

Gaussian Random Variables

A specific (very useful!) continuous random variable

$$X \sim \underline{N(\mu, \sigma^2)}$$

$$\text{density } \gamma_{\mu, \sigma}(x) = \frac{e^{-(x-\mu)^2/2\sigma^2}}{\sqrt{2\pi}\sigma}$$

Normal distribution
with mean μ , variance σ^2

$$X \sim N(0, 1)$$

$$\gamma(x) = \frac{e^{-x^2/2}}{\sqrt{2\pi}}$$

(Standard Normal Distribution)

$$P[X \in [a, b]] = \int_a^b \frac{e^{-x^2/2}}{\sqrt{2\pi}} dx$$

What's up with $\sqrt{2\pi}$?

$$I = \int_{-\infty}^{\infty} e^{-x^2/2}$$

► I is bounded.

Proof:
$$I = 2 \int_0^{\infty} e^{-x^2/2} dx = 2 \int_0^2 \underbrace{e^{-x^2/2}}_{\leq 1} dx + 2 \int_2^{\infty} \underbrace{e^{-x^2/2}}_{\leq e^{-x}} dx$$
$$\leq 2(1 + e^{-2})$$

► $I^2 = 2\pi$

Proof:
$$I^2 = \int_{-\infty}^{\infty} e^{-x^2/2} dx \int_{-\infty}^{\infty} e^{-y^2/2} dy = \iint e^{-(x^2+y^2)/2} dx dy$$
$$= \iint e^{-r^2/2} r dr d\theta = 2\pi$$

Transformations of Gaussians

Say $X \sim N(0, 1)$ $Y = \sigma \cdot X + \mu$ ($\sigma > 0$)

$$\begin{aligned} P[Y \leq t] &= P\left[X \leq \frac{t - \mu}{\sigma}\right] = \int_{-\infty}^{(t - \mu)/\sigma} \frac{e^{-x^2/2}}{\sqrt{2\pi}} dx \\ &= \int_{-\infty}^t \frac{e^{-(y - \mu)^2/2\sigma^2}}{\sqrt{2\pi} \sigma} dy \end{aligned}$$

$$\text{Density of } Y = \frac{e^{-(y - \mu)^2/2\sigma^2}}{\sqrt{2\pi} \sigma}$$

Multiple independent Gaussians

$X = (X_1, \dots, X_n)$ each $X_i \sim \mathcal{N}(0, 1)$ independent

$$p(x_1, \dots, x_n) = \frac{e^{-x_1^2/2}}{\sqrt{2\pi}} \cdot \dots \cdot \frac{e^{-x_n^2/2}}{\sqrt{2\pi}}$$

$$= \frac{e^{-(x_1^2 + \dots + x_n^2)/2}}{(2\pi)^{n/2}}$$

$$= \frac{e^{-\|x\|^2/2}}{(2\pi)^{n/2}}$$

Linear Combinations of Independent Gaussians

► X, Y independent
density f → density g

$$\begin{aligned} P[X + Y \leq t] &= \int_{-\infty}^{\infty} \int_{-\infty}^{t-x} f(x) g(y) dx dy \\ &= \int_{-\infty}^t \left(\underbrace{\int_{-\infty}^{\infty} f(x) g(z-x) dx}_{h(z)} \right) dz \end{aligned}$$

$z = x + y$

► $X_1, X_2 \sim N(0, 1)$ independent.

$$\begin{aligned} \text{density of } X_1 + X_2 &= \int_{-\infty}^{\infty} \frac{e^{-x^2/2}}{\sqrt{2\pi}} \frac{e^{-(z-x)^2/2}}{\sqrt{2\pi}} dx \\ &= \left(\int_{-\infty}^{\infty} \underbrace{e^{-x^2 + zx}}_{e^{-(x - \frac{z}{2})^2 + z^2/4}} dx \right) \cdot \frac{e^{-z^2/2}}{2\pi} = \frac{e^{-z^2/4}}{\sqrt{2} \cdot \sqrt{2\pi}} \end{aligned}$$

Ex: For $X_1, \dots, X_n$ i.i.d. $N(0, 1)$, show that $\sum c_i X_i \sim N(0, \|c\|^2)$

Gaussian Expectations

Let $X \sim \mathcal{N}(0, 1)$

$$\mathbb{E}[X] = \int_{-\infty}^{\infty} x \cdot r(x) dx = \int_{-\infty}^{\infty} x \cdot \frac{e^{-x^2/2}}{\sqrt{2\pi}} dx = 0 = \mathbb{E}[X^{2k+1}]$$

for any $k \in \mathbb{N}$

$$\mathbb{E}[X^2] = \int_{-\infty}^{\infty} x^2 \cdot \frac{e^{-x^2/2}}{\sqrt{2\pi}} dx = 1$$

$$\mathbb{E}[e^{\lambda X^2}] = \int_{-\infty}^{\infty} e^{\lambda x^2} \cdot \frac{e^{-x^2/2}}{\sqrt{2\pi}} dx = \int_{-\infty}^{\infty} \frac{e^{-(1-2\lambda)x^2/2}}{\sqrt{2\pi}} dx \quad \left\{ y = \sqrt{1-2\lambda} x \right.$$

($\lambda < \frac{1}{2}$)

$$= \int_{-\infty}^{\infty} \frac{e^{-y^2/2}}{\sqrt{2\pi}} \frac{dy}{\sqrt{1-2\lambda}} = \frac{1}{\sqrt{1-2\lambda}}$$

