

Exercise 2

i let $f(y_1, y_2) = k \exp \left(\frac{-2y_1^2 - y_2^2 + 2y_1 y_2}{-22y_1 - 14y_2 + 65} \right)$

(a) Show for appropriate k this is the density of a normal.

Soln. $(y - \mu)^T \underbrace{\begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}}_{\Sigma} (y - \mu)$

$$= 2y_1^2 + y_2^2 + 2y_1 y_2 + \mu^T \Sigma \mu - 2y^T \Sigma \mu$$

From y_1, y_2 terms:

$$4\mu_1 + 2\mu_2 = +22$$

$$2\mu_1 + 3\mu_2 = +14$$

$$\Rightarrow \mu_1 = 4, \mu_2 = 3$$

$$\mu^T \Sigma \mu = 49$$

$$k = \exp\left(-\frac{65^{-49}}{2}\right) (2\pi)^{-1} \frac{\det^k \Sigma}{1}$$

(b) Find μ, Σ .

Soln. See above.

2. Let U have density g and A be nonsingular. Define $Y = A(U + c)$.
Find the density of y .

Soln. By Radon-Nikodym, density f of Y is the unique fn (if it exists) satisfying for all Borel S that:

$$P\{Y \in S\} = \int_S f \, dy$$

Note the transform $t(U) = Y$
is invertible and bijective. So

$$\begin{aligned} P\{Y \in S\} &= P\{U \in t^{-1}(S)\} = \int_{t^{-1}(S)} g \, du \\ &= \int_{t^{-1}(S)} g \circ t^{-1} \circ t \, du = \int_S g \circ t^{-1} \, dt \end{aligned}$$

The pushforward is given by

$$dt_* u = \frac{det(\tilde{\epsilon}^{-1})^T}{\det A} dy$$

Viewing this as a differential n-form.

3. Show the matrix

(a)

$$A = \begin{pmatrix} 1 & \rho \\ \rho & 1 \end{pmatrix}$$

is PD if $\rho > -\frac{1}{2}$.

Soln. Prove by Schur complement.

$$A = \begin{pmatrix} X & Y \\ Z & W \end{pmatrix} \text{ with } W=1 \text{ PD},$$

$$\text{Then } X - YW^{-1}Z = \begin{pmatrix} 1-\rho \\ \rho \end{pmatrix} - \rho^2$$

Consider the vector (a, b) on this:

$$(a^2 + b^2)(1 - \rho^2) + \rho(1 - \rho)2ab \\ (1 - \rho)[(1 + \rho)(a^2 + b^2) + \rho^2 - ab]$$

Since $a^2 + b^2 \geq 2ab$ by
AM-GM,

$$\text{for } p > -y_2,$$

$$(1-p) \geq 0$$

$$(1+p)(a^2+b^2) + 2pb + s \geq 0$$

(b) Let $\Sigma = \begin{bmatrix} 1 & e \\ e & 1 \end{bmatrix}$. Find Σ^{-1} .

$$\text{Solve: } \det(\Sigma - \lambda I) = 0$$

$$= (1-\lambda)^2 - e^2$$

$$= (1-\lambda-e)(1-\lambda+e)$$

$$\lambda = 1-e, 1+e,$$

Solve for v_i , $(\Sigma - \lambda_i I)v_i = 0$,

$$\begin{pmatrix} -e & e \\ e & -e \end{pmatrix} v = 0 \rightarrow (1, 1)$$

$$\begin{pmatrix} e & e \\ e & e \end{pmatrix} v = 0 \rightarrow (1, -1)$$

$$\Sigma = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} 1+\rho & 0 \\ 0 & 1-\rho \end{pmatrix} \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

$$\text{So } \Sigma' = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} \sqrt{1+\rho} & 0 \\ 0 & \sqrt{1-\rho} \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

assuming $|\rho| \leq 1$ (for R-valued solutions)

Exercises 2b

1. Let $X \sim N(\mu, \Sigma)$ with $\mu = \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix}$
 and $\Sigma = \begin{pmatrix} \sigma_x^2 & \sigma_{xy}\rho \\ \sigma_{xy}\rho & \sigma_y^2 \end{pmatrix}$.

What is its MGF?

