

## The Signal Extraction Problem

### I. A Regression Analogy

- A. Suppose that we have a sample of past values of  $x$  and  $y$  (expressed as deviations from means).
- B. Now suppose that today we observe  $x_t$ , but our interest is in predicting  $y_t$ .
- C. For the regression model:

$$y_t = bx_t + e_t$$

we would estimate  $b$ :

$$\hat{b} = \frac{\sum_t x_t y_t}{\sum_t x_t^2}$$
$$\hat{b} = \frac{\text{sample covariance}(x, y)}{\text{sample variance}(x)}$$

and we would forecast:

$$\hat{y}_t = \hat{b}x_t.$$

### II. Signal Extraction

- A. Now suppose that two random errors,  $e_t$  and  $u_t$  are to be drawn. These errors have the following properties:

$$E(e_t) = E(u_t) = E(e_t u_t) = 0$$
$$E(e_t^2) = \mathbf{s}_e^2 \quad E(u_t^2) = \mathbf{s}_u^2$$

- B. Assume that the distributional information presented above is known.
- C. Now suppose that  $e_t$  and  $u_t$  values are drawn, but only the sum,  $s_t = e_t + u_t$ , is observed.
- D. I want to forecast  $u_t$  given my knowledge of  $s_t$ .
  - 1. I know:

$$\text{Var}(s_t) = \text{Var}(\mathbf{e}_t + u_t) = \mathbf{s}_e^2 + \mathbf{s}_u^2$$

$$\text{Cov}(s_t, u_t) = E[(\mathbf{e}_t + u_t)u_t] = \mathbf{s}_u^2$$

2. For a “regression” of  $u_t$  on  $s_t$  we have:

$$\hat{b} = \frac{\text{Cov}(s_t, u_t)}{\text{Var}(s_t)}$$

$$\hat{b} = \frac{\mathbf{s}_u^2}{\mathbf{s}_e^2 + \mathbf{s}_u^2}$$

3. So we forecast:

$$u_t = \left( \frac{\mathbf{s}_u^2}{\mathbf{s}_e^2 + \mathbf{s}_u^2} \right) s_t$$