

9. Wiener Filter Theory

$$d = R s + n$$

$$d^i = R_x^i s^x + n^i$$

$$\mathcal{P}(n, s) = \mathcal{G}(s, S) \mathcal{G}(n, N)$$

$$\mathcal{P}(s|d) = \mathcal{G}(s - m, D)$$

$$m = Dj = D^{xy}j_y$$

$$D = (S^{-1} + \underbrace{R^\dagger N^{-1} R}_{=M})^{-1}$$

$$j = R^\dagger N^{-1} d$$

$$j_x = \bar{R}_x^i (N^{-1})_{ij} d^j$$

9.1 Statistical Homogeneity

- ▶ $s, d, n : \mathbb{R}^u \rightarrow \mathbb{R}, \mathbb{C}$
- ▶ complete data: $d = s + n$
- ▶ statistical homogeneous signal:

$$S^{xy} = \langle s^x s^y \rangle_{(s)} = C_s(x - y)$$

- ▶ statistical homogeneous noise:

$$N^{xy} = \langle n^x n^y \rangle_{(n)} = C_n(x - y)$$

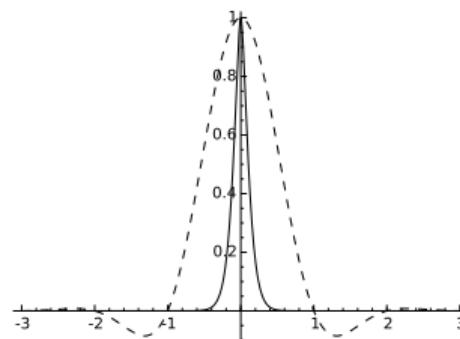


Figure: Possible correlation functions

9.2 Fourier Space

Fourier transform of a function $f : \mathbb{R}^u \rightarrow \mathbb{C}$:

$$\begin{aligned} f(\mathbf{k}) &= \int dx e^{2\pi i \mathbf{k} \cdot \mathbf{x}} f(x) \\ f(x) &= \int d\mathbf{k} e^{-2\pi i \mathbf{k} \cdot \mathbf{x}} f(\mathbf{k}) \end{aligned}$$

Absorb 2π -factor, $k = 2\pi|\mathbf{k}|$:

$$\begin{aligned} f_k = f(k) &= \int dx^u e^{ikx} f(x) \\ f_x = f(x) &= \int \frac{dk^u}{(2\pi)^u} e^{-ikx} f(k) \end{aligned}$$

Fourier Transformation Operator

Fourier transformation operator:

$$F_x^k = e^{ikx} \text{ applied via scalar product } a^\dagger b = \int dx \bar{a}_x b_x$$

$$(F^{-1})_k^x = e^{-ikx} \text{ applied via scalar product } a^\dagger b = \int \frac{dk}{(2\pi)^u} \bar{a}_k b_k$$

$$F^{-1} = F^\dagger \Rightarrow F \text{ orthonormal transformation}$$

9.3 Power Spectra

$$\begin{aligned}
 S^{xy} &= C_s(x - y) \text{ correlation translational invariant} = \text{homogeneous} \\
 S^{kk'} &= \left\langle s^k \bar{s}^{k'} \right\rangle_{(s)} = \left\langle (Fs)^k \overline{(Fs)}^{k'} \right\rangle_{(s)} = \left\langle (Fs)^k (Fs)^{\dagger k'} \right\rangle_{(s)} \\
 &= \left\langle (Fs)^k \left(s^{\dagger} F^{\dagger} \right)^{k'} \right\rangle_{(s)} = \left(F \left\langle s s^{\dagger} \right\rangle_{(s)} F^{\dagger} \right)^{kk'} = \left(F S F^{\dagger} \right)^{kk'} \\
 &= \left(F_x^k S^{xy} F_y^{\dagger k'} \right) \Big| \text{Einstein sum} = \int dx e^{ikx} \int dy S^{xy} e^{-ik'y} \\
 &= \int dx \int dy e^{i(kx - k'y)} C_s(\underbrace{x - y}_{=:r}) = \int dx \int dr e^{i(kx - k'(x - r))} C_s(r) \\
 &= \int dx e^{i(k - k')x} \int dr e^{ik'r} C_s(r) \\
 &= (2\pi)^u \delta(k - k') P_s(k)
 \end{aligned}$$