Soln. $\mathbb{E} \exp(t \cdot X) =$

$$\exp\left(\frac{1}{2} t \cdot \mu + t^T \Sigma'/_2\right)$$

$$= \exp\left(\frac{1}{2}(t_1\mu_1 + t_2\mu_2 + t_1^2\sigma_x^2 + 2t_1t_2\sigma_{xy}\rho + t_2^2\sigma_y^2)\right)$$

2. If $\mathbf{Y} \sim N(\mu, \Sigma)$ show $Y_i \sim N(\mu_i, \sigma_{ii})$.
 Soln. By choice of $e_i^T \mathbf{Y}$ as the
 Normal. Exponential MGF.

3. Let $\mu = \begin{pmatrix} 2 \\ 1 \\ 2 \end{pmatrix}$ and $\Sigma = \begin{pmatrix} 2 & 1 & 1 \\ 1 & 1 & 0 \\ 1 & 0 & 3 \end{pmatrix}$.

If $\mathbf{Y} \sim N(\mu, \Sigma)$ then find
 the joint of $Z_1 = Y_1 + Y_2 + Y_3$
 and $Z_2 = Y_1 - Y_2$.

Soln. Notice $Z = \underbrace{\begin{pmatrix} 1 & 1 & 1 \\ 1 & -1 & 0 \end{pmatrix}}_C \mathbf{Y}$

so $Z \sim N\left(C\mu, C\Sigma C^T\right)$.

4. Show $a^T \mathbf{Y} \perp \!\!\! \perp b^T \mathbf{Y}$ for $a \perp b$
 and $\mathbf{Y} \sim N(\mu, I)$.

Soln. Factorize the MGF:

$$M_{(a+b)^T \mathbf{Y}}(t) = \exp\left(\frac{1}{2}(a+b)^T \mu + \frac{1}{2}(a+b)^T I(a+b)\right)$$

Since cross terms cancel,

$$M_{(a+b)^T Y}(t) = M_{a^T Y}(t) M_{b^T Y}(t)$$

5. Observe (X_i, Y_i) bivariate r.v.s,
n indep observations. Find the global
of (\bar{X}, \bar{Y}) .

Soln. If $(X_i, Y_i) \sim N(\mu, \Sigma)$

then its avg $\rightarrow N(\mu, \Sigma/n)$

6. If $Y_1 + Y_2, Y_1 - Y_2$ are
indep std r.v.s, what is $Y = (Y_1, Y_2)$

Soln. Let $X_1 = Y_1 + Y_2, X_2 = Y_1 - Y_2$.

$$\begin{aligned} M_{X_1, -X_2}(t) &= M_{X_1}(t) M_{-X_2}(t) \\ &= \exp(t^2) \end{aligned}$$

$$\begin{aligned} \text{Some } \lim X_1 + X_2. \mathbb{E}[(X_1 + X_2)(X_1 + X_2)] \\ = D \end{aligned}$$

$$\text{B.s.t } t_1(x_1 - x_2) + t_2(x_1 + x_2)$$

$$= \underbrace{(t_1 + t_2)x_1 - (t_2 - t_1)x_2}_{\text{Jointly Gaussian.}}$$

So $y \sim N(0, \Sigma)$ too.

1. Let $f(x, y) = \frac{1}{2\pi} \exp\left(-\frac{1}{2}(x^2 + y^2)\right)$

$$x \left[1 - \frac{xy}{(1+x^2)(1+y^2)} \right].$$

What are the marginals?

Soln. By symm. suff. to show for X .

$$\int f dy = \frac{1}{\sqrt{2\pi}} e^{-x^2/2} \int \frac{1}{\sqrt{2\pi}} e^{-y^2/2} \left(1 - \frac{xy}{(1+x^2)(1+y^2)}\right) dy$$

$$= \gamma(x) \left[\int \gamma(y) dy - \int y \frac{\gamma(y)x}{(1+x^2)(1+y^2)} dy \right]$$

γ gaussian density.

But left term is 1
and right term is odd.

