

concept learning 1

- i) Prove that each minimal specialization h of $g \in G$ (g not consistent with d) is consistent with all the training examples observed thus far
- (-ve example)
1. some member of S is more specific than or equal to h (candidate elimination algorithm)
 $\exists s \in S \quad h \geq g \quad s$
 2. since $s \in S$, s is consistent with all training examples so far (given in assumption)
 3. Therefore, every true training example satisfies s (property 1)
 4. since $h \geq g \quad s$, $\forall x \in X, s(x)=1 \rightarrow h(x)=1$ (def of $\geq g$)
 5. Every positive training observed thus far satisfies h
 6. Thus, h is consistent with all the training examples observed thus far.
- ii) Prove that each minimal specialization h of $g \in G$ is consistent with all -ve training examples observed thus far, including d

1. $\forall g \in G$, g is consistent with all training examples observed so far (not including d)
2. Therefore, every -ve training example not including d does not satisfy g (prop 2)
3. since h is a minimal specialization of g , $g \geq g \quad h$
 $\forall x \in X, g(x)=0 \rightarrow h(x)=0$ (contrapositive of $\geq g$)
4. every negative training example thus far does not satisfy h (not including d)
5. since h is consistent with d , h is consistent with all negative training examples observed thus far.

concept learning 2

i) proof by contradiction

$$\begin{aligned} \text{suppose } & \sim \left[(\forall g \in G \quad g \not\geq g h') \rightarrow (\forall h \in VSH, D \quad h \not\geq g h') \right] \\ & \equiv (\forall g \in G \quad g \not\geq g h') \wedge \sim (\forall h \in VSH, D \quad h \not\geq g h') \quad \downarrow \sim(p \rightarrow q) \\ & \equiv (\forall g \in G \quad g \not\geq g h') \wedge (\exists h \in VSH, D \quad h \geq g h') \quad \equiv \sim(\sim p \vee q) \\ & \equiv p \wedge \sim q \end{aligned}$$

1. $\exists h \in VSH, D$ such that $h \geq g h'$
2. $\exists g \in G$ such that $g \geq g h$ (VSR7)
3. By transitive property of \geq_g relation, $\exists g \in G \quad g \geq g h \geq g h' \quad g \geq g h'$
4. contradiction with $\forall g \in G \quad g \not\geq g h'$, assumption must be false
5. Thus $(\forall g \in G \quad g \not\geq g h') \rightarrow (\forall h \in VSH, D \quad h \not\geq g h')$ must be true

ii) proof by contradiction

$$\begin{aligned} \text{suppose } & \sim \left[(\forall s \in S \quad h' \not\geq g s) \rightarrow (\forall h \in VSH, D \quad h' \not\geq g h) \right] \\ & \equiv \forall s \in S \quad h' \not\geq g s \wedge \exists h \in VSH, D \quad h' \geq g h \end{aligned}$$

1. $\exists h \in VSH, D$ such that $h' \geq g h$
2. $\exists s \in S$ such that $h' \geq g s$ (VSR7)
3. By transitive property of \geq_g relation, $\exists s \in S \quad h' \geq g h \geq g s \quad h' \geq g s$
4. contradiction with $\forall s \in S \quad h' \not\geq g s$, initial assumption is false
5. Thus $(\forall s \in S \quad h' \not\geq g s) \rightarrow (\forall h \in VSH, D \quad h' \not\geq g h)$ is true

Concept Learning 3

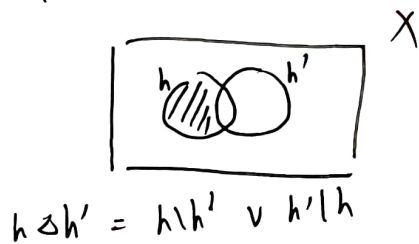
$$G_0 = \{ \langle \text{????} \rangle \Delta \langle \emptyset, \emptyset, \emptyset, \emptyset \rangle \} = \{ \langle \text{????} \rangle \}$$

