

## Solution to BIOS760: 2011 FALL SEMESTER MIDTERM EXAM II

1. (a) Let  $X_n$  and  $Y_n$  be independent normal distributions. Then  $X_n \rightarrow_d X_1$  and  $Y_n \rightarrow_d -X_1$  but  $X_n + Y_n \sim N(0, 2)$  does not converge to  $X_1 - X_1 = 0$ .
- (b) The characteristic function of  $(X_n, Y_n)$  is

$$\begin{aligned}\phi(t_1, t_2) &= E[\exp\{it_1X_n + it_2Y_n\}] = E[e^{it_1X_n}]E[e^{it_2Y_n}] \\ &\rightarrow_d E[e^{it_1X}]E[e^{it_2Y}] = E[\exp\{it_1X + it_2Y\}].\end{aligned}$$

Thus,  $(X_n, Y_n) \rightarrow_d (X, Y)$ . Since  $(x, y) \rightarrow x + y$  is continuous,  $X_n + Y_n \rightarrow_d X + Y$  follows from the continuous mapping theorem.

2. (a) This follows from  $E[\sum_{i=1}^n w_i Y_i / x_i] = \sum_{i=1}^n w_i E[Y_i] / x_i = \beta$ .
- (b) The variance is given as

$$\sum_{i=1}^n w_i^2 x_i / x_i^2 = \sum_{i=1}^n w_i^2 / x_i.$$

We minimize the above term subject to constraint  $\sum_{i=1}^n w_i = 1$ . The Lagrange-multiplier gives

$$w_i = x_i / \sum_{j=1}^n x_j.$$

- (c) For the optimal weights, we have

$$\sum_{i=1}^n w_i Y_i / x_i = \beta + \frac{\sum_{i=1}^n \epsilon_i \sqrt{x_i}}{\sum_{i=1}^n x_i}.$$

Since  $\max x_i / \sum_{j=1}^n x_j \rightarrow 0$ , by the weighted CLT,

$$\frac{\sum_{i=1}^n \epsilon_i \sqrt{x_i}}{\sqrt{\sum_{i=1}^n x_i}} \rightarrow_d N(0, 1).$$

Thus,

$$\left\{ \sum_{i=1}^n x_i \right\}^{1/2} \left[ \sum_{i=1}^n w_i Y_i / x_i - \beta \right] \rightarrow_d N(0, 1).$$