8. Let $y_i \sim N(0, 1)$. Show

$$\bar{Y} \perp\!\!\!\perp \sum_i (y_i - \bar{Y})^2.$$

Sln. Define $\bar{X} = (\bar{Y}, Y_1 - \bar{Y}, \dots, Y_n - \bar{Y})$.

$$\begin{aligned}
 M_{\bar{X}}(t) &= \mathbb{E} \exp(t_0 \bar{Y} + (Y - \bar{Y})^T t) \\
 &= \mathbb{E} \exp(t^T Y - \frac{1_n^T Y}{n} \left(\underbrace{1_n^T t - t_0}_n \right)) \\
 &= \mathbb{E} \exp \left(\left(t - \underbrace{\left(\frac{1_n^T t - t_0}{n} \right) 1_n}_n \right)^T Y \right) \\
 &= \exp \left(\frac{1}{2} \| \alpha \|^2 \right) \\
 &= \exp \left(\frac{1}{2} \left(\| t \|_2^2 - \frac{2(1_n^T t)^2}{n} + t_0 \frac{1_n^T t}{n} + \frac{(1_n^T t - t_0)^2}{n} \right) \right)
 \end{aligned}$$

$$= \exp \left(\frac{1}{2} \left(\|t\|_2^2 - \frac{(1_n^\top t)^2}{n} - t_0^\top \frac{1_n^\top t}{n} + \frac{t_0^\top t}{n} \right) \right)$$

Scratch above. Consider

$$C = \begin{pmatrix} 1_n 1_n^\top \\ I_n - \frac{1_n 1_n^\top}{n} \end{pmatrix}$$

Then $\bar{X} = C Y$

$$\begin{aligned} \text{and } MGF_{\bar{X}}(t) &= \exp \left(\frac{1}{2} t^\top C C^\top t \right) \\ &= \exp \left(\frac{1}{2} t^\top \begin{pmatrix} 1_n & 0 \\ 0 & I_n - \frac{1_n 1_n^\top}{n} \end{pmatrix} t \right) \\ &= MGF_{\bar{Y}}(t_0) MGF_{Y_{-i}}(t_{1:n}) \end{aligned}$$

\rightarrow ii.

9. Let $X \sim N(0, I_3)$ and
obtain

$$Y = \begin{pmatrix} Y_{N_3} & Y_{N_3} & Y_{N_3} \\ Y_{N_2} & -Y_{N_2} & 0 \\ Y_{N_6} & Y_{N_6} & -Y_{N_6} \end{pmatrix} X.$$

What is the distribution of Y ?

Polar Since X bivariate is ortho,

$$Y \sim N(0, I_n).$$

Exercise 2e

1. Are pairwise indep. Mults w/
mutually independent? MVN joint

Solu. Yes, $b(c) \text{ cov} = 0$ off-diag.

2. Let $Y \sim N(\mu_{\mathbb{I}_n}, \Sigma)$

where $\Sigma = (1-p)I_n + pJ_n$
and $J_n = \mathbb{1}_n \mathbb{1}_n^T$. Let $p > \frac{1}{n-1}$.

When $P = O$, \bar{Y} and $Y - \bar{Y}$
are indep. What about $P \neq O$?

$$\text{fdn. } \begin{pmatrix} \bar{Y} \\ Y - \bar{Y} \end{pmatrix} = \begin{pmatrix} \mathbf{1}_n^T / n \\ \mathbf{I} - \frac{\mathbf{1}_n \mathbf{1}_n^T}{n} \end{pmatrix} Y$$

So by thm 2.2,

indep if

$$O = \frac{\mathbf{1}_n^T}{n} \left[(1-p) \mathbf{I} + p \mathbf{J} \right] \left(\mathbf{I} - \frac{\mathbf{J}}{n} \right)$$

$$= \underbrace{\frac{p \mathbf{1}_n^T}{n} \mathbf{J}}_{p \mathbf{1}_n^T} \left(\mathbf{I} - \frac{\mathbf{J}}{n} \right)$$

$\rightarrow N \otimes O \Rightarrow$ Not indep.

3. Suppose $Y \sim N(\mu, \Sigma)$ with

$$\Sigma = \begin{pmatrix} 1 & \rho & \rho \\ \rho & 1 & \rho \\ \rho & \rho & 1 \end{pmatrix}. \text{ For}$$

what ρ values are

$$Z_1 = Y_1 + Y_2 + Y_3$$

$$Z_2 = Y_1 - Y_2 - Y_3$$

- indep?

