

$$\mu_1 = E(\bar{x}_1), \quad \mu_2 = E(\bar{x}_2), \quad \sigma_1^2 = \text{Var}(\bar{x}_1), \quad \sigma_2^2 = \text{Var}(\bar{x}_2)$$

$$\rho = \frac{E(\bar{x}_1 - \mu_1)(\bar{x}_2 - \mu_2)}{(E(\bar{x}_1 - \mu_1)^2)^{\frac{1}{2}} (E(\bar{x}_2 - \mu_2)^2)^{\frac{1}{2}}}$$

$$n: f_{\bar{x}}(x) = \frac{1}{(2\pi)^{\frac{n}{2}} (\det \Sigma)^{\frac{1}{2}}} \exp\left(-\frac{1}{2}(x - \mu)^T \Sigma^{-1}(x - \mu)\right), \quad \bar{x}, \mu \in \mathbb{R}^n, \quad \Sigma \in \mathbb{R}^{n \times n}$$

$$\Sigma = E(\bar{x} - \mu)(\bar{x} - \mu)^T, \quad \Sigma = \begin{pmatrix} \sigma_1^2 & \rho\sigma_1\sigma_2 \\ \rho\sigma_1\sigma_2 & \sigma_2^2 \end{pmatrix}^{-1}$$

$$\text{Characteristic Function. } \bar{x} \in \mathbb{R} \quad \phi_{\bar{x}}(w) = E(\exp(jw\bar{x}))$$

Gaussian Processes.

$$\bar{x}(t), \quad \forall n, \quad \forall t_1 \leq t_2 \leq \dots \leq t_n, \quad \bar{x} = (\bar{x}(t_1), \dots, \bar{x}(t_n))$$

$$\Leftrightarrow \bar{x} \sim N(\mu, \Sigma), \quad n\text{-dimensional Joint Gaussian Distribution}$$

$$n=1: f_{\bar{x}}(x) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$$

$$n=2: f_{\bar{x}_1, \bar{x}_2}(x_1, x_2) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp\left(-\frac{1}{2(1-\rho^2)}\left(\left(\frac{x_1-\mu_1}{\sigma_1}\right)^2 + \left(\frac{x_2-\mu_2}{\sigma_2}\right)^2 - 2\rho\left(\frac{x_1-\mu_1}{\sigma_1}\right)\left(\frac{x_2-\mu_2}{\sigma_2}\right)\right)\right)$$

$$(\bar{x}_1, \bar{x}_2) \sim N(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2, \rho)$$

$$\Phi_Z(\omega) = \int_{-\infty}^{\infty} \exp(i\omega x) f_Z(x) dx.$$

$$E(g(Z, Y)) = E(E(g(Z, Y)|Y))$$

$$P(A) = E(I_A).$$

Sum of Random Variables. Z, Y , independent, $f_Z(\omega), f_Y(\omega)$

$$Z = X + Y. \quad f_Z(z) = (f_X \otimes f_Y)(z) = \int_{-\infty}^{\infty} f_X(z-t) f_Y(t) dt.$$

$$F_Z(t) = P(Z \leq t) = P(\underline{X+Y} \leq t) = \int_{-\infty}^{\infty} P(X+y \leq t | Y=y) f_Y(y) dy$$

$$\Omega: A \subseteq \Omega \quad P(A) = \sum_{x \in A} P(x) = \sum_{\omega \in \Omega} I_A(\omega) P(\omega) = E(I_A)$$

$$F_Z(t) = \int_{-\infty}^{\infty} P(X \leq t-y) f_Y(y) dy = \int_{-\infty}^{\infty} F_X(t-y) f_Y(y) dy.$$

$$f_Z(t) = \frac{d}{dt} F_Z(t) = \frac{d}{dt} \int_{-\infty}^{\infty} F_X(t-y) f_Y(y) dy.$$

$$= \int_{-\infty}^{\infty} f_X(t-y) f_Y(y) dy = \boxed{(f_X \otimes f_Y)(t)}$$

$$\begin{aligned} \Phi_Z(\omega) &= E(\exp(i\omega Z)) = E(\exp(i\omega(X+Y))) = E(\exp(i\omega X) \exp(i\omega Y)) \\ &= E(\exp(i\omega X)) E(\exp(i\omega Y)) = \Phi_X(\omega) \Phi_Y(\omega) \end{aligned}$$

$$\Phi_x(\omega) = \int_{-\infty}^{+\infty} f_x(x) \exp(j\omega x) dx.$$

$$= \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2} + j\omega x\right) dx.$$