$$S_0 = \{ \langle \emptyset, \emptyset, \emptyset, \emptyset \rangle \Delta \langle \emptyset, \emptyset, \emptyset, \emptyset \rangle, \langle \text{????} \rangle \Delta \langle \text{????} \rangle, \dots \} = \{ \langle \emptyset, \emptyset, \emptyset, \emptyset \rangle \}$$

$$h \setminus h' = h(x)=1 \wedge h'(x)=0$$

$$G_1 = G_0$$

$$S_1 = \{ \langle \text{sometimes}, \text{Good}, \text{Average}, \text{Yes} \rangle \Delta \langle \emptyset, \emptyset, \emptyset, \emptyset \rangle \}$$



$$G_2 = G_1$$

$$S_2 = \{ \langle \text{sometimes}, \text{Good}, \text{Average}, \text{Yes} \rangle \Delta \langle \text{Always}, \text{Good}, \text{Average}, \text{Yes} \rangle \}$$



$$G_3 = G_2$$

$$S_3 = \{ \langle \text{?}, \text{Good}, \text{Average}, \text{Yes} \rangle \Delta \langle \text{Rarely}, \text{Good}, \text{Good}, \text{Yes} \rangle \}$$

$$G_4 = G_3$$

$$S_4 = \{ \langle \text{?}, \text{Good}, \text{Average}, \text{Yes} \rangle \Delta \langle \text{?}, \text{Good}, \text{Good}, \text{Yes} \rangle \}$$

$$S_5 = S_4$$

$$G_5 = \{ \langle \text{????} \rangle \Delta \langle \text{Rarely}, \text{Poor}, \text{Good}, \text{Yes} \rangle \}$$

Part 4 EM

$$\langle x_1, x_2, x_3 \rangle = \langle 4, 5, 6 \rangle, \quad M=2, \quad \sigma^2 = 0.5$$

$$\langle \mu_1, \mu_2 \rangle = \langle 4, 6 \rangle$$

run EM for 3 iterations, estimate μ_1, μ_2 up to 6 dp

intermediate values dep

Iteration 1

E step

$$E[Z_{11}] = \frac{e^{-\frac{1}{\sigma^2}(4-4)^2}}{\sum_{k=1}^2 e^{-\frac{1}{\sigma^2}(4-\mu_k)^2}} = \frac{1}{1.01831564} = 0.98201379$$

$$E[Z_{12}] = \frac{e^{-\frac{1}{\sigma^2}(4-6)^2}}{1.01831564} = 0.01798621$$

$$E[Z_{21}] = \frac{e^{-\frac{1}{\sigma^2}(5-4)^2}}{e^{-\frac{1}{\sigma^2}(5-4)^2} + e^{-\frac{1}{\sigma^2}(5-6)^2}} = \frac{1}{2}$$

$$E[Z_{22}] = \frac{1}{2}$$

$$E[Z_{31}] = \frac{e^{-\frac{1}{\sigma^2}(6-4)^2}}{1.01831564} = 0.01798621$$

$$E[Z_{32}] = 0.98201379$$

M step

$$\mu'_1 = \frac{E[Z_{11}] \cdot x_1 + E[Z_{21}] \cdot x_2 + E[Z_{31}] \cdot x_3}{E[Z_{11}] + E[Z_{21}] + E[Z_{31}]} = 4.35731495$$