9.3.1 Units

- ▶ $[s^k] = [\int dx e^{ikx} s^x] = V [s^x]$
- ▶ $[C_s(r)] = [\langle s^x s^{x+r} \rangle] = [s^x]^2$
- ▶ $[P_s(k)] = [\int dr e^{ikr} C_s(r)] = V [s^x]^2 = \frac{[s^k]^2}{V}$
- ▶ $[\delta(k - k')] = \left[\frac{1}{k\text{-Volume}} \right] = V$

$$\Rightarrow P_s(k) = \frac{\langle |s^k|^2 \rangle}{V}$$

$$\begin{aligned} S^{kk'} &= (2\pi)^u \delta(k - k') P_s(k) = \langle s^k \bar{s}^{k'} \rangle_{(s)} = \mathbb{1}^{kk'} \frac{\langle |s^k|^2 \rangle}{V} \\ \mathbb{1}^{kk'} &= (2\pi)^u \delta(k - k') \end{aligned}$$

9.3.2 Wiener-Khintchin Theorem

- ▶ statistical homogeneous signal s
- ▶ stationary auto-correlation $S^{xy} = \langle s^x \bar{s}^y \rangle_{(s)} = C_s(x - y)$

\Rightarrow diagonal covariance matrix in Fourier space

$$S^{kk'} = \langle s^k \bar{s}^{k'} \rangle_{(s)} = (2\pi)^u \delta(k - k') C_s(k)$$

$$P_s(k) = \lim_{V \rightarrow \infty} \frac{1}{V} \langle \left| \int_V dx s^x e^{ikx} \right|^2 \rangle_{(s)} = C_s(k)$$

9.3.2 Wiener-Khintchin Theorem

- ▶ statistical homogeneous noise n
- ▶ stationary auto-correlation $N^{xy} = \langle n^x \bar{n}^y \rangle_{(n)} = C_n(x - y)$

\Rightarrow diagonal covariance matrix in Fourier space

$$N^{kk'} = \langle n^k \bar{n}^{k'} \rangle_{(s)} = (2\pi)^u \delta(k - k') C_n(k)$$

$$P_n(k) = \lim_{V \rightarrow \infty} \frac{1}{V} \left\langle \left| \int_V dx n^x e^{ikx} \right|^2 \right\rangle_{(s)} = C_n(k)$$