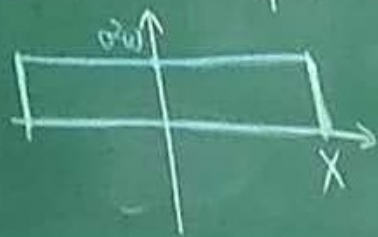
$$= \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{+\infty} \exp\left(-\frac{1}{2\sigma^2}(x^2 - 2\mu x + \mu^2) + j\omega x\right) dx$$

$$= \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{+\infty} \exp\left(-\frac{1}{2\sigma^2}(x^2 - 2\mu x - 2j\sigma^2\omega x + (\mu + j\sigma^2\omega)^2) + \frac{1}{2\sigma^2}(\mu + j\sigma^2\omega)^2 - \frac{1}{2\sigma^2}\mu^2\right) dx$$

$$= \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{+\infty} \exp\left(-\frac{(x - \mu - j\sigma^2\omega)^2}{2\sigma^2}\right) dx \cdot \exp(j\mu\omega - \frac{1}{2}\sigma^2\omega^2).$$

Cauchy Integral. $f(z)$ Analytic $\oint_C f(z) dz = 0$.

$$f(z) = \exp\left(-\frac{z^2}{2\sigma^2}\right), \quad \int_0^{j\sigma^2\omega} \exp\left(-\frac{(x+jy)^2}{2\sigma^2}\right) dy = \int_0^{j\sigma^2\omega} \exp\left(-\frac{x^2}{2\sigma^2}\right) \exp\left(-\frac{y^2}{2\sigma^2}\right) dy$$



$$= \left[\exp\left(-\frac{x^2}{2\sigma^2}\right) \int_0^{j\sigma^2\omega} \exp\left(-\frac{y^2}{2\sigma^2}\right) \exp\left(j\frac{xy}{\sigma^2}\right) dy \right]_{x=-\infty}^{x=\infty} \xrightarrow{x \rightarrow \infty} 0$$

Law of Large Number: X_1, \dots, X_n i.i.d. $E(X_1) = \mu$.

$$\frac{1}{n} \sum_{k=1}^n X_k \xrightarrow{n \rightarrow \infty} \mu.$$



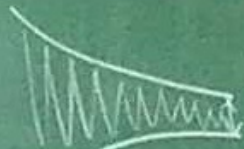
$$(1 + \frac{1}{n})^n \rightarrow e.$$

$$(1 + \frac{a}{n})^n \rightarrow e^{a\mu}$$

$$(1 + \frac{a}{n} + o(\frac{1}{n}))^n \rightarrow e^a$$

Central Limit Theorem: $E(X_k) = 0, \text{Var}(X_k) = 1$

$$\frac{X_1 + \dots + X_n}{\sqrt{n}} \xrightarrow{n \rightarrow \infty} N(0, 1)$$



$$\phi_{X_1}(\frac{\omega}{n}) = E(\exp(j\omega \frac{X_1}{n})) = E(1 + \frac{j\omega X_1}{n} + o(\frac{1}{n})) = 1 + \frac{j\omega \mu}{n} + o(\frac{1}{n}).$$

$$\begin{aligned} \phi_{\frac{X_1 + \dots + X_n}{n}}(\omega) &= E(\exp(j\omega (\frac{X_1 + \dots + X_n}{n}))) \\ &= \prod_{k=1}^n E(\exp(j\omega \frac{X_k}{n})) = \prod_{k=1}^n \phi_{X_k}(\frac{\omega}{n}) = (\phi_{X_1}(\frac{\omega}{n}))^n \\ &= (1 + \frac{j\omega \mu}{n} + o(\frac{1}{n}))^n \xrightarrow{n \rightarrow \infty} \exp(j\mu \omega) = \phi_{\mu}(\omega) \end{aligned}$$

