

CHAPTER 10

→ Appendix

↳ Basic Forecasting Formula

$$\hat{r} = E\{r\} + \text{Cor}\{r, g\} \cdot \text{Var}^{-1}\{g\} \cdot (g - E\{g\}) \quad (10A.1)$$

→ \hat{r} is an unbiased estimate of r

→ If $\{r, g\}$ have a normal distribution, then:

○ \hat{r} is the max. likelihood estimate of r given g

↳ Technical treatment of Examples

→ Consider one asset with K forecasts $g = [g_1, g_2, \dots, g_K]$

$$\text{Var}\{g\} = \begin{bmatrix} \text{std}\{g_1\} & 0 \\ 0 & \text{std}\{g_K\} \end{bmatrix} \cdot p_g \cdot \begin{bmatrix} \text{std}\{g_1\} & 0 \\ 0 & \text{std}\{g_K\} \end{bmatrix} \quad (10A.18)$$

→ Covariance b/w K signals will involve K ICs

$$\text{Cor}\{r, g\} = \omega \cdot [IC_1, \dots, IC_K] \cdot \begin{bmatrix} \text{std}\{g_1\} & 0 \\ 0 & \text{std}\{g_K\} \end{bmatrix} \quad (10A.19)$$

→ Input into basic forecasting formula

$$\phi = \text{Cor}\{r, g\} \cdot \text{Var}^{-1}\{g\} \cdot (g - E\{g\})$$

$$= \omega \cdot [IC_1, \dots, IC_K] \cdot p_g^{-1} \cdot \begin{bmatrix} \frac{1}{\text{std}\{g_1\}} & 0 \\ 0 & \frac{1}{\text{std}\{g_K\}} \end{bmatrix} \cdot \begin{bmatrix} g_1 - E\{g_1\} \\ \vdots \\ g_K - E\{g_K\} \end{bmatrix} \quad (10A.20)$$

simplifies to:

$$\phi = \omega \cdot IC^T \cdot p_g^{-1} \cdot z \quad (10A.21)$$

where $z_i = (g_i - E\{g_i\}) \cdot \text{std}^{-1}\{g_i\}$