Soln. $Z = \underbrace{\begin{pmatrix} 1 & 1 & 1 \\ 1 & -1 & -1 \end{pmatrix}}_A Y$

$Z_1 \perp\!\!\!\perp Z_2$ when $\underbrace{A \Sigma A^T}_{= 0 \text{ off diag.}} = 0$

$$\begin{pmatrix} 1+\rho & 2\rho+1 & 1+\rho \\ 1-\rho & 1 & -\rho-1 \end{pmatrix} A^T$$

$$= \begin{pmatrix} 3+\rho & -1-2\rho \\ -1-2\rho & 1 \end{pmatrix} \rightarrow \rho = -1/2$$

Exercises 2d

1. Show that the MGF of

$$\sum_{i=1}^n d_i Z_i^2 \sim \chi^2_{\sum d_i}$$

$\sim \chi^2_{\sum d_i}$

$$\text{is } \prod_{i=1}^n (1 - 2d_i)$$

Soln. Follows by Z_1^2 MGF $(1-2t)^{\frac{n}{2}}$
 w/independent.

2. Let $Y \sim N(0, I_n)$ and A be symm.

(a). What is the MGF of Y ?

$$\begin{aligned} \text{Soln. MGF}_Y(t) &= \int p(y) \exp(y^T A y + t) dy \\ &= \int_{\mathbb{R}^n} \exp\left(\frac{1}{2} y^T (I - 2At)y\right) dy \end{aligned}$$

For small t , $I \geq 2At$ so this is a
 Normal tilt; and the argument is the
 partition, $\det(I - 2At)^{-\frac{1}{2}}$.

(b) If A is independent of rank r ,
how does the answer change?

Soln. $MGF_Y(t) = (1-2t)^{-r/2}$ by
Corr. 2.8 & MGF of χ_r^2 .

((c)) What happens if $Y \sim N(0, \Sigma)$?

Soln. By eigen $\Sigma = Q \Lambda Q^T$
w/ Q full rank and unitary - U.S.W.

Let $\Sigma^{1/2} = Q \Lambda^{1/2} Q^T$.

If Λ is not full rank, det $\Lambda^{-1/2}$
w/ $0^{-1/2} = 0$, then let $Z = \Sigma^{-1/2} Y$
so $Z \sim N(0, (I_r \ 0))$

When $r = \text{rank } \Sigma$. As

$$Y^T A Y = Z^T (\Sigma^{1/2})^T A \Sigma^{1/2} Z$$

We can handle numbers

$\Sigma^{1/2} A$, and wlog take $v=n$.

so then $\Sigma^{1/2}$ is full rank. By
2a,

$$\text{MGF}_{Y \sim \mathcal{A}(\Sigma)}(t) = \text{MGF}_{\Sigma^{1/2}}(t) \quad (f)$$

$$= \det(I - 2(\Sigma^{1/2})^\top A \Sigma^{1/2} t)$$
$$= \det(\Sigma^{-1/2} \Sigma^{1/2} - 2 \sum_i A \Sigma^{1/2} t)$$

$$= \underbrace{\det(\Sigma^{-1/2}) \det(\dots) \det(\Sigma^{1/2})}_{x=1}.$$

$$= \det(I - 2A\Sigma t)$$

3. Find a, b st

$$a(Y_1 - Y_2)^2 + b(Y_1 + Y_2)^2 \sim \chi^2_2.$$

Sln. = $\frac{(a+b)(Y_1^2 + Y_2^2)}{+ (a+b)(Y_1 Y_2)} \rightarrow \chi^2_2$

$$\rightarrow a = b = Y_2 \text{ Only}$$

Sln. by inspecting MLE.

4. Let Y_i be iid $N(0, 1)$. i.e. $i \in [n]$

$$\text{Defn } Y_{n+1} \equiv Y_1.$$

$$\text{Then } \sum_{i=1}^n (Y_{i+1} - Y_i)^2$$

$$\chi^2 - \text{distributed?}$$

Sln. Sln. $\rightarrow Y^T A^T A Y$ for

$$A = \text{band} (0 \rightarrow 1, 1 \rightarrow -1)$$

(i.e. the matrix of terms

$$a_{ij} = \begin{cases} 1 & \text{if } j-i \equiv 0 \pmod{n} \\ -1 & \text{if } j-i \equiv 1 \pmod{n} \end{cases}$$

$$\text{Then } A^T A = \text{band} (-1 \rightarrow -1, 0 \rightarrow 2, 1 \rightarrow -1)$$

$\gamma_3(A^T A)$ is rank-2 idempotent

but by inspection of the $(n, 1)$ -st entry, $(A^T A)^2 \neq (A^T A)\alpha$
for any α . for any $n > 3$.

