# Skill, ROC curves, and optimal forecast combination

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# Probability forecastMedical diagnosisraw $\widetilde{X} \in [0, 1]$ $\widetilde{X} \in \Re$ decision $X = I(\widetilde{X} > \gamma)$ $X = I(\widetilde{X} > x_c)$

**POP** 50% chance rain Appendicitis white blood count

Loss ( $k_{YX}$ ) matrix

$$\begin{array}{c|c} Y = 1 & Y = 0 \\ X = 1 & 0 & \gamma \\ X = 0 & 1 - \gamma & 0 \end{array}$$

**Optimal Naive Forecast** 

$$\theta_y = P(Y = 1)$$

$$ONF = I(\theta_y > \gamma) \equiv 0$$

### Model

$$P(y, x) = \theta_{yx} = \theta_{y|x} \theta_x$$

$$L(y, x|\theta) = \theta_{1|1}^{yx} (1 - \theta_{1|1})^{(1-y)x} \times \theta_{0|0}^{(1-y)(1-x)} (1 - \theta_{0|0})^{y(1-x)} \times \theta_{x}^{x} (1 - \theta_{x})^{1-x}$$

$$p(\theta_{1|1}, \ \theta_{0|0}, \theta_x) = p(\theta_{1|1})p(\theta_{0|0})p(\theta_x)$$
  
~ "Be\_{\theta\_{1|1}}(0,0) × Be\_{\theta\_{0|0}}(0,0) × Be\_{\theta\_x}(0,0)"

$$p(\theta_{1|1}|y,x) \sim Be(n_{11},n_{01})$$

$$p(\theta_{0|0}|y,x) \sim Be(n_{00},n_{10})$$

$$p(\theta_{x}|y,x) \sim Be(n_{.1},n_{.0})$$

$$E(\theta_{1|1}|y,x) = \frac{n_{11}}{n_{11} + n_{01}}$$

$E(\theta_{\alpha \alpha} y x)$	_	<u> </u>
L(0 0 9,x)	—	$n_{00} + n_{10}$
		$n_{11} + n_{01}$

$$E(\theta_x|y,x) = \frac{n_{11} + n_{01}}{n_{\cdots}}$$

### Skill score

$$\begin{array}{c|c} Y = 1 & Y = 0 \\ X = 1 & 0 & \gamma \\ \text{Loss matrix} & X = 0 & 1 - \gamma & 0 \end{array}$$

$$E(k^{ONF}) = \theta_y(1-\gamma)$$
  
=  $\left(\theta_{1|1}\theta_x + (1-\theta_{0|0})(1-\theta_x)\right)(1-\gamma)$ 

 $E(k^E) = \theta_{1|1}\theta_x\gamma + (1-\theta_{0|0})(1-\theta_x)(1-\gamma)$ 

$$K_{\gamma} = \frac{E(k^{ONF}) - E(k^{E})}{E(k^{ONF})}$$
$$= \frac{\theta_{x}(\theta_{1|1} - \gamma)}{\theta_{y}(1 - \gamma)}$$

**Skill** : 
$$K_{1/2} > 0$$
 **Value** :  $K_{\gamma} > 0$ 

$$P(K_{\gamma} > 0|y, x) \longleftrightarrow P(\theta_{1|1} > \gamma|y, x)$$

$$E(K_{\gamma}|y,x) = \frac{n_{11}(1-\gamma) - n_{01}\gamma}{(n_{11}+n_{10})(1-\gamma)}$$

$$E(K_{1/2}|y,x) = \frac{n_{11} - n_{01}}{n_{11} + n_{10}}$$

### **Example 1**: Mammograms

$$\begin{array}{c|c} Y = 1 & Y = 0 \\ X = 1 & 7 & 70 \\ X = 0 & 1 & 922 \end{array}$$

Accuracy 
$$\left(X^{ONF}\right) = 97\%$$
 Accuracy  $\left(X^{E}\right) = 93\%$   
 $\widehat{K}_{1/2} = -7.9$   $\left(\gamma < \frac{1}{11} \to K_{\gamma} > 0\right)$ 

 $P(K_{1/2} > 0|y, x) = P(\theta_{1|1} > 1/2|y, x) \approx 10^{-15}$ 

Continuous forecasts skill score

$$F_j = P(Y = j, X \le x_c)$$

$$K_{\gamma,x_c} = \frac{(1-\gamma)F_1 - \gamma F_0}{\theta_y(1-\gamma)}I_0 +$$

$$\frac{-(1-\gamma)F_1+\gamma F_0}{(1-\theta_y)\gamma}(1-I_0)$$

$$(I_0 = I(\theta_y \le \gamma))$$

**Optimal decision point** 

$$x_* = \left\{ x : \frac{f(x|Y=1)}{f(x|Y=0)} = \frac{\gamma}{1-\gamma} \frac{1-\theta_y}{\theta_y} \right\}$$

- $\equiv$  Bayesian classifier
- $\equiv$  ROC curve decision

## X = I(White Blood Count > $x_c$ )

= 1Cut open= 0Send home













**Example 3**: Combine Two Forecasts:  $A, B \in [0, 1]$  $X = I(wA + (1 - w)B > \gamma)$ (A = Humans, B = MOS)If  $A, B \in \{0, 1\}, \qquad K_B > K_A$ Then

$$w = 0, \qquad X \equiv B$$









Y=1

![](_page_21_Figure_0.jpeg)

![](_page_22_Figure_0.jpeg)

Best w

![](_page_23_Figure_0.jpeg)

**Actual forecast** 

$$X = I(wA + (1 - w)B > \gamma)$$
  
$$X = I\left(B > \frac{\gamma}{(1 - w)} - \frac{w}{(1 - w)}A\right)$$

![](_page_25_Figure_0.jpeg)

 $\mathbf{X}_{\mathsf{A}}$ 

![](_page_26_Figure_0.jpeg)

 $\mathbf{x}^{\mathsf{B}}$ 

![](_page_26_Figure_2.jpeg)