

2.51

- a) For  $h = 10^5$ , the following sample values are created

	Sample	Theoretical
$\mu(z_0)$	0.5985	0.6
$\mu(z_1)$	0.6035	0.6
$\tilde{C}_{00}$	0.2403	0.24
$\tilde{C}_{11}$	0.2393	0.24
$\tilde{C}_{01}$	0.0402	0.04

(Code appended)

- b) Since there is a finite number states one can calculate the cumulative probability simply by summing. If one now draws  $p \in [0,1]$  from a uniform distribution, it follows:

$$P(p \in [P_{\text{cum}}(X_{i-1}), P_{\text{cum}}(X_i)]) = P(X_i), X_i \in \Omega$$

Therefore we can use draws from  $[0,1]$  to generate random samples from our Probability distribution.

(pseudorandom)

2.5.4

$$\tilde{x}_i^k(t) = d_i^k(t) + c^k(t)$$

$\Rightarrow \tilde{x}_i^k(t) = \tilde{d}_i^k(t) + \tilde{c}^k(t)$ , with expected values (linear  
and  $d_i, d_j, c$  independent, follows

$$\Rightarrow \tilde{C}_{ij}(s) = \delta_{ij} \tilde{d}_d(s) + \tilde{\sigma}_c(s)$$

$$= \delta(s)[\alpha v + \delta_{ij}(1-\alpha)v]$$

$$\Rightarrow \tilde{C}_{ij}(\omega) = \mathbb{F}[\tilde{C}_{ij}(s)](\omega)$$

$$= e^{is} (\alpha v + \delta_{ij}(1-\alpha)v)$$

$$= (\alpha + \delta_{ij}(1-\alpha))v$$

$$= \begin{cases} \alpha & i \neq j \\ 1 & i = j \end{cases}$$

$$\Rightarrow K'_{ij} = \frac{\tilde{C}_{ij}(\omega)}{\sqrt{\tilde{C}_{ii}(\omega)\tilde{C}_{jj}(\omega)}}, \text{ for } i \neq j$$

$$= \alpha$$

2.5.2

$$\mathcal{Z}_2. \langle \tilde{x}_i^k(-\omega) \tilde{x}_j^k(\omega) \rangle_k = \mathbb{F}[\tilde{C}_{ij}(s)](\omega)$$

$$\text{Bew. } \langle \tilde{x}_i^k(-\omega) \tilde{x}_j^k(\omega) \rangle_k = \langle Sd + d\tilde{f} e^{i(t-f)\omega} \tilde{x}_i^k(t) \tilde{x}_j^k(f) \rangle_k$$

$$= \langle Sd + d\tilde{f} e^{-is\omega} \tilde{x}_i^k(t) \tilde{x}_j^k(t+s) \rangle_k$$

$$= Sd \int e^{-is\omega} \langle \langle \tilde{x}_i^k(t) \tilde{x}_j^k(t+s) \rangle_k \rangle_f$$

$$= \mathbb{F}[\langle \langle \tilde{x}_i^k(t) \tilde{x}_j^k(t+s) \rangle_k \rangle_f](\omega)$$

$$= \mathbb{F}[\langle \langle \tilde{C}_{ij}(t, t+s) \rangle_k \rangle](\omega) = \mathbb{F}[\tilde{C}_{ij}(s)](\omega) \square$$

2. S. 3

$$a) k'(w) = \frac{\langle \tilde{x}_i^k(w) \tilde{x}_j^k(w) \rangle_c}{\sqrt{\langle |\tilde{x}_i^k(w)|^2 \rangle_c \langle |\tilde{x}_j^k(w)|^2 \rangle_c}}$$

$$|k'(w)|^2 = |k'(w)|^2 = \frac{\langle \tilde{x}_i^k(w) \tilde{x}_i^k(w) \rangle_c \langle \tilde{x}_j^k(w) \tilde{x}_j^k(w) \rangle_c}{\langle \tilde{x}_i^k(w) \tilde{x}_i^k(w)^* \rangle_c \langle \tilde{x}_j^k(w) \tilde{x}_j^k(w)^* \rangle_c}$$

For a deterministic process:

$$\Rightarrow |k'(w)|^2 = \frac{\langle \tilde{x}^k \rangle_c = \tilde{x}^k = z}{\frac{\tilde{x}_i^k(w) \tilde{x}_j^k(w) \tilde{x}_i^k(w) \tilde{x}_j^k(w)^*}{\tilde{x}_i^k(w) \tilde{x}_i^k(w)^* \tilde{x}_j^k(w) \tilde{x}_j^k(w)^*}} = 1$$