S. Let A, B be Symmetric.

Find joint mgf of $Y^T A Y, Y^T B Y$.

Use this to characterize when the two forms are independent.

Sdn. Use the same procedure
as in 2(A), yellow

$$\text{MGF}_{(Y^T A Y, Y^T B Y)}^{(s, t)} = \det(I - 2sA - 2tB)^{-1/2}.$$

If $AB = 0$,

$$\begin{aligned} \text{Notice } & (I - 2sA)(I - 2tB) \\ &= I - 2sA - 2tB + \underbrace{4stAB}_0 \\ &= I - 2sA - 2tB. \end{aligned}$$

But then MGF of pair = product of MGFs
(det distributes over product).
 \rightarrow Indep.

Miscellaneous Exercises 2.

1. Let $\varepsilon \sim N(0, \sigma^2)$ and $Y_{i+1} = \rho Y_i + \varepsilon_i$,
with $Y_0 \sim N(0, \sigma_0^2)$.

(a) Find $\text{var } Y_n$.

Soln. Note Y_0, ε_i are mutually indep.

$$\text{var } Y_i = \rho^2 \text{var } Y_{i-1} + \sigma^2$$

$$\text{cov}(Y_i, Y_{i+1}) = \rho \text{var } Y_i$$

$$Y_{i+2} = \rho^2 Y_i + \rho \varepsilon_{i+1} + \varepsilon_{i+2}$$

$$\text{cov}(Y_i, Y_{i+2}) = \rho^2 \text{var } Y_i$$

$$\text{var } Y = \begin{pmatrix} \overbrace{\rho^2 \sigma_0^2 + \sigma^2}^n & & & \\ \vdots & \overbrace{\rho \sigma_0^2 + (\rho^2 + 1) \sigma^2}^n & & \\ \rho^2 \dots & \dots & \ddots & \rho^6 \sigma_0^2 + (\rho^4 + \rho^2 + 1) \sigma^2 \\ & & & \sigma^2 \end{pmatrix} \quad \text{nsr rorv}$$

(b) What's the协方差 of Y ?

Soln. $\hat{Y} \rightarrow \text{MIN S/C}$

y_i are jointly Normal.

$$a^T y = b^T \begin{pmatrix} \varepsilon \\ y_0 \end{pmatrix} \text{ Then } b^T$$

and any a^T .

2. Let $y \sim N(0, I_n)$. Define:

$$x = Ay, u = Bu, v = Cv$$

for unknowns A, B, C . Suppose

$$\text{cov}(x, u) = \text{cov}(x, v) = 0.$$

Then show $X \perp\!\!\!\perp U+V$.

Soln. Since every thing is normal,

$$X \perp\!\!\!\perp U, X \perp\!\!\!\perp V. \text{ Thus}$$

$$X \perp\!\!\!\perp U+V.$$

3. $i \in \{n\}$, $Y_i \stackrel{iid}{\sim} \mathcal{N}(\mu, \sigma^2)$.

Show $\bar{Y} \perp \text{I.I.} \sum_i (Y_i - \bar{Y})^2$.

$$\text{Soh. } \bar{Y} = \frac{1}{n} \sum_{i=1}^n Y_i$$

$$Q = \bar{Y}^T A^T A \bar{Y},$$

$$A = \text{circulant}(0 \rightarrow 1, 1 \rightarrow -1)$$

$$\therefore \text{Meet } A^T = \text{circulant}(-1 \rightarrow -1, 0 \rightarrow 1)$$

Let F be the DFT. Then

If C_A = first column of A :

$$A = F^{-1} \text{diag}(F C_A) F$$

$$\therefore A^T A = F^{-1} \left(\text{diag}(F C_A) \times \text{diag}(F C_{A^T}) \right) F$$