$$\phi_{\frac{X_1 + \dots + X_n}{\sqrt{n}}}(\omega) = (\phi_{X_1}(\frac{\omega}{\sqrt{n}}))^n = (1 + \frac{j\omega E(X_1)}{\sqrt{n}} + \frac{1}{2} (\frac{j\omega X_1}{\sqrt{n}})^2 + o(\frac{1}{n}))^n$$

$$= \left(1 - \frac{E(\bar{x})^2 \omega^2}{2n} + o\left(\frac{1}{n}\right)\right)^n = \left(1 - \frac{\omega^2}{2n} + o\left(\frac{1}{n}\right)\right)^n \quad \bar{x} \sim \left[\begin{matrix} (- \\ F \end{matrix}\right]$$

$$\xrightarrow{n \rightarrow \infty} \exp\left(-\frac{\omega^2}{2}\right) = \phi_{\bar{x}}(\omega). \quad \bar{x} \sim N(0,1) \quad \text{Var}(a\bar{x}) = a^2 \text{Var}(\bar{x})$$

Maximum Entropy. $\Omega \rightarrow (\#\Omega) \rightarrow \log(\#\Omega)$ Boltzmann.

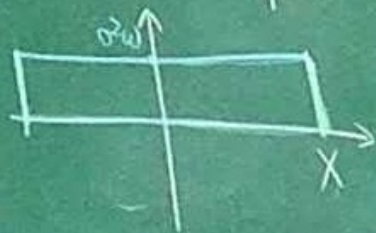
$$E(f) = \left[- \int_{-\infty}^{+\infty} f_{\bar{x}}(x) \log f_{\bar{x}}(x) dx \right] = E(\log f(\bar{x})). \quad \text{Shannon Randomness}$$

$$\max_f E(f), \quad \text{s.t.} \quad \int_{-\infty}^{+\infty} x f(x) dx = \mu, \quad \int_{-\infty}^{+\infty} x^2 f(x) dx = \sigma^2$$

$$= \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{+\infty} \exp\left(-\frac{(x-\mu-\frac{j\sigma^2\omega}{2})^2}{2\sigma^2}\right) dx \quad \boxed{\exp(j\mu\omega - \frac{1}{2}\sigma^2\omega^2)}$$

Cauchy Integral. $f(z)$ Analytic $\oint_{\mathcal{C}} f(z) dz = 0$. $\Phi_{\bar{x}}(\underline{\omega})$

$$f(z) = \exp\left(-\frac{z^2}{2\sigma^2}\right). \quad \int_0^{\sigma^2\omega} \exp\left(-\frac{(x+jy)^2}{2\sigma^2}\right) dy = \int_0^{\sigma^2\omega} \exp\left(-\frac{x^2}{2\sigma^2}\right) \exp\left(-\frac{y^2}{2\sigma^2}\right)$$



$$= \boxed{\exp\left(-\frac{x^2}{2\sigma^2}\right)} \int_0^{\sigma^2\omega} \exp\left(-\frac{y^2}{2\sigma^2}\right) \exp(j\frac{xy}{\sigma^2}) dy \xrightarrow{x \rightarrow \infty} \exp\left(-j\frac{x y}{\sigma^2}\right) dy$$

$$\tilde{G}(t) = \int_{-\infty}^{+\infty} (f_0 + tg) \log(f_0 + tg) dx + \lambda_1 \left(\int_{-\infty}^{+\infty} x(f_0 + tg) dx - \mu \right) + \lambda_2 \left(\int_{-\infty}^{+\infty} x^2(f_0 + tg) dx - \sigma^2 \right).$$

$$\frac{d}{dt} \tilde{G}(t) = \int_{-\infty}^{+\infty} g(x) (\log(f_0 + tg) + \lambda_1 x + \lambda_2 x^2 + 1) dx.$$

$$\left. \frac{d}{dt} \tilde{G}(t) \right|_{t=0} = \int_{-\infty}^{+\infty} \underline{g(x)} (\log(\underline{f_0}) + \lambda_1 x + \lambda_2 x^2 + 1) dx = 0$$

$$f_0(x) = \exp(-\lambda_2 x^2 - \lambda_1 x - 1)$$

Variational Technique.

$E(f)$. Functional. f_0 optimal Function. $\forall f$. $E(f) \leq E(f_0)$.

