

Statistical Data Analysis

Discussion notes – week 4

- Problem sheet 1
- Test of Gaussian mean
- Chain rule for pdfs

Problem sheet 1

Exercise 1: A beam of particles consists of a fraction 10^{-4} electrons and the rest photons. The particles pass through a double-layered detector which gives signals in either zero, one or both layers. The probabilities of these outcomes for electrons (e) and photons (γ) are

$$\begin{array}{ll} P(0|e) = 0.001 & \text{and} \quad P(0|\gamma) = 0.99899 \\ P(1|e) = 0.01 & P(1|\gamma) = 0.001 \\ P(2|e) = 0.989 & P(2|\gamma) = 10^{-5}. \end{array}$$

(a) [4 marks] What is the probability for the particle to be a photon given a detected signal in one layer only?

1) Use Bayes' theorem

$$P(A|B) = \frac{P(B|A) P(A)}{\sum_i P(B|A_i) P(A_i)}$$

$$a) \quad P(\gamma|1) = \frac{P(1|\gamma) P(\gamma)}{P(1|\gamma) P(\gamma) + P(1|e) P(e)}$$

$$= \frac{0.001 \times 0.9999 \quad \leftarrow 1 - 10^{-4}}{0.001 \times 0.9999 + 0.01 \times 10^{-4}} = 0.999 \text{ (000899)}$$

1(b)

(b) [4 marks] What is the probability for a particle to be an electron given a detected signal in both layers?

$$\begin{aligned} \text{b) } P(e|z) &= \frac{P(z|e) P(e)}{P(z|e) P(e) + P(z|\gamma) P(\gamma)} \\ &= \frac{0.989 \times 0.0001}{0.989 \times 0.0001 + 10^{-5} \times (1 - 0.0001)} \\ &= \underline{0.90818} \end{aligned}$$

Exercise 2: Consider the joint probability density for two continuous variables x and y given by

$$f(x, y) = \begin{cases} x + y & 0 \leq x \leq 1, 0 \leq y \leq 1, \\ 0 & \text{otherwise.} \end{cases}$$

(a) [6 marks] Find the marginal pdfs $f_x(x)$ and $f_y(y)$ and indicate what they look like with a simple sketch. Are x and y independent? Explain.

$$a) \quad f_x(x) = \int f(x, y) dy$$

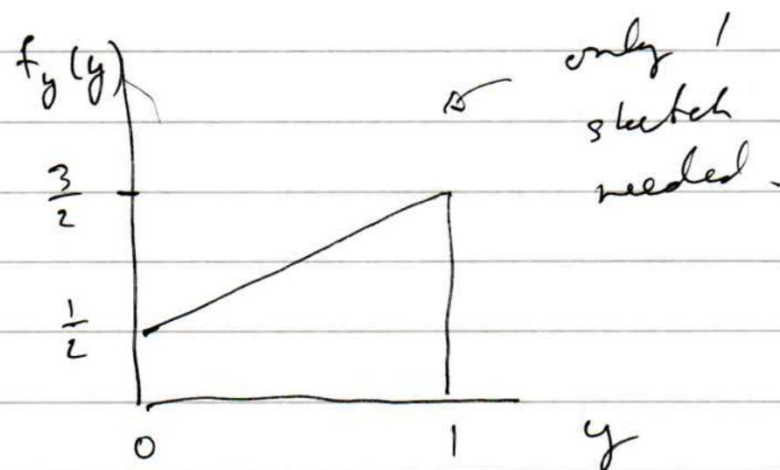
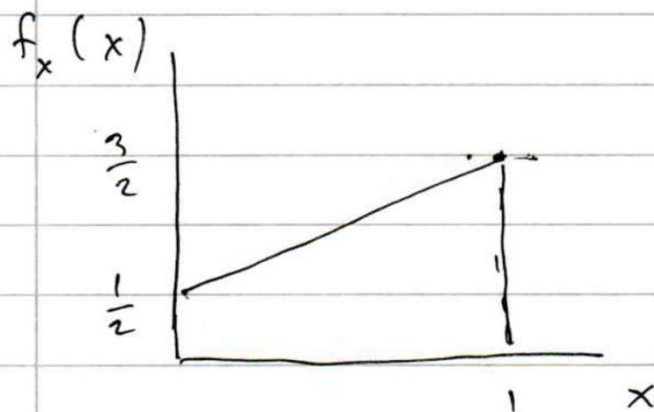
$$= \int_0^1 (x + y) dy$$

$$= \left(xy + \frac{y^2}{2} \right) \Big|_0^1 = \frac{1}{2} + x,$$

$$0 \leq x \leq 1$$

2(a) cont.