$= \text{circulant}(C_A * C_{A^T}) \rightarrow \text{convolution.}$

$$\therefore A^T A = \text{correlated} (-1)$$

$$CA = \begin{array}{c} -1 & 0 & 1 & \dots & n \\ \hline 0 & 1 & -1 & 0 \end{array}$$

$$\text{flip} = \begin{array}{c} - & 0 & -1 & 1 & 0 \\ \hline \end{array}$$

$$C_A^T = \begin{array}{c} 0 & -1 & 1 & 0 \\ \hline \end{array}$$

idx -1: idx 0: idx 1

$$\begin{array}{r} -2 \\ -1 \\ \hline -1 \end{array} \quad \begin{array}{r} -1 \\ 1 \\ \hline \text{flip} \end{array}$$

$$\begin{array}{r} -1 \\ \cancel{0} \\ \hline -1 \end{array} \quad C_A^T$$

$$\begin{array}{r} 2 \\ \hline -1 \end{array}$$

$$A^T A = \text{correlated} (-1 \rightarrow -1 \\ 0 \rightarrow 2 \\ 1 \rightarrow 1)$$

Above is neat but correlated
to the problem 1st

By Thm 2.5, $A\bar{Y} \perp\!\!\!\perp \bar{Y}$

If cov is 0 , $\Rightarrow \|A\bar{Y}\|^2 \perp\!\!\!\perp \bar{Y}$

$$\text{cov}\left(\frac{1_n}{n} Y, A\bar{Y}\right)$$

$$= \frac{1^T}{n} \underbrace{\text{cov}(Y, Y)}_{\sigma^2 I} A^T$$

$$= 0 \quad b/c \quad A 1_n = 0$$

4. If X, Y are ^{indep.} MVN

then show $aX + bY$ is

Soln. $\begin{pmatrix} X \\ Y \end{pmatrix}$ is MVN, so linear $\begin{pmatrix} aI & 0 \\ 0 & bI \end{pmatrix}$ transform is.

5. If $Y \sim N(0, I_n)$, $a \neq 0$

Show $Y^T Y | a^T Y = 0 \sim \chi_{n-1}^2$.

Soln. Let $B \in \mathbb{R}^{n-1 \times n}$

be the orthonormal basis completion
after a , w log $\|a\|_2 = 1$. Then

$Q = [B \ a]$ is unitary w/

$$Q^T Y = \begin{bmatrix} B^T \\ a^T \end{bmatrix} Y.$$

$$\begin{aligned} Y^T Y &= Y^T Q Q^T Y \\ &= \|Q^T Y\|_2^2 = \|B^T Y + a^T Y\|_2^2. \end{aligned}$$

Condition on $a^T Y = 0$, thus is
 $\|B^T Y\|_2^2$. By constr. $B^T Y \sim N(0, I_{n-1})$

So χ^2_{n-1} holds under each hypothesis.

6. Let $Y \sim N(\mu I_n, \Sigma)$ where

$$\Sigma = (1-p)I_n + pJ_n \text{ w/ } p > -\frac{1}{n-1}.$$

$$\text{Show } S^2 = \sum_i (Y_i - \bar{Y})^2 / (-p) \sim \chi^2_{n-1}$$

Soln.

$$S^2 f(-p) = \left\| \underbrace{\left(I_n - \frac{1}{n} J_n \right) Y}_2 \right\|^2$$

$\sim N(0, \Sigma')$ where

$$\begin{aligned} \Sigma' &= (1-p)I_n + \left(p + \frac{1-p}{n} \right) J_n \\ &= ((1-p)\left[I_n - \frac{1}{n} J_n \right]) \end{aligned}$$

Above is just $(I_n - \frac{1}{n} J_n)(\Sigma)$,
 Missing right mult by $(I_n - \frac{1}{n} J_n)$, but idempotent

$$S_0 \quad S^2 = \|N(0, I_n - \frac{1}{n} J_n)\|_2^2$$

by cancellation of $(-P)$ term.

Cov is rank $n-1$ & idempotent
is sym, so $S^2 \sim \chi_{n-1}^2$.

7. Let $Y_i \stackrel{iid}{\sim} N(\mu, \Sigma)$.

Show $\frac{1}{n-1} \sum_{k=1}^n (Y^{(k)} - \bar{Y})(Y^{(k)} - \bar{Y})^T$
is an unbiased est. of Σ .