$G(t) = E(f_0 + tg)$. $\forall g$. $G(t) \leq G(0)$. $\forall t$.

$$\frac{d}{dt} G(t) = \frac{d}{dt} \int_{-\infty}^{+\infty} (f_0 + tg) \log(f_0 + tg) dx$$

$$= \int_{-\infty}^{+\infty} g \log(f_0 + tg) dx + \int_{-\infty}^{+\infty} g dx$$

$$\mathbf{x} \in \mathbb{R}^n. \quad \mathbf{x} = (x_1, \dots, x_n). \quad \Phi_{\mathbf{x}}(\omega) = E(\exp(j\omega^T \mathbf{x}))$$

$$\exp(j\mu\omega - \frac{1}{2}\sigma^2\omega^2) \rightarrow \exp(j\mu^T \omega - \frac{1}{2}\omega^T \Sigma \omega) \quad \omega \in \mathbb{R}^n.$$

$$f_{\mathbf{x}}(\mathbf{x}) = \frac{1}{(2\pi)^{\frac{n}{2}} (\det \Sigma)^{\frac{1}{2}}} \exp(-\frac{1}{2}(\mathbf{x} - \mu)^T \Sigma^{-1}(\mathbf{x} - \mu)) \quad \mathbf{x}, \mu \in \mathbb{R}^n$$

$$\int_{\mathbb{R}} f_{\mathbf{x}}(\mathbf{x}) d\mathbf{x} = \frac{1}{(2\pi)^{\frac{n}{2}} (\det \Sigma)^{\frac{1}{2}}} \int_{\mathbb{R}} \exp(-\frac{1}{2}(\mathbf{x}' - \mu)^T \Sigma^{-1}(\mathbf{x}' - \mu)) d\mathbf{x}$$

$$= \frac{1}{(2\pi)^{\frac{n}{2}} (\det \Sigma)^{\frac{1}{2}}} \int_{\mathbb{R}} \exp(-\frac{1}{2}(\mathbf{x}')^T \Sigma^{-1}(\mathbf{x}')) d\mathbf{x}'$$

$$= \left(1 - \frac{E(\mathbf{x}_1)^2 \omega^2}{2n} + o\left(\frac{1}{n}\right)\right)^n = \left(1 - \frac{\omega^2}{2n} + o\left(\frac{1}{n}\right)\right)^n \quad \mathbf{x} \sim \boxed{F}$$

$$\xrightarrow{n \rightarrow \infty} \exp\left(-\frac{\omega^2}{2}\right) = \Phi_{\mathbf{x}}(\omega). \quad \mathbf{x} \sim N(0, 1) \quad a\mathbf{x} \sim \boxed{F}$$

Maximum Entropy. $\Omega \rightarrow (\#\Omega) \rightarrow \log(\#\Omega)$ Boltzmann.

$\text{Var}(a\mathbf{x}) = a^2 \text{Var}(\mathbf{x})$

$$E(f) = \boxed{-\int_{-\infty}^{+\infty} f_{\mathbf{x}}(\mathbf{x}) \log f_{\mathbf{x}}(\mathbf{x}) d\mathbf{x}} = E(\log f(\mathbf{x})). \quad \text{Shannon Randomness}$$

$$\max_f E(f). \quad \text{s.t.} \quad \int_{-\infty}^{+\infty} x f(x) dx = \mu, \quad \int_{-\infty}^{+\infty} x^2 f(x) dx = \sigma^2$$

$$\Sigma = E(\bar{x} - \mu)(\bar{x} - \mu)^T. \quad \Sigma^T =$$

$$\Sigma^T = U^T \Lambda U. \quad U U^T = U^T$$

$$\Sigma^T = U^T \Lambda^{\frac{1}{2}} \Lambda^{\frac{1}{2}} U = L^T L.$$

$$y = Lx' = \underline{U^T \Lambda^{\frac{1}{2}} U \cdot U^T \Lambda^{\frac{1}{2}} U} = L^2$$

$$= \frac{1}{(2\pi)^{\frac{n}{2}} (\det \Sigma)^{\frac{1}{2}}} \int_{\mathbb{R}^n} \exp(-\frac{1}{2} y^T y) \frac{1}{|\det L|} dy$$

