

$$3. \quad \mathbb{E} [| \hat{\phi}_n(t) - \phi_x(t) |^2] = \frac{1}{n} (1 - \phi_x(t)^2)$$

$$= n^{-1} (1 - \phi_x(t)^2)$$

↳ the order of magnitude
is $O(n^{-1})$

Since $\hat{\phi}_n$ is unbiased, its quadratic risk is equal to its variance.

2.) Fix $t \in \mathbb{R}$, $\hat{\Phi}(t)$ is an unbiased estimator of $\Phi_X(t)$

$$\mathbb{E}[\hat{\Phi}_n(t)] = \frac{1}{n} \sum_{k=1}^n \mathbb{E}[e^{itX_k}] = \Phi_X(t)$$

Lemma 1.2

$$\mathbb{E}[\hat{\Phi}_n(t)] = \frac{1}{n} \sum_{k=1}^n \mathbb{E}[e^{itX_k}] = \frac{1}{n} \cdot n \cdot \Phi_X(t) = \Phi_X(t)$$

$$\mathbb{E}[\phi_n(t)^2]$$

$$= \mathbb{E}[\phi_n(t)\phi_n(-t)] = \mathbb{E}\left[\sum_{k=1}^n \sum_{j=1}^n e^{i(X_k - X_j)t}\right] \frac{1}{n^2}$$

when $k=j$, $e^{i(X_k - X_j)t} = 1$

$$\mathbb{E}\left[\sum_{k,j: k \neq j} e^{i(X_k - X_j)t} + \sum_{k=j} 1\right] \frac{1}{n^2} = \mathbb{E}\left[\sum_{k,j: k \neq j} e^{i(X_k - X_j)t} + n\right] \frac{1}{n^2}$$

$$= \frac{1}{n^2} \sum_{k \neq j} \mathbb{E}[e^{i(X_k - X_j)t}] + \frac{1}{n} =$$

for k in range(n)
 for j in range(n) \Rightarrow $n(n-1)$
 if $j = k \pm 1$

$$= \frac{(n-1)}{n} |\phi(t)|^2 + \frac{1}{n}$$

$$\begin{aligned}
 \text{var}(\hat{\phi}_n(t)) &= \mathbb{E}[(\hat{\phi}_n(t) - \mathbb{E}[\hat{\phi}_n(t)])^2] \\
 &= \mathbb{E}[(\hat{\phi}_n(t) - \phi(t))^2] = \mathbb{E}[\hat{\phi}_n(t)^2] - \mathbb{E}[\hat{\phi}_n(t)]^2 \\
 &= \frac{n-1}{n} |\phi(t)|^2 + \frac{1}{n} - |\phi(t)|^2 = \left(1 - \frac{1}{n}\right) |\phi|^2 + \frac{1}{n} - |\phi|^2 \\
 &= -\frac{|\phi(t)|^2}{n} + \frac{1}{n} = \frac{1}{n} (1 - |\phi(t)|^2)
 \end{aligned}$$

$$\phi_X(t) = \mathbb{E}[e^{itx}] = \int_{\mathbb{R}} e^{itx} dF_X(x) = \int_{\mathbb{R}} e^{itx} f_X(x) dx$$

↑
characteristic
function

X is a real valued
random variable
defined on a prob.
space (Ω, Σ, P)

def. Expectation
real valued
random
variables

Lebesgue
Integral

1.)

$$\Phi_X(t) = \mathbb{E}[e^{itx}] = \int_{\mathbb{R}} e^{itx} p(t) dt$$

$$= \int_{\mathbb{R}} e^{itx} dF(t)$$

→ def of expectation
arbitrary real -
valued random
var.

prob
dens.
function

$$\Phi_n(\omega) = \int_{-\infty}^{\infty} e^{it\omega} dF_n(t) = \frac{1}{n} \sum_{j=1}^n e^{ix_j t}$$

The estimator is obtained by replacing the probability
function $F(t)$

by the empirical distribution function

$$(\Omega, F, P) \rightarrow \hat{F}_n(t) = \frac{1}{n} \sum_{i=1}^n \mathbb{1}(X_i \leq t) *$$