Soln. Inspect ij -th term:

$$\Sigma_{ij} = \frac{1}{n-1} \sum_k (Y_{;i}^{(k)} - \bar{Y}_i)(Y_{;j}^{(k)} - \bar{Y}_j)$$

Each term is identically distributed:

$$\Sigma_{ij} = \frac{n}{n-1} \mathbb{E}(Y_{;i}^{(1)} - \bar{Y}_i)(Y_{;j}^{(1)} - \bar{Y}_j)$$

$$= \frac{1}{n-1} \mathbb{E} \left[\left(\frac{n-1}{n} Y_i^{(1)} - \frac{1}{n} \sum_{k \neq i} Y_k^{(1)} \right) \times \right.$$

$$\left. (\dots j \dots) \right]$$

$$= \frac{1}{n(n-1)} \mathbb{E} \left[\left((n-1) Y_i^{(1)} - \sum_{k \neq i} Y_k^{(1)} \right) \right.$$

$$\left. \left((n-1) Y_i^{(1)} - \sum_{k \neq i} Y_k^{(1)} \right) \right]$$

$$(n-1)^2 \mathbb{E} \left[Y_i^{(1)} Y_j^{(1)} \right]$$

$$- (n-1) \mathbb{E} Y_i^{(1)} \mathbb{E} \sum_j Y_j^{(k)}$$

$$- (n-1) \mathbb{E} Y_i^{(1)} \mathbb{E} \sum_{j \neq i} Y_j^{(1)}$$

$$+ \sum_{k,m \geq 1} \mathbb{E} Y_i^{(k)} Y_j^{(m)} \quad k \neq 1$$

$$k, m \geq 1$$

$$\begin{aligned}
&= \frac{1}{n(n-1)} \left[\mathbb{E} Y_i^{(1)} Y_j^{(1)} ((n-1)^2 \right. \\
&\quad \left. - 2(n-1) \mu_i \mu_j \right. \\
&\quad + (n-1) \mathbb{E} Y_i^{(1)} Y_j^{(1)} \\
&\quad \left. + n(n-1) \mu_i \mu_j \right] \\
&\quad - (n-1)^2 \mu_i \mu_j \\
&\quad + (n-1) \mu_i \mu_j \\
&= \mathbb{E} Y_i^{(1)} Y_j^{(1)} - \mu_i \mu_j \\
&= \text{cov}(Y_i^{(1)}, Y_j^{(1)}) \\
&= \sum_{ij}
\end{aligned}$$

8. Let $Y \sim N(0, I_n)$

and $AB = BA = 0$ for

symm. independent submatrices A, B .

Show quadratic forms over

$A, B, I - A - B$ are indep χ^2 .

Ex. For A, B , this follows

by Thm 2.7 + Ex 2.12.

$A + B$ is still sym, indep.

So by Rx 2.11 $I - A - B$

is sym, indep, and

indep χ^2 as well from $A + B$.

We check only w/ constraints A, B :

$$A(I - A - B) = A \underbrace{- A^2}_{\underbrace{A}_0} - \underbrace{AB}_0.$$

Same reasoning for B.

Q. Let $(X_i, Y_i) \stackrel{\text{iid}}{\sim} N \begin{pmatrix} \mu_x \\ \mu_y \end{pmatrix}, \begin{pmatrix} \sigma_x^2 & \rho_{xy} \sigma_y \\ \rho_{xy} \sigma_x & \sigma_y^2 \end{pmatrix}$.

Define $W = \begin{pmatrix} X \\ Y \end{pmatrix}$

(a) Find distab. of W.

Soln. All bivariate are jointly Gaussian

by II so

$$W \sim N \begin{pmatrix} \mu_1 I_n \\ \mu_2 I_n \end{pmatrix} \mid \begin{pmatrix} \sigma_x^2 I_n & \rho \sigma_x \sigma_y I_n \\ \rho \sigma_x \sigma_y I_n & \sigma_y^2 I_n \end{pmatrix}$$

(b) Find $Y|X$.

