

$$\text{loss} = D * LHD$$

$$\text{class} = cPD * cLHD$$

$$EL = PD * ELHD$$

$$\text{Dcorr} = \frac{PDJ - PD_1 PD_2}{\sqrt{PD_1(1-PD_1)PD_2(1-PD_2)}} = \frac{\text{Cov}(D_1, D_2)}{\sqrt{\text{Var}(D_1)\text{Var}(D_2)}}$$

$$PDJ_{1,2} = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \phi_2(z_1, z_2, \rho) dz_2 dz_1$$

$$\text{Bayes} \Rightarrow P(A|B) = \frac{P(A \cap B)}{P(B)}$$

$$\text{Range of } PDJ_{1,2,3} \Rightarrow [0, \min(PDJ_{i,j})] \quad i, j \in \{1, 2, 3\}$$

$$EL = IE(\text{class}) = \sum \text{class} * \text{prob.}, \text{ where: } \text{class} = cPD * cLHD$$

$$ELHD = \frac{EL}{PD}, \text{ Note: } IE(cLHD) < ELHD.$$

$$\begin{aligned} \text{Cov}(D_1, D_2) &= E[(X - EX)(Y - EY)] = E[XY] - E[X]E[Y] \\ &= E[D_1 D_2] - E[D_1]E[D_2] \\ &= PDJ - PD_1 PD_2 \end{aligned}$$

Bernoulli trials!!!

$$PD = IE[cPD] = \sum cPD * \text{prob. state.}$$

$$cLHD = IE(cLHD) = \sum cLHD * \text{prob. state.}$$