

Inductive Learning

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$$\begin{aligned}\frac{\partial}{\partial \theta_i} J(\theta_i) &= \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) x_j^{(i)} \\ \frac{\partial}{\partial \theta_0} J(\theta_0, \theta_1) &= \frac{1}{4} ((\theta_0 + 3\theta_1 - 2) + (\theta_0 + \theta_1 - 2) + (\theta_0 - 1) + (\theta_0 + 4\theta_1 - 3)) \\ &= \theta_0 + 2\theta_1 - 2 \\ \frac{\partial}{\partial \theta_1} J(\theta_0, \theta_1) &= \frac{1}{4} (3(\theta_0 + 3\theta_1 - 2) + (\theta_0 + \theta_1 - 2) + 0(\theta_0 - 1) + 4(\theta_0 + 4\theta_1 - 3)) \\ &= 2\theta_0 + 6.5\theta_1 - 5\end{aligned}$$

Let θ_0^i denotes the i th round value of θ_0 , $\alpha = 0.1$.

- Round 0

$$\theta_0^0 = 0 \tag{1}$$

$$\theta_1^0 = 1 \tag{2}$$

- Round 1

$$\theta_0^1 = 0 - 0.1 \times (0 + 2 - 2) = 0 \tag{3}$$

$$\theta_1^1 = 1 - 0.1 \times (2 \times 0 + 6.5 \times 1 - 5) = 0.85 \tag{4}$$

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- Round 2

$$\theta_0^2 = 0 - 0.1 \times (0 + 2 \times 0.85 - 2) = 0.03 \quad (5)$$

$$\theta_1^2 = 0.85 - 0.1 \times (2 \times 0 + 6.5 \times 0.85 - 5) = 0.7975 \quad (6)$$

- Round 3

$$\theta_0^3 = 0.03 - 0.1 \times (0.03 + 2 \times 0.7975 - 2) = 0.0675 \quad (7)$$

$$\theta_1^3 = 0.7975 - 0.1 \times (2 \times 0.03 + 6.5 \times 0.7975 - 5) = 0.773125 \quad (8)$$

- Round 4

$$\theta_0^4 = 0.0675 - 0.1 \times (0.0675 + 2 \times 0.773125 - 2) = 0.106125 \quad (9)$$

$$\theta_1^4 = 0.773125 - 0.1 \times (2 \times 0.0675 + 6.5 \times 0.773125 - 5) = 0.7571 \quad (10)$$

- Round 5

$$\theta_0^5 = 0.106125 - 0.1 \times (0.106125 + 2 \times 0.7571 - 2) = 0.1441 \quad (11)$$

$$\theta_1^5 = 0.7571 - 0.1 \times (2 \times 0.106125 + 6.5 \times 0.7571 - 5) = 0.7438 \quad (12)$$

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- False Positive: 10%
- False Negative: 20%

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a) Specific hypothesis (S)

Pros Less false positive error;

Cons More false negative errors.

b) General hypothesis (G)

Pros Less false negative errors.

Cons More false positive error;

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Definition 1. A hypothesis h is *consistent* with a set of training examples D if and only if $h(x) = c(x)$ for each example $\langle x, c(x) \rangle$ in D .

$$\text{Consistent}(h, D) \equiv (\forall \langle x, c(x) \rangle \in D) h(x) = c(x)$$

Definition 2. The *version space*, denoted $VS_{H,D}$, with respect to hypothesis space H and training examples D , is the subset of hypotheses from H consistent with the training examples in D .

$$VS_{H,D} \equiv \{h \in H | \text{Consistent}(h, D)\}$$

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The most general hypothesis has (?) value for each attribute.