$$y = Lx'$$

$$dy = \left| \det \left(\frac{dy}{dx'} \right) \right| \cdot dx'$$

$$= |\det L| \cdot dx'$$

$$dx' = \frac{1}{|\det L|} dy \quad \Sigma^T = L^T L$$

$$\det \Sigma^T = (\det L)^2$$

$$(\det \Sigma)^{\frac{1}{2}} = \frac{1}{|\det L|}$$

$$= \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} \exp(-\frac{1}{2} y^T y) dy$$

$$= \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} \exp(-\frac{1}{2} \sum_{k=1}^n y_k^2) dy_1 \cdots dy_n$$

$$= \left(\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} \exp(-\frac{y^2}{2}) dy \right)^n = 1$$

$$\cdot \left(\int_{-\infty}^{+\infty} x (f_0 + tg) dx - \mu \right) \\ x^2 (f_0 + tg) dx - \sigma^2 \right)$$

$$\frac{dy}{dx'} = \frac{\partial(y_1, \dots, y_n)}{\partial(x'_1, \dots, x'_n)}$$

$$= \begin{pmatrix} \frac{\partial y_1}{\partial x'_1} & \cdots & \frac{\partial y_1}{\partial x'_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial y_n}{\partial x'_1} & \cdots & \frac{\partial y_n}{\partial x'_n} \end{pmatrix} = L$$

$$\begin{aligned}
 & \frac{1}{(2\pi)^{\frac{n}{2}} (\det \Sigma)^{\frac{1}{2}}} \int_{\mathbb{R}^n} \exp \left(-\frac{1}{2} (x-\mu)^T \Sigma^{-1} (x-\mu) + j\omega^T x \right) dx \cdot \overset{(AB)^T = B^T A^T}{=} (L^T L)^{-1} \\
 & \Sigma^{-1} = L^T L \quad y = L(x-\mu) \Rightarrow x = L^{-1}y + \mu \quad \Rightarrow \quad \overset{(L^{-1})^T}{=} (\Sigma^{-1})^{-1} \\
 & = \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} \exp \left(-\frac{1}{2} y^T y + j\omega^T (L^{-1}y + \mu) \right) dy = \Sigma \\
 & = \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} \exp \left(-\frac{1}{2} (y - j(L^{-1})^T \omega)^T (y - j(L^{-1})^T \omega) + j\omega^T \mu - \frac{1}{2} \omega^T (L^{-1}) (L^{-1})^T \omega \right) dy \\
 & = \exp(j\mu^T \omega - \frac{1}{2} \omega^T \Sigma \omega) \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} \exp \left(-\frac{1}{2} (y - j(L^{-1})^T \omega)^T (y - j(L^{-1})^T \omega) \right) dy
 \end{aligned}$$

$$\begin{aligned}
 & \bar{x} \in \mathbb{R}^n, \bar{x} = (\bar{x}_1, \dots, \bar{x}_n), \Phi_{\bar{x}}(\omega) = E(\exp(j\omega^T \bar{x})) \\
 & \exp(j\mu\omega - \frac{1}{2}\sigma^2\omega^2) \rightarrow \exp(j\mu^T \omega - \frac{1}{2}\omega^T \Sigma \omega), \omega \in \mathbb{R}^n. \\
 & f_{\bar{x}}(x) = \frac{1}{(2\pi)^{\frac{n}{2}} (\det \Sigma)^{\frac{1}{2}}} \exp \left(-\frac{1}{2} (x-\mu)^T \Sigma^{-1} (x-\mu) \right), x, \mu \in \mathbb{R}^n \\
 & \int_{\mathbb{R}^n} f_{\bar{x}}(x) dx = \frac{1}{(2\pi)^{\frac{n}{2}} (\det \Sigma)^{\frac{1}{2}}} \int_{\mathbb{R}^n} \exp \left(-\frac{1}{2} (x-\mu)^T \Sigma^{-1} (x-\mu) \right) dx \\
 & = \frac{1}{(2\pi)^{\frac{n}{2}} (\det \Sigma)^{\frac{1}{2}}} \int_{\mathbb{R}^n} \exp \left(-\frac{1}{2} (x')^T \Sigma^{-1} (x') \right) dx' = \frac{1}{(2\pi)^{\frac{n}{2}} (\det \Sigma)^{\frac{1}{2}}} \int_{\mathbb{R}^n} \exp \left(-\frac{1}{2} (x')^T L^T L^{-1} L x' \right) dx'
 \end{aligned}$$