

6. $\therefore f_{\text{MKT}}(\vec{x}) = \text{sign}(P(y=+1 | \vec{x}) - \alpha)$

\therefore if $\begin{cases} P(y=+1 | \vec{x}) > \alpha, \text{ then } f_{\text{MKT}}(\vec{x}) = +1 \\ P(y=+1 | \vec{x}) \leq \alpha, \text{ then } f_{\text{MKT}}(\vec{x}) = -1 \end{cases}$

		\vec{y}	
		+1	-1
\vec{y}	+1	0	① FP.
	-1	② FN	0

If we consider the points on the decision boundary, the expected loss for assigning any such \vec{x} to either side should be the same (because the decision boundary is the place where we believe to have same probability to classify points to both sides)

Therefore, for \vec{x} on the boundary, we have that equality holds:

(FN) $\underbrace{10}_{\text{the probability of having the point } \vec{x} \text{ (which is on the boundary)}}$ $P(y=+1 | \vec{x}) = 1 \cdot \underbrace{P(y=-1 | \vec{x})}_{\text{to have true label } y=-1}$

$\Rightarrow 10 P(y=+1 | \vec{x}) = 1 \cdot (1 - P(y=+1 | \vec{x}))$

$\Rightarrow 10 P(y=+1 | \vec{x}) = 1 - P(y=+1 | \vec{x})$

$\Rightarrow 11 P(y=+1 | \vec{x}) = 1$

$\Rightarrow P(y=+1 | \vec{x}) = \frac{1}{11}$

Therefore, we assign α as $\alpha = \frac{1}{11}$, making $f_{\text{MKT}}(\vec{x}) = \text{sign}(P(y=+1 | \vec{x}) - \frac{1}{11})$ \square

because this means:

if $P(y=+1 | \vec{x}) > \frac{1}{11}$, then $f_{\text{MKT}}(\vec{x}) = 1$

\rightarrow we only need the probability to be more than $\frac{1}{11}$ (but not 0.5)

to make the model classify \vec{x} to be a member (+1)

\rightarrow we classify a customer to be a member more often because the cost of FN is greater.