6

a) $|X| = 3^4 = 81$.

b) $(\text{instance}) = 2^{16}$.

c) $\binom{4}{2} \times 4 = 24$

7

$$h_0 \leftarrow \langle \emptyset, \emptyset, \emptyset, \emptyset, \emptyset \rangle$$

$$h_1 \leftarrow \langle 1, 1, 0, 1, 1 \rangle$$

$$h_2 \leftarrow \langle 1, 1, 0, 1, 1 \rangle$$

$$h_3 \leftarrow \langle 1, 1, ?, 1, ? \rangle$$

$$h_4 \leftarrow \langle 1, 1, ?, 1, ? \rangle$$

$$h_5 \leftarrow \langle 1, 1, ?, 1, ? \rangle$$

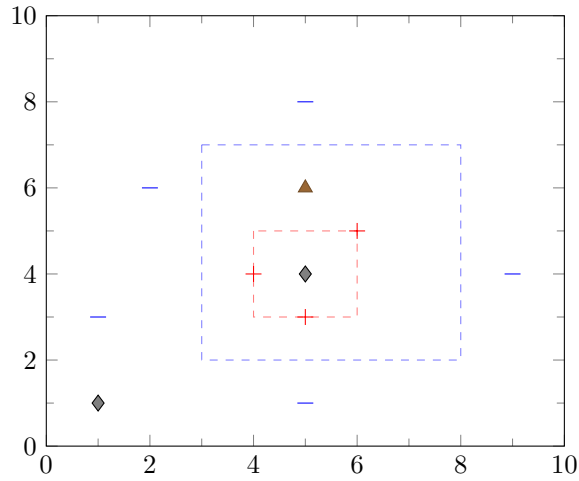
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$$(GPA < 3.5 \wedge Exp \geq 3) \vee (GPA \geq 3.5 \wedge Exp \geq 1) \quad (13)$$

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- a) $S \equiv (4 \leq x \leq 6) \wedge (3 \leq y \leq 5)$, shown as red dash line square in fig. 1. Points on the line should be included.
- b) $G \equiv (3 \leq x \leq 8) \wedge (2 \leq y \leq 7)$, shown as blue dash line square in fig. 1. Points on the line should be excluded.
- c) A query lying between S and G would guarantee to reduce the size of the version space, for example $x = 5, y = 6$, shown as triangle in fig. 1. A query outside G or inside S would not reduce size of the version space, for example $x = 1, y = 1$ and $x = 5, y = 4$, shown as diamond in fig. 1.

Figure 1: S and G in Diagram



- d) Four training sample would be sufficient to achieve a particular target concept via CANDIDATE-ELIMINATION algorithm. Take concept $3 \leq x \leq 5, 2 \leq y \leq 9$ as an example. The following four training sample would be sufficient.

+ (3,2)

- + (5,9)
- (2,1)
- (6,10)

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a) Trace:

i. Original hypothesis.

ii. + $\langle\langle \text{male brown tall } US \rangle \langle \text{female black short } US \rangle \rangle$

$$S \equiv \langle\langle \text{male brown tall } US \rangle \langle \text{female black short } US \rangle \rangle \quad (14)$$

$$G \equiv \langle\langle ? ? ? ? \rangle \langle ? ? ? ? \rangle \rangle \quad (15)$$

iii. + $\langle\langle \text{male brown short French} \rangle \langle \text{female black short } US \rangle \rangle$

$$S \equiv \langle\langle \text{male brown ? ?} \rangle \langle \text{female black short } US \rangle \rangle \quad (16)$$

$$G \equiv \langle\langle ? ? ? ? \rangle \langle ? ? ? ? \rangle \rangle \quad (17)$$

iv. - $\langle\langle \text{female brown tall German} \rangle \langle \text{female black short Indian} \rangle \rangle$

$$S \equiv \langle\langle \text{male brown ? ?} \rangle \langle \text{female black short } US \rangle \rangle \quad (18)$$

$$G \equiv \langle\langle \text{male ? ? ?} \rangle \langle ? ? ? ? \rangle, \langle\langle ? ? ? ? \rangle \langle ? ? ? ? US \rangle \rangle \quad (19)$$

v. + $\langle\langle \text{male brown tall Irish} \rangle \langle \text{female brown short Irish} \rangle \rangle$

$$S \equiv \langle\langle \text{male brown ? ?} \rangle \langle \text{female ? short ?} \rangle \rangle \quad (20)$$

$$G \equiv \langle\langle \text{male ? ? ?} \rangle \langle ? ? ? ? \rangle \rangle \quad (21)$$

b) $2^8 = 256$

c) Each time specify a query with one attribute with mark ‘-’ would reduce the general hypothesis by half or some part.

d)