$$\mu'_2 = 5.64268505$$

$$\sum_{k=1}^2 e^{-\frac{1}{\sigma^2}(4-\mu_k)^2} = 0.85014028 + 0.06731217$$

Iteration 2

E step

$$E[Z_{11}] = 0.92895457$$

$$E[Z_{12}] = 0.07104543$$

$$E[Z_{21}] = \frac{1}{2}$$

$$E[Z_{22}] = \frac{1}{2}$$

$$E[Z_{31}] = 0.07104543$$

$$E[Z_{32}] = 0.92895457$$

M step

$$\mu'_1 = 4.42806057$$

$$\mu'_2 = 5.57193943$$

$$= 0.94745245$$

$$= \sum_{k=1}^2 e^{-\frac{1}{\sigma^2}(6-\mu_k)^2}$$

$$\sum_{k=1}^2 e^{-\frac{1}{\sigma^2}(4-\mu_k)^2} = 0.83257177 + 0.08450086$$

Iteration 3

E step

$$E[Z_{11}] = 0.90785805$$

$$E[Z_{12}] = 0.09214195$$

$$E[Z_{21}] = E[Z_{22}] = \frac{1}{2}$$

$$E[Z_{31}] = 0.09214195$$

$$E[Z_{32}] = 0.90785805$$

M step

$$\boxed{\begin{aligned} \mu'_1 &= 4.45618927 \\ \mu'_2 &= 5.54381073 \end{aligned}}$$

$$= 0.91707263$$

Part 5 CLT not-toried ?

Part 6 Neural Networks

i) $w_0 = -1, w_2 = 1$ $\text{AND}(x_1, x_2)$

Case 1: Perceptron outputs true if both inputs are true

$$w_0 + w_1 x_1 + w_2 x_2 = -1 + w_1 + 1 \\ = w_1 > 0$$

Case 2: Perceptron outputs false if at least 1 input is false

$$(1, -1) \Rightarrow -1 + w_1 - 1 = w_1 - 2 \leq 0 \Rightarrow w_1 \leq 2$$

$$(-1, 1) \Rightarrow -1 - w_1 + 1 = -w_1 \leq 0 \Rightarrow w_1 \geq 0$$

$$(-1, -1) \Rightarrow -1 - w_1 - 1 = -w_1 - 2 \leq 0 \Rightarrow w_1 \geq -2$$

} restricted by $w_1 > 0$ in case 1

$$\therefore 0 < w_1 \leq 2$$

ii) $w_1, w_2, \dots, w_n = -1$ NOR function

Case 1: Perceptron outputs true if all n Boolean inputs are false

$$\sum_{i=1}^n x_i = \sum_{i=1}^n -1 = -n$$

$$w_0 + (-1) \sum_{i=1}^n x_i = w_0 + n > 0 \Rightarrow w_0 > -n$$

Case 2: Perceptron outputs false if at least 1 input is true

$$\sum_{i=1}^n x_i \geq 1 + n(-1) = 2 - n$$

$$- \sum_{i=1}^n x_i \leq n - 2$$

$$w_0 + (-1) \sum_{i=1}^n x_i \leq w_0 + n - 2 \leq 0 \Rightarrow w_0 \leq 2 - n$$

$$-n < w_0 \leq 2 - n$$

iii)

x_1	y_2	z_k
1	1	1
1	-1	-1
-1	1	-1
-1	-1	1

xw13

x	weights	z_k
$(1, 1)$	$(w_1 + w_2) > 0? \quad w_3 : -w_3 > 0$ $+ w_4 + w_5$	$2 > 0$

$$(1, -1) \quad (w_1 - w_2) > 0 \quad ? \quad w_3 : -w_3 \leq 0 \quad 0 \leq 0$$

$w_4 \neq w_5$ else either $(1,1) (2,2)$
or $(-1,-1) (-2,-2)$

have a -2 or $-4 \Rightarrow \neq 0$ with $+2$

$$(-1, 1) \quad (-w_1 + w_2) > 0 \quad w_3 : -w_3 \leq 0 \quad 0 \leq 0$$

$$-w_4 \leq w_5$$

$$\text{fng } (-1, 1) \quad \left. \begin{array}{l} u_1 + u_3 = 0 \\ -u_1 - u_3 = 0 \end{array} \right\} \begin{array}{l} \text{same as} \\ \text{with 2} \\ -u_1 - u_3 \end{array}$$

$$\Rightarrow u_3 = 1 \text{ or } -2$$

$$(-1, -1) \quad (-w_1 - w_2) > 0 \quad w_3 := w_4 > 0 \quad 270$$

$$-w_4 + w_5 = 2$$

$$\boxed{w_3 = -2}$$

$$(-1, 1, -2, -1, 1)$$