By symmetry, $f_y(y) = \frac{1}{2} + y$, $0 \leq y \leq 1$



$$f(x, y) = x + y$$

$$\neq f_x(x) f_y(y) = \left(\frac{1}{2} + x\right) \left(\frac{1}{2} + y\right)$$

$\Rightarrow x, y$ not independent.

2(b)

(b) [6 marks] Find the conditional probabilities $f(x|y)$ and $f(y|x)$. State how these two densities are related by Bayes theorem, and demonstrate that the relation holds using the conditional pdfs you have found together with the marginal pdfs from (a).

$$\begin{aligned} 2b) \quad f(x|y) &= \frac{f(x,y)}{f_y(y)} \\ &= \frac{x+y}{\frac{1}{2}+y}, \quad 0 \leq x \leq 1 \end{aligned}$$

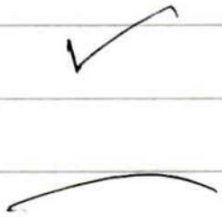
By symmetry

$$f(y|x) = \frac{x+y}{\frac{1}{2}+x}, \quad 0 \leq y \leq 1$$

2(b) cont.

Bayes' thm :

$$f(x|y) = \frac{f(y|x) f_x(x)}{f_y(y)}$$

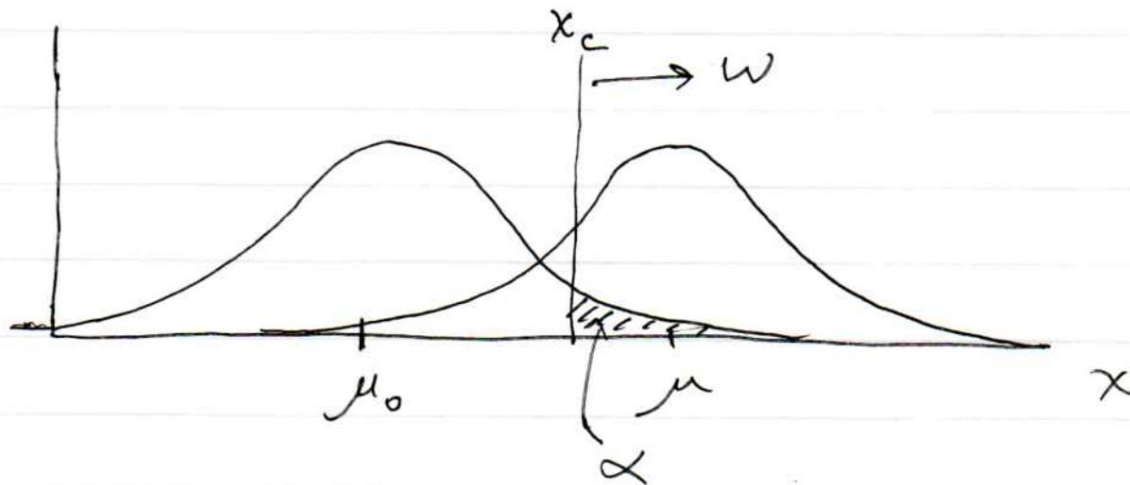
$$\frac{x+y}{\frac{1}{2}+y} = \frac{\frac{x+y}{\frac{1}{2}+x} \cdot (\frac{1}{2}+x)}{\frac{1}{2}+y}$$


From Week 4 extra slides

$x \sim \text{Gauss}(\mu, \sigma)$
 want to test μ
 σ known.

$$H_0: \mu = \mu_0 \quad \text{vs.} \quad H_1: \mu > \mu_0$$

$$\text{Take } W = \{x : x \geq x_c\}$$



$$\alpha = P(x \geq x_c | \mu_0)$$

$$= \int_{x_c}^{\infty} \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{(x-\mu_0)^2}{2\sigma^2}} dx$$