9.3.2 Wiener-Khintchin Theorem

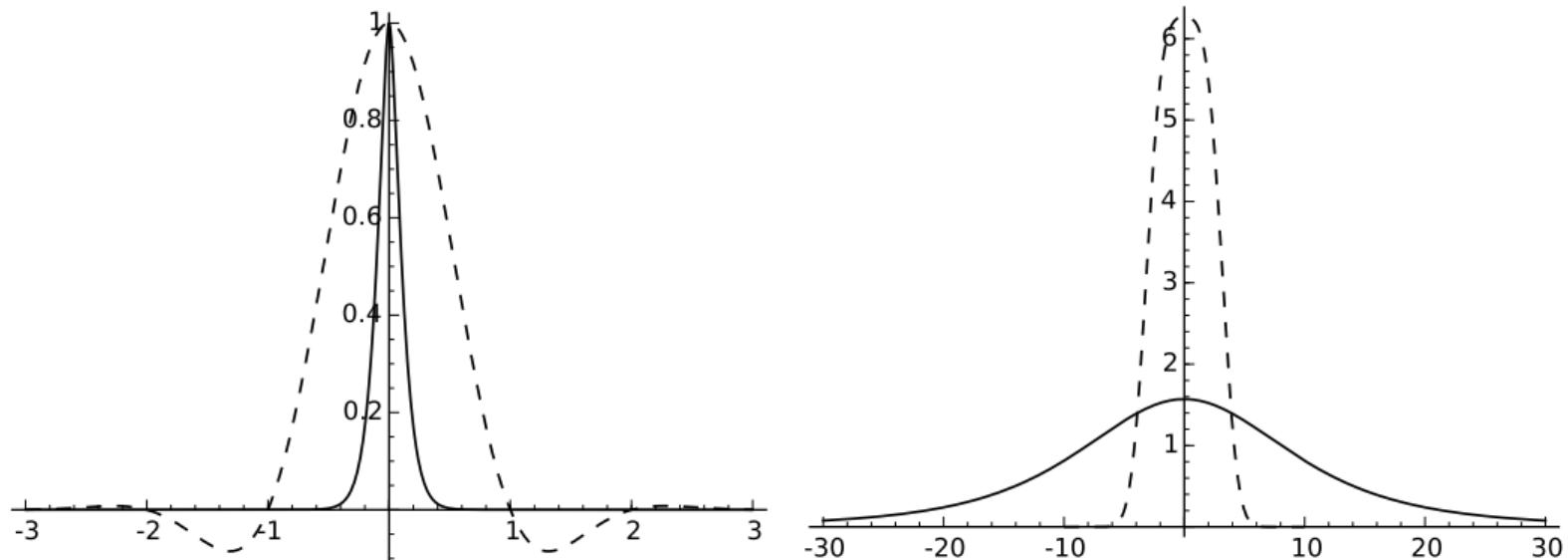


Figure: Possible correlation functions (left) and corresponding power spectra (right)

9.3.3 Fourier Space Filter

- ▶ Assume: $d = s + n$, $R = \mathbb{1}$, $\mathcal{P}(n, s) = \mathcal{G}(n, N) \mathcal{G}(s, S)$, $S^{xy} = C_s(x - y)$, $N^{xy} = C_n(x - y)$
- ▶ Gaussian posterior $\mathcal{P}(s|d) = \mathcal{G}(s - m, D)$
- ▶ mean $m = DN^{-1}d$
- ▶ variance $D = (S^{-1} + N^{-1})^{-1}$

Calculation of S^{-1} :

$$\begin{aligned}
 \mathbb{1}_q^k &= (\textcolor{blue}{S} \textcolor{red}{S^{-1}})_q^k \\
 (2\pi)^u \delta(k - q) &= \textcolor{blue}{S}^{kk'} (\textcolor{red}{S^{-1}})_{k'q} = \int \frac{dk'}{(2\pi)^u} (2\pi)^u \delta(k - k') P_s(k) (\textcolor{blue}{S^{-1}})_{k'q} \\
 &= P_s(k) (\textcolor{blue}{S^{-1}})_{kq} \\
 \Rightarrow (\textcolor{red}{S^{-1}})_{kq} &= \frac{(2\pi)^u \delta(k - q)}{P_s(k)}, \quad (N^{-1})_{kq} = \frac{(2\pi)^u \delta(k - q)}{P_n(k)} \\
 \Rightarrow M_{kq} &= (R^\dagger N^{-1} R)_{kq} = (N^{-1})_{kq}
 \end{aligned}$$

9.3.3 Fourier Space Filter

$$\begin{aligned}\Rightarrow D^{kq} &= (S^{-1} + \underbrace{R^\dagger N^{-1} R}_{=M})^{-1} e^{kq} \\ &= (2\pi)^u \delta(k - q) ([P_s(k)]^{-1} + [P_n(k)]^{-1})^{-1}\end{aligned}$$

$$\begin{aligned}\Rightarrow j_k &= (R^\dagger N^{-1} d)_k = (N^{-1} d)_k \\ &= \int \frac{dk'}{(2\pi)^u} (2\pi)^u \delta(k' - k) [P_n(k)]^{-1} d_{k'} \\ &= \frac{d_k}{P_n(k)}\end{aligned}$$

9.3.3 Fourier Space Filter

$$\begin{aligned}\Rightarrow m^k &= (Dj)^k = D^{kk'} j_{k'} \\ &= \int \frac{dk'}{(2\pi)^u} \frac{(2\pi)^u \delta(k - k')}{\frac{1}{P_s(k)} + \frac{1}{P_n(k)}} \frac{d^{k'}}{P_n(k')} \\ &= \underbrace{\frac{1}{1 + \frac{P_n(k)}{P_s(k)}}}_{f(k)=\text{filter function}} d^k\end{aligned}$$

Filter Function

$\Rightarrow f(k)$ reweights all Fourier modes of the data independently

$$f(k) = \frac{1}{1 + \frac{P_n(k)}{P_s(k)}}$$
$$= \begin{cases} 1 & \text{if } P_s(k) \gg P_n(k) \Rightarrow \text{perfect pass through} \\ \underbrace{\frac{P_s(k)}{P_n(k)}}_{\ll 1} & \text{if } P_s(k) \ll P_n(k) \Rightarrow \text{signal-to-noise weighting} \end{cases}$$

$$\Rightarrow m^k = f(k) d^k = f(k) (s^k + n^k) = \left(\frac{s + n}{1 + P_n/P_s} \right)^k$$

End