## Ralph's Normal Fallacy

Ralph is faced with two decisions and can acquire either of two, both, or no costly information systems. Each decision entails making a point estimate, $\widehat{x}_{j}$, regarding the eventual value of an outcome, $x_{j}$, where $j=1$ or 2 . Ralph's knowledge of the first variable is

$$
x_{1}=\mu_{1}+k \varepsilon_{1}+\varepsilon_{2}
$$

and the second variable is

$$
x_{2}=\mu_{2}+\varepsilon_{1}+\varepsilon_{2}
$$

where $\mu_{1}, \mu_{2}$, and $k$ are known constants and $\varepsilon_{1}$ and $\varepsilon_{2}$ are mean zero normal random variables with variances, $\sigma_{1}^{2}$ and $\sigma_{2}^{2}$, respectively. The first information system reports $\varepsilon_{1}$ at cost $c_{1}$ while the second information system reports $\varepsilon_{2}$ at $\operatorname{cost} c_{2}$. For simplicity, the cost of decision error is $E\left[\left(\sum_{j=1}^{2} x_{j}-\widehat{x}_{j}\right)^{2}\right]$ where $\widehat{x}_{j}$ is our best guess for $x_{j}$ given the information for decision $j=1$ or 2 . Hence, if both information sources are acquired, perfect prediction is possible and the $\operatorname{cost}$ is $c_{1}+c_{2}$. While if neither information sources are acquired, the cost is

$$
\begin{aligned}
& L^{T}\left[\begin{array}{cccc}
k^{2} \sigma_{1}^{2}+\sigma_{2}^{2}+2 k \sigma_{12} & k \sigma_{1}^{2}+\sigma_{2}^{2}+(1+k) \sigma_{12} & k \sigma_{1}^{2}+\sigma_{12} & k \sigma_{12}+\sigma_{2}^{2} \\
k \sigma_{1}^{2}+\sigma_{2}^{2}+(1+k) \sigma_{12} & \sigma_{1}^{2}+\sigma_{2}^{2}+2 \sigma_{12} & \sigma_{1}^{2}+\sigma_{12} & \sigma_{12}+\sigma_{2}^{2} \\
k \sigma_{1}^{2}+\sigma_{12} & \sigma_{1}^{2}+\sigma_{12} & \sigma_{1}^{2} & \sigma_{12} \\
k \sigma_{12}+\sigma_{2}^{2} & \sigma_{12}+\sigma_{2}^{2} & \sigma_{12} & \sigma_{2}^{2}
\end{array}\right] L \\
& \quad=k^{2} \sigma_{1}^{2}+\sigma_{2}^{2}+2 k \sigma_{12}+\sigma_{1}^{2}+\sigma_{2}^{2}+2 \sigma_{12}+2\left(k \sigma_{1}^{2}+\sigma_{2}^{2}+(1+k) \sigma_{12}\right) \\
& \\
& =(k+1)^{2} \sigma_{1}^{2}+4 \sigma_{2}^{2}+2(2+2 k) \sigma_{12}
\end{aligned}
$$

where $L^{T}=\left[\begin{array}{llll}1 & 1 & 0 & 0\end{array}\right]$ is the transpose of $L$, a column vector, and $\sigma_{12}$ refers to the covariance between $\varepsilon_{1}$ and $\varepsilon_{2}$. Since $\varepsilon_{1}$ and $\varepsilon_{2}$ are normally distributed random variables so are $x_{1}$ and $x_{2}$. Let $x=\left[\begin{array}{l}x_{1} \\ x_{2}\end{array}\right]$, Ralph knows

$$
\begin{aligned}
E\left[x \mid \varepsilon_{1}=e_{1}\right] & =\left[\begin{array}{c}
\mu_{1}+k e_{1}+\frac{\sigma_{12}}{\sigma_{1}^{2}} e_{1} \\
\mu_{2}+e_{1}+\frac{\sigma_{12}}{\sigma_{1}^{2}} e_{1}
\end{array}\right] \\
\operatorname{Var}\left[x \mid \varepsilon_{1}\right] & =\left[\begin{array}{ll}
\sigma_{2}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{2}^{2}} & \sigma_{2}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{2}^{2}} \\
\sigma_{2}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{1}^{2}} & \sigma_{2}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{1}^{2}}
\end{array}\right]
\end{aligned}
$$

Hence, the cost of decision error given $\varepsilon_{1}$ is acquired is

$$
\begin{aligned}
& c_{1}+\left[\begin{array}{ll}
1 & 1
\end{array}\right]\left[\begin{array}{ll}
\sigma_{2}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{1}^{2}} & \sigma_{2}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{1}^{1}} \\
\sigma_{2}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{1}^{2}} & \sigma_{2}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{1}^{2}}
\end{array}\right]\left[\begin{array}{l}
1 \\
1
\end{array}\right] \\
= & c_{1}+4\left(\sigma_{2}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{1}^{2}}\right)
\end{aligned}
$$

$$
\begin{aligned}
E\left[x \mid \varepsilon_{2}=e_{2}\right] & =\left[\begin{array}{c}
\mu_{1}+e_{2}+k \frac{\sigma_{12}}{\sigma_{2}^{2}} e_{2} \\
\mu_{2}+e_{2}+\frac{\sigma_{12}}{\sigma_{2}^{2}} e_{2}
\end{array}\right] \\
\operatorname{Var}\left[x \mid \varepsilon_{2}\right] & =\left[\begin{array}{cc}
k^{2}\left(\sigma_{1}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{2}^{2}}\right) & k\left(\sigma_{1}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{2}^{2}}\right) \\
k\left(\sigma_{1}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{2}^{2}}\right) & \sigma_{1}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{2}^{2}}
\end{array}\right]
\end{aligned}
$$

Hence, the cost of decision error given $\varepsilon_{2}$ is acquired is

$$
\begin{aligned}
& c_{2}+\left[\begin{array}{ll}
1 & 1
\end{array}\right]\left[\begin{array}{cc}
k^{2}\left(\sigma_{1}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{2}^{2}}\right) & k\left(\sigma_{1}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{2}^{2}}\right) \\
k\left(\sigma_{1}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{2}^{2}}\right) & \sigma_{1}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{2}^{2}}
\end{array}\right]\left[\begin{array}{l}
1 \\
1
\end{array}\right] \\
= & c_{2}+(k+1)^{2}\left(\sigma_{1}^{2}-\frac{\sigma_{12}^{2}}{\sigma_{2}^{2}}\right)
\end{aligned}
$$

Consider the following six cases.

| case | $k$ | $\sigma_{1}^{2}$ | $\sigma_{2}^{2}$ | $\sigma_{12}$ | $c_{1}$ | $c_{2}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 99 | 1 | 1 | 0 | 10 | 10 |
| 2 | 1 | 9 | 9 | 0 | 10 | 10 |
| 3 | 0 | 9 | 9 | 0.9 | 5 | 20 |
| 4 | 0 | 9 | 9 | 9 | 12 | 15 |
| 5 | 1 | 9 | 9 | 9 | 7 | 8 |
| 6 | 1 | 9 | 9 | 0 | 40 | 40 |

Required:

1. For cases $1-6$, what information systems, if any, are acquired and what is the total (information and decision error) cost considering both information systems and their full decision implications?
2. What does this example suggest about viewing accounting choice in isolation (of other information and other decisions)?