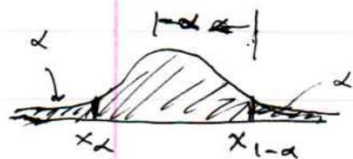
$$\text{let } y = \frac{x - \mu_0}{\sigma}$$

$$= \int_{\frac{x_c - \mu_0}{\sigma}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-y^2/2} dy$$

$$= 1 - \Phi\left(\frac{x_c - \mu_0}{\sigma}\right)$$

↖ Standard Gauss. cumul. dist.

$$\Rightarrow x_c = \mu_0 + \sigma \Phi^{-1}(1 - \alpha)$$



↖ Std. Gauss. quantile

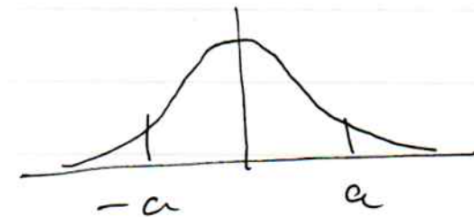
$$\Phi^{-1}(\alpha) = -\Phi^{-1}(1 - \alpha)$$

$$\text{Power } M = P(x \geq x_c | \mu)$$

$$= \int_{x_c}^{\infty} \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

$$= 1 - \Phi\left(\frac{x_c - \mu}{\sigma}\right)$$

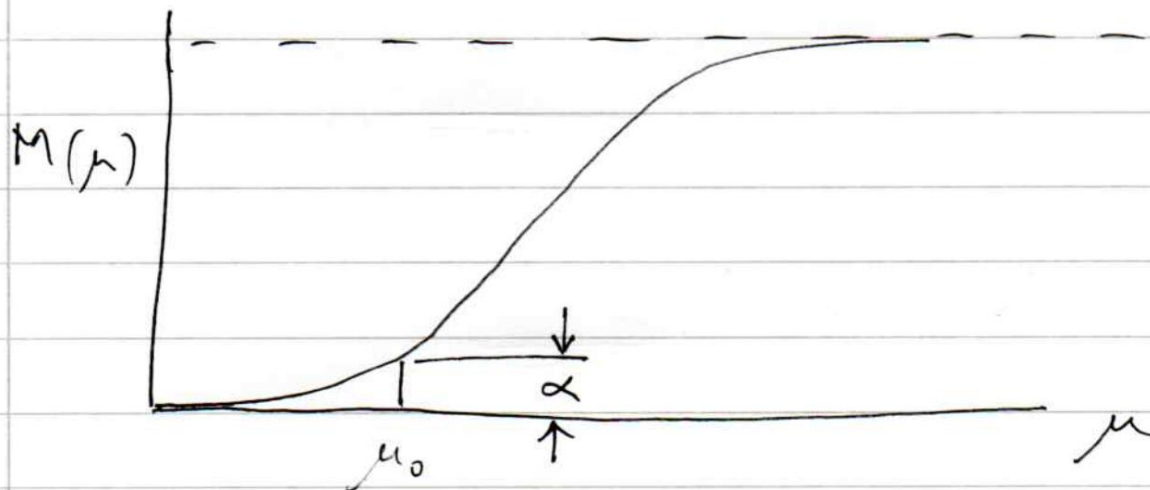
$$= \Phi\left(\frac{\mu - x_c}{\sigma}\right)$$



$$\Phi(-a) = 1 - \Phi(a)$$

$$= \Phi\left(\frac{\mu - \mu_0 - \sigma \Phi^{-1}(1-\alpha)}{\sigma}\right)$$

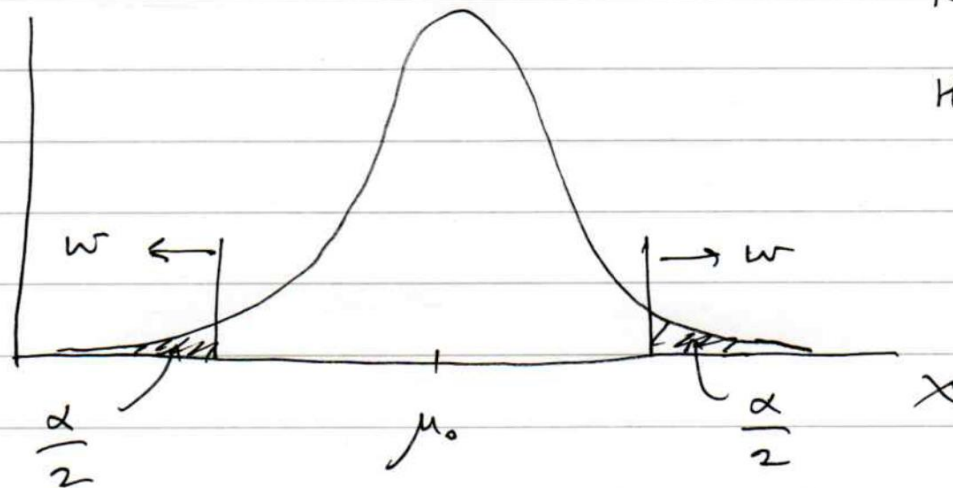
$$= \Phi\left(\frac{\mu - \mu_0}{\sigma} + \Phi^{-1}(\alpha)\right)$$