Johnson-Lindenstrauss dimension reduction

[JL 84]

Let $P \subseteq \mathbb{R}^d$ be a set of n points. $\forall \epsilon > 0$

$\exists$ a linear transformation $\varphi: \mathbb{R}^d \rightarrow \mathbb{R}^k$ s.t.

$$\forall u, v \in P \quad (1-\epsilon) \|u-v\|^2 \leq \|\varphi(u) - \varphi(v)\|^2 \leq (1+\epsilon) \|u-v\|^2$$

$$\text{and } k \leq \frac{12 \ln n}{\epsilon^2 - \epsilon^3}$$

Random φ :

$$\varphi \equiv \frac{1}{\sqrt{k}} \begin{bmatrix} G \end{bmatrix}$$

d

k

$$\forall i, j \quad G_{ij} \sim \mathcal{N}(0, 1) \quad \text{i.i.d.}$$

Studying just one pair (u, v)

Suffices to show that for any fixed $u, v \in \mathcal{P}$

$$\mathbb{P}\left[(1-\epsilon)\|u-v\|^2 \leq \|\varphi(u)-\varphi(v)\|^2 \leq (1+\epsilon)\|u-v\|^2\right] \geq 1 - \frac{2}{n^3}$$

$\varphi = \frac{1}{\sqrt{k}} \cdot G$

Say we can prove above. Then,

$$\mathbb{P}\left[\forall u, v \in \mathcal{P} \quad (1-\epsilon)\|u-v\|^2 \leq \|\varphi(u)-\varphi(v)\|^2 \leq (1+\epsilon)\|u-v\|^2\right].$$

$$\geq 1 - \binom{n}{2} \cdot \frac{2}{n^3}$$

$$\geq 1 - \frac{1}{n}$$

Studying a single unit vector.

$$\text{Let } W = \frac{u-v}{\|u-v\|}$$

$$\varphi = \frac{1}{\sqrt{2}} G$$

$$(1-\epsilon) \|u-v\|^2 \leq \|\varphi(u) - \varphi(v)\|^2 \leq (1+\epsilon) \|u-v\|^2$$

||

$$(1-\epsilon) \cdot \cancel{\|u-v\|^2} \leq \left\| \frac{1}{\sqrt{k}} G \frac{(u-v)}{\|u-v\|} \right\|^2 \leq (1+\epsilon) \cdot \cancel{\|u-v\|^2}$$

||

$$k(1-\epsilon) \leq \|Gw\|^2 \leq k(1+\epsilon)$$

Suffices to prove $\mathbb{P}_G \left[(1-\epsilon)k \leq \|Gw\|^2 \leq (1+\epsilon)k \right] \geq 1 - \frac{2}{\sqrt{m}}$

Understanding a single w

Fix $w \in \mathbb{R}^d$, $\|w\| = 1$

$$\begin{bmatrix} G \end{bmatrix} \begin{bmatrix} w \end{bmatrix} = \begin{bmatrix} x \end{bmatrix} \rightarrow x_i = G_{i1} w_1 + \dots + G_{id} w_d$$

$x_i \sim \mathcal{N}(0, \|w\|^2)$
 $\sim \mathcal{N}(0, 1)$

$$Z = \|Gw\|^2 = x_1^2 + \dots + x_k^2$$

Sums of squares of independent Gaussians

$$Z = X_1^2 + \dots + X_k^2 \quad \underbrace{X_1, \dots, X_k}_{i.i.d} \sim \mathcal{N}(0, 1)$$

$$\mathbb{E}[Z] = \mathbb{E}[X_1^2] + \dots + \mathbb{E}[X_k^2] = k$$

$$\mathbb{P}[Z \geq (1+\epsilon) \cdot k] = \mathbb{P}[e^{\lambda Z} \geq e^{(1+\epsilon)k}]$$

$$\leq \frac{\mathbb{E}[e^{\lambda Z}]}{e^{(1+\epsilon) \cdot k}} = \prod_{i=1}^k \frac{\mathbb{E}[e^{\lambda X_i^2}]}{e^{(1+\epsilon) \cdot k}}$$

$$\mathbb{E}[e^{\lambda X_i^2}] = \frac{1}{\sqrt{1-2\lambda}}$$

$$(\lambda < 1/2)$$

Choosing λ and k

$$P[Z \geq (1+\epsilon)k] \leq \frac{\left(\frac{1}{\sqrt{1-2\lambda}}\right)^k}{e^{\lambda(1+\epsilon)k}} = \left(\frac{e^{-2\lambda(1+\epsilon)}}{1-2\lambda}\right)^{k/2} \leq \frac{1}{n^3}$$

using $\lambda = \frac{\epsilon}{2(1+\epsilon)}$

$$\begin{aligned} P[Z \geq (1+\epsilon)k] &\leq \left(\underbrace{(1+\epsilon) \cdot e^{-\epsilon}}_{\leq (1+\epsilon)(1-\epsilon+\frac{\epsilon^2}{2})}\right)^{k/2} \\ &= 1 - \frac{\epsilon^2}{2} + \frac{\epsilon^3}{2} \leq e^{-(\epsilon^2-\epsilon^3)/2} \end{aligned}$$

$$P[\dots] \leq \frac{1}{n^3} \quad \text{for } k = \frac{12 \ln n}{\epsilon^2 - \epsilon^3}$$

Ex: Obtain similar bound for $P[Z \leq (1-\epsilon)k]$