

$$P_x [N(\mu, \sigma^2) = x] \\ = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x-\mu}{\sigma} \right)^2}$$

$$= \frac{1}{\sigma} \left[\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x' - 0}{1} \right)^2} \right] \quad \text{Let } x' = \frac{x-\mu}{\sigma}$$

$$= \frac{1}{\sigma} P_x [N(0, 1) = x']$$

$$= \frac{1}{\sigma} P_N(x')$$

$$X_A = T_A + E_A \quad X_B = T_B + E_B \quad (1)$$

$$D = T_B - T_A$$

$$\text{Let } \gamma = \text{Exp}[E_A] - \text{Exp}[E_B]$$

$$P_c[d] = P_c[D \in [d, d+dd]] / dd$$

= height of probability distribution

$$P_c[d \text{ and } X_A = X_B] = P_c[X_A = X_B \mid d] \times P_c[d]$$

$$P_c[X_A = X_B]$$

$$P_c[d]$$

$$D = T_B - T_A$$

$$= N(0, 2\sigma_T^2)$$

$$= P_c[N(0, 2\sigma_T^2) = d]$$

$$= \frac{1}{\sqrt{2}\sigma_T} P_c \left(\frac{d}{\sqrt{2}\sigma_T} \right)$$

$$P_c[X_A = X_B] = P_c[X_B - X_A = 0]$$

$$X_B - X_A = \cancel{T_B} + E_B [T_B - T_A] + [E_B - E_A]$$

$$\text{Exp}[X_B - X_A] = 0 + \gamma$$

$$\text{Var}[X_B - X_A] = 2\sigma_T^2 + 2\sigma_E^2$$

$$= \frac{1}{(\sqrt{2}\sigma_T + \sqrt{2}\sigma_E)} P_c \left(\frac{\gamma}{(\sqrt{2}\sigma_T + \sqrt{2}\sigma_E)} \right)$$

(2)

$$P_0 [X_A = X_B \mid \cancel{T_B} - T_A = d]$$

$$X_B - X_A = \underbrace{[T_B - T_A]}_{\text{Fixed to } d} + [E_B - E_A]$$

$$Exp = d + r$$

$$Var = 2\sigma_E^2$$

$$\text{Need } X_B - X_A = 0$$

$$= P_0 \left[N(d+r, 2\sigma_E^2) = 0 \right]$$

$$= \frac{1}{\sqrt{2\sigma_E}} P_N\left(\frac{d+r}{\sqrt{2\sigma_E}}\right)$$

$$P_0 [d \mid X_A = X_B] = \frac{1}{\sqrt{2\sigma_E}} P_N\left(\frac{d+r}{\sqrt{2\sigma_E}}\right) \times \frac{1}{\sqrt{2\sigma_r}} P_N\left(\frac{d}{\sqrt{2\sigma_r}}\right)$$

$$\frac{1}{\sqrt{2\sigma_r + 2\sigma_E}} P_N\left(\frac{r}{\sqrt{2\sigma_E + 2\sigma_r}}\right)$$

$$e^{-\frac{1}{2} (d+r)^2 + d^2 - \left(\frac{\sigma}{2}\right)^2} = e^{-\frac{1}{2} \left[2d^2 + 2dr + \frac{3}{4} r^2 \right]}$$

$$\cancel{e} \sigma = \sqrt{\frac{2}{3}}$$

$$P_r[d | X_A = X_B] = e^{-\frac{1}{2} [2d^2 + 2dr + \frac{3}{4}r^2]} \quad (3)$$

$$= e^{-\frac{1}{2} [2(d + \frac{r}{2})^2]}$$

$$2(d + \frac{r}{2})^2 \neq \therefore = 2d^2 + 2dr + \frac{3}{4}r^2$$

$$\cancel{2d^2} + \cancel{2dr} + \frac{r^2}{4} \quad \therefore =$$

$$\therefore = (\frac{r}{2})^2$$

$$= e^{-\frac{1}{2} \left[\left(\sqrt{2}d + \frac{r}{\sqrt{2}} \right)^2 \right]} \times \underbrace{e^{-\frac{1}{2} \left(\frac{r}{2} \right)^2}}_{\text{constant}}$$

$$e^{-\frac{1}{2} \left(\frac{d + r/2}{1/\sqrt{2}} \right)^2}$$

$$D' \sim \text{Norm} \left(-\frac{r}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right)$$

$$P_r[T_B > T_A | X_A = X_B]$$

$$P_r[D' > 0] = P_r \left[\text{Norm} \left(-\frac{r}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right) > 0 \right]$$

$$\propto \frac{1}{2}$$

o.p

Jeff's model

Let $T_A \sim T_B \sim T$

Let Error be an arbitrary distribution with mean zero

Let $X = \text{Grade}_A(T + \text{Error})$
 $= \text{Grade}_B(T + \text{Error})$

←
The Grade B would be fixed to be the same

Karan's model

$T_A \sim T_B \sim T$

Let $E_A + E_B$ be arbitrary distribution

Without loss of generality, ~~let~~

$E_A \approx EA(\text{Error})$
where EA is some bijective function

$E_B \approx EB(\text{Error})$

Let $X = \text{Outcome}(T, E)$

$= \text{Grade}_A(T + EA^{-1}(E))$

→ Unless the 10^{-100} digit of T is always even
of EA " " even
 EB " " odd

Then outcome can know A vs B

Kara's model
is more General

$$Eq x = Outcome(T, E) = T \times E$$

$$Jeff's \quad x = GradeA(T + E)$$

can't compute $T \times E$
from $T + E$

~~notions~~