

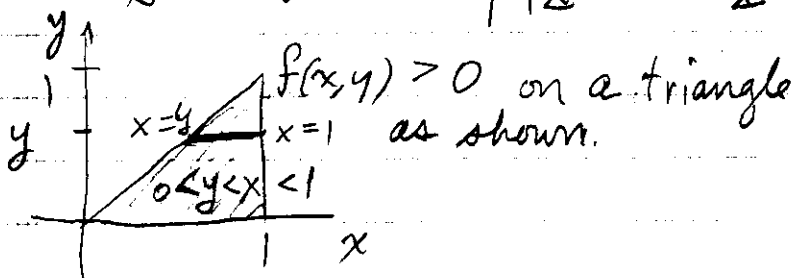
The method of conditional density.

Suppose X and Y are jointly distributed as follows. X is uniformly distributed on $(0,1)$, and given $X=x$, Y is uniformly distributed on $(0,x)$. Thus we are given

$$f_X(x) = 1, 0 < x < 1, \text{ and } f_{Y|X}(y|x) = \frac{1}{x}, 0 < y < x.$$

Thus the joint density is the product:

$$f(x,y) = f_{Y|X}(y|x) f_X(x) = \begin{cases} 1 \cdot \frac{1}{x} & 0 < y < x < 1 \\ 0 & \text{else.} \end{cases}$$



We find the marginal density of Y as:

$$f_Y(y) = \int_{-\infty}^{\infty} f(x,y) dx = \int_y^1 \frac{1}{x} dx = \ln x \Big|_y^1$$

$$= \ln 1 - \ln y = -\ln(y), 0 < y < 1.$$

Hence the conditional density of X given $Y=y$ is:

$$f_{X|Y}(x|y) = \frac{f(x,y)}{f_Y(y)} = \begin{cases} \frac{1}{-x \ln(y)}, & y < x < 1 \\ 0 & \text{all other } x \end{cases}$$

$$\text{Hence } E(X|Y=y) = \int_{-\infty}^{\infty} x f_{X|Y}(x|y) dx$$

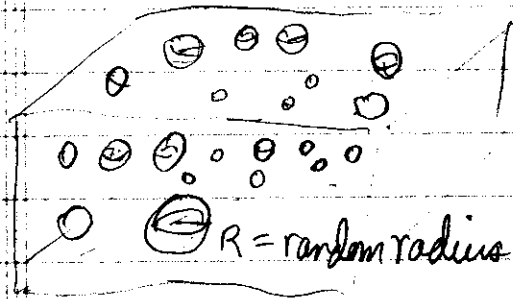
integrand is constant in x!

$$= \int_y^1 x \left(\frac{1}{-x \ln(y)} \right) dx = \int_y^1 \left(\frac{-1}{\ln(y)} \right) dx = \frac{1-y}{-\ln(y)}$$

at $y = .44$ this becomes $\frac{.56}{-\ln(.44)} = \boxed{.68211}$

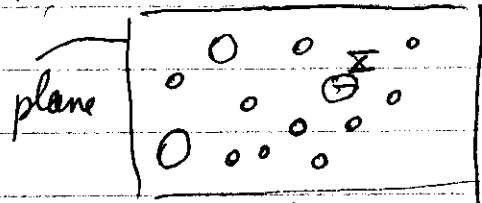
(p. 2)

3.5.2 Example B.



Σ = random radius of cross-sectioned circles when a random plane cuts the material.

R = random radius of spheres.



The holes in a slice of Swiss cheese.

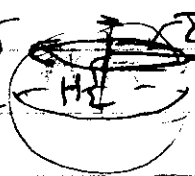
Consider a subset of the spheres defined by the condition:

$$R \in [r, r+dr],$$

that is R is infinitesimally close to r

for some fixed radius r , so all the spheres that remain in the material with this condition have the same radii. The plane cutting at random is the condition that the height H at which these spheres is cut is uniformly distributed on $[0, r]$.

Hence radius r { Σ given $R=r$ cross-section of radius Σ given $R=r$



Since $f_H(h) = \frac{1}{r} \quad 0 < h < r$

and since given $R=r$, $\Sigma^2 + H^2 = r^2$, we have the conditional cdf of Σ given $R=r$ is

$$\begin{aligned} F_{\Sigma|R}(x|r) &= P(\Sigma \leq x | R=r) \\ &= P(\sqrt{r^2 - H^2} \leq x) \\ &= P(H \geq \sqrt{r^2 - x^2}) = \int_{\sqrt{r^2 - x^2}}^r \frac{1}{r} dh \end{aligned}$$

constant integrand!

$= 1 - \frac{\sqrt{r^2 - x^2}}{r}$. Thus the conditional density of Σ given $R=r$

is $f_{\Sigma|R}(x|r) = \frac{d}{dx} (1 - \frac{\sqrt{r^2 - x^2}}{r}) = \frac{x}{r\sqrt{r^2 - x^2}} \quad 0 < x < r$

Thus $f(x,r) = \frac{x}{r\sqrt{r^2 - x^2}} f_R(r)$ and $f_{\Sigma}(x) = \int_x^{\infty} \frac{x}{r\sqrt{r^2 - x^2}} f_R(r) dr$, See p. 91.

(p. 3)

Bayesian Inference. If θ is a parameter for a density $f(x|\theta)$, then we may regard θ as an unknown and so give θ an initial density $g(\theta)$. Then the joint density

of X and θ is $f(x, \theta) = f(x|\theta)g(\theta)$

so the marginal density of X is

$f_X(x) = \int f(x|\theta)g(\theta)d\theta$ and the so-called posterior density of θ given $X=x$ is:

$$f(\theta|x) = \frac{f(x|\theta)g(\theta)}{f_X(x)}$$

Exercise 3.34 Suppose $g(\theta) = \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} \theta^{a-1} (1-\theta)^{b-1}$, $0 < \theta < 1$

as in 2.2.4, p. 58 (the beta density with parameters a and b). Take $a=b=3$. Then

$$g(\theta) = \frac{\Gamma(6)}{\Gamma(3)\Gamma(3)} \theta^2 (1-\theta)^2 = \frac{5!}{2!2!} \theta^2 (1-\theta)^2 = 30 \theta^2 (1-\theta)^2, \quad 0 < \theta < 1$$

Let now X have (discrete) freq. function $p(x|\theta) = \binom{n}{x} \theta^x (1-\theta)^{n-x}$, $x=0, 1, 2, \dots, n$

say for $n=10$. Then the

marginal freq. fn of X is: $p_X(x) = \int_0^1 \binom{10}{x} \theta^x (1-\theta)^{10-x} \cdot 30 \theta^2 (1-\theta)^2 d\theta$

$$= 30 \binom{10}{x} \cdot \int_0^1 \theta^{2+x} (1-\theta)^{12-x} d\theta = 30 \binom{10}{x} \frac{\Gamma(3+x)\Gamma(13-x)}{\Gamma(16)}$$

$x=0, 1, \dots, 10$. So the posterior density of θ given $X=x$ is:

$$\frac{30 \binom{10}{x} \theta^{2+x} (1-\theta)^{12-x}}{30 \binom{10}{x} \Gamma(3+x)\Gamma(13-x)} \Gamma(16), \quad 0 < \theta < 1, \text{ or beta with parameters: } \boxed{a'=3+x, b'=13-x}$$