Given. By given MUN condition,

$$Y|X \sim$$

$$N\left(\mu_y I_n + \rho \sigma_x \sigma_y I_n \begin{pmatrix} \sigma_x^2 I_n \\ X - \mu_x I_n \end{pmatrix}, \right.$$

$$\left. \sigma_y^2 I_n - \rho^2 \sigma_x^2 \sigma_y^2 \begin{pmatrix} \sigma_x^2 \end{pmatrix}^{-1} \right)$$

$$= N\left(\mu_y I_n + \frac{\rho \sigma_y}{\sigma_x} \begin{pmatrix} X - \mu_x I_n \end{pmatrix}, \right.$$

$$\left. \sigma_y^2 (1 - \rho^2) I_n \right)$$

i.e. Suppose $Y \sim N_2(0, \Sigma)$.

Show $Y^\top \Sigma^{-1} Y - Y_1^2 / \Sigma_{11} \sim \chi^2_1$

Soln. Define $\vec{z} = \sum^{-1/2} \vec{y}$.

So $\vec{z} \sim N(0, I_n)$ & orthogonal

ჩამდება $\vec{z}^T \vec{z} - z_1^2 = z_2^2 \sim \chi^2$

II. Let y_i be MA(1)

W. parameter ϕ & noise $N(0, \sigma^2)$.

How is y distributed?

Soln. $y_i = \varepsilon_i + \phi y_{i-1}$

$$\therefore y_i = \sum_{j=0}^i \phi^{i-j} \varepsilon_j$$

Since y_i is finite sum of
n Gaussians $\{\varepsilon_j\}_j$, all y_i are JG.
 $\rightarrow MVN$.

$$Y = \begin{pmatrix} \phi & & \\ & \ddots & \\ & & \ddots & \\ & & & \phi_1 \end{pmatrix} \epsilon$$

A

$$\sim N(0, \sigma^2 A A^\top)$$

13. Let $Y_i \stackrel{iid}{\sim} N(0, 1)$ for $i \in \{3\}$
 and find the MGF of

$$2(Y_1 Y_2 - Y_2 Y_3 - Y_3 Y_1)$$

and show via egene. χ^2 approach.

Soln. The answer is

$$(Y_1 + Y_2 - Y_3)^2 - Y_1^2 - Y_2^2 - Y_3^2$$

$$Y^T \left(-I_3 + \underbrace{\begin{pmatrix} 1 & 1 & -1 \end{pmatrix}^T}_{\begin{pmatrix} 1 & 1 & -1 \end{pmatrix}} \times \right)$$

A

$\times Y$

So by (2.10) it's
the eigenvalues of the
above matrix, $2, -1, -1$,

$$\text{so } Y^T A Y \sim \{ \lambda_i \}^2 \lambda_i$$

13. Use Normal distribution
through joint Gaussianity
to show facts about its distribution

(a). $Z_i \stackrel{iid}{\sim} N(0, 1)$ then $Z \sim N(0, I)$.

Soln. $a^T Z = \sum a_i Z_i \sim N(0, \|a\|_2^2)$
as desired.

(b) Show its MGF.

Sln. As $t^T X \sim N(t^T \mu, t^T \Sigma t)$

by independence, and usual
MGF is the usual

$$\exp\left(s\mu + \frac{1}{2}s^T \Sigma s\right),$$

apply to $t^T X_1, t$ fixed, where $s =$.

(c). Find its density.

Sln. $Y \sim N(\mu, \Sigma)$.

$$\Sigma^{-1/2} (Y - \mu)$$
 is standard,

so just a product of Normals
 \rightarrow Use rule of Jacobian.

(4). [last term is odd in y_i :]

(5. Let $Y \sim N_4(0, I)$

and $Q = Y_1 Y_2 - Y_3 Y_4$

(a) Why is Q not \mathbb{R}^2 ?

Soln. Q is symmetric about 0 .

(b) Find the MGF of Q .

Soln. Note Q is $Y^T A Y$ and

$$A = \underbrace{\left(-I_4 + \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}^T \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \right)}_{\frac{1}{2}}$$

eigenvalues are $0, 0, 1, -1$

then use

$$\text{MGF} = \det(I - 2tA)^{-1/2}$$

Ex. 2 & #4

16. This was done in a previous exercise, pt 1.)

17. $Y \sim N(\mu, \Sigma)$,

show

$$\begin{aligned} \text{var}(Y^T A Y) \\ = \text{Tr}((A \Sigma)^2) / 2 \\ + \cancel{Y^T A \Sigma A Y} \end{aligned}$$

(wlog A Symm).

Soln. Follows directly from Thm. 16.