"Two-sided" test:

We might consider $\mu < \mu_0$ and $\mu > \mu_0$
as equally relevant alternatives.

⇒ choose two-sided critical region:

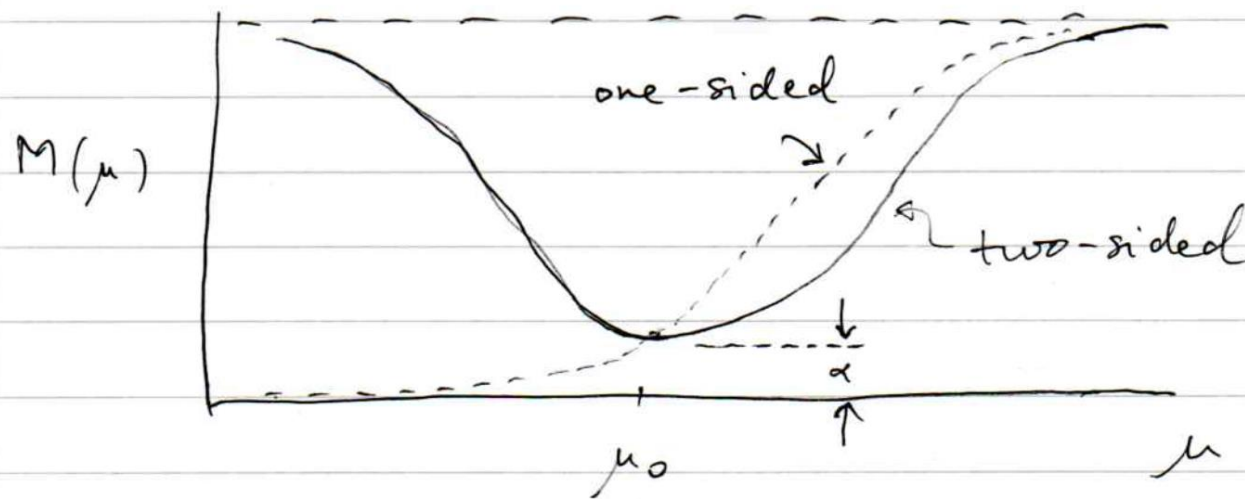


$$H_0: \mu = \mu_0$$

$$H_1: \mu \neq \mu_0$$

Two-sided w gives (some) power for both $\mu < \mu_0$ and $\mu > \mu_0$,

but less power for $\mu > \mu_0$ than the original one-sided test.



Chain rule for pdfs

Recall conditional pdf

$$f(x_2, x_1) = f(x_2 | x_1) f(x_1)$$

Generalize to

$$f(x_n, \dots, x_1) = f(x_n | x_{n-1}, \dots, x_1) \underbrace{f(x_{n-1}, \dots, x_1)}$$

↓

$$f(x_{n-1}, \dots, x_1) = f(x_{n-1} | x_{n-2}, \dots, x_1) \underbrace{f(x_{n-2}, \dots, x_1)}$$

↓

$$f(x_{n-2}, \dots, x_1) = f(x_{n-2} | x_{n-3}, \dots, x_1) f(x_{n-3}, \dots, x_1)$$

⋮
⋮
⋮

Assemble the factors:

$$f(x_n, \dots, x_1) = \prod_{i=1}^n f(x_i | \underbrace{x_{i-1}, \dots, x_1}_{\text{if } i > 1, \text{ else, no condition}})$$

e.g. $f(x_3, x_2, x_1) = f(x_3 | x_2, x_1)$

$$\times f(x_2 | x_1)$$
$$\times f(x_1)$$