**Second Question:** Again assume you are given arbitrary cdfs  $F_X$  and  $F_Y$  describing random variables X and Y. Find a nontrivial *lower* bound to  $\overline{d}(F_X, F_Y)$  which depends only on  $\sigma_X^2$  and  $\sigma_Y^2$ .

**Solution:** The trivial lower bound is 0, since the expected value of the square of a real random variable is nonnegative. As before we know that

$$E[(X - Y)^{2}] = E(X^{2}) + E(Y^{2}) - 2E(XY) = \sigma_{X}^{2} + \sigma_{Y}^{2} - 2E(XY)$$

so to get a lower bound to  $\overline{d}(F_X, F_Y)$  we want an upper bound to the correlation E(XY). One of the most important bounds in probability has exactly this form. The Cauchy-Schwartz inequality states that

$$\mid E(XY) \mid \leq \sqrt{E(X^2)E(Y^2)} = \sigma_X \sigma_Y$$

so that for any joint cdf with the required marginals,

$$E[(X - Y)^2] \ge \sigma_X^2 + \sigma_Y^2 - 2\sigma_X\sigma_Y = (\sigma_X - \sigma_Y)^2$$

and hence

$$\overline{d}(F_X, F_Y) \ge (\sigma_X - \sigma_Y)^2$$

is the desired lower bound.

A few people used the equivalent fact that the correlation coefficient  $E(XY)/\sigma_X\sigma_Y$  has magnitude less than 1.