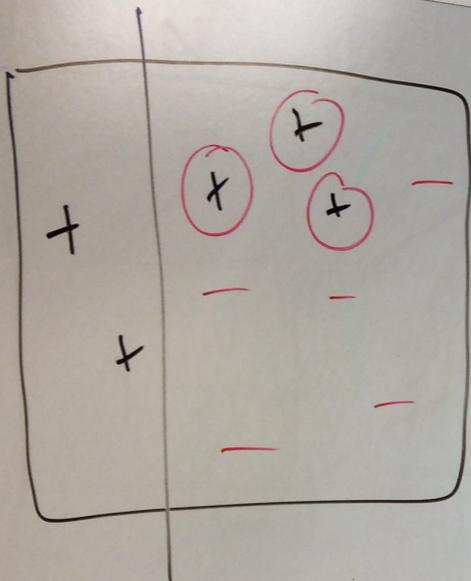
Handout 14: Round 2 & 3 (exercise!)



Handout 14: final classifier



$$h(\mathbf{x}) = \text{sign}\left(0.42 \cdot h^{(1)}(\mathbf{x}) + 0.65 \cdot h^{(2)}(\mathbf{x}) + 0.92 \cdot h^{(3)}(\mathbf{x})\right)$$



$$w_i^{(1)} = \frac{1}{10} \quad \forall i$$

(b) weighted error:

$$\epsilon_1 = \frac{3}{10}$$

$$\begin{aligned} \alpha_1 &= \frac{1}{2} \log\left(\frac{1 - \frac{3}{10}}{\frac{3}{10}}\right) = \frac{1}{2} \log \frac{7}{3} \\ &= \log \sqrt{\frac{7}{3}} \end{aligned}$$

missclassified

$$\begin{aligned} w_i^{(2)} &= c_i \frac{1}{10} \exp(\log \sqrt{\frac{7}{3}}) \\ &= c_i \frac{1}{10} \sqrt{\frac{7}{3}} \end{aligned}$$

correct

$$w_i^{(2)} = c_i \frac{1}{10} \sqrt{\frac{3}{7}}$$

$$3 \left(c_1 \frac{1}{10} \sqrt{\frac{7}{3}} \right) + 7 \left(c_1 \frac{1}{10} \sqrt{\frac{3}{7}} \right) = 1$$

incorrect : $\frac{1}{6}$

correct : $\frac{1}{14}$

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- See back of Handout 13

- Note: entropy calculation should have a negative sign in front!