

Matrix

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \quad \begin{bmatrix} A & B \\ C & D \end{bmatrix}^{-1} =$$

$$\begin{bmatrix} A^{-1} + A^{-1}B(D-CA^{-1}B)^{-1}CA^{-1} & -A^{-1}B(D-CA^{-1}B)^{-1} \\ -(D-CA^{-1}B)^{-1}CA^{-1} & (D-CA^{-1}B)^{-1} \end{bmatrix}$$

$\det A = |A| = \sum_{j=1}^N (-1)^{i+j} a_{ij} M_{ij} = \sum_{i=1}^N (-1)^{i+j} a_{ij} M_{ij}$
 where M is the minor: The *Minor* of entry a_{ij} , denoted by M_{ij} , is the determinant of the sub-matrix obtained by deleting i th row and j th col
Cramer's Rule: for equation $Ax = b$, Define A_i to be equal to the matrix but with col i replaced by b , then we have

$$x_i = \frac{\det A_i}{\det A}$$

Example:

$$\begin{bmatrix} 3 & 1 & 0 \\ -1 & 0 & 1 \\ 1 & 1 & 4 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} -3 \\ -2 \\ 0 \end{bmatrix}$$

thus $\det A = 2$ and

$$\det(A_1) = \begin{vmatrix} -3 & 0 & 0 \\ -2 & 1 & 1 \\ 0 & 1 & 4 \end{vmatrix} = 11$$

then $x = 11/2$, similarly, $y = -39/2, z = 7/2$
Basis: if \mathbf{c}, \mathbf{b} are two basis, the coordinates under \mathbf{c} is x_c , then the coordinates under \mathbf{b} will satisfy

$$Bx_b = Cx_c \implies x_b = B^{-1}Cx_c$$

Projection: Suppose there a vector λw on w , and the new vector $v - \lambda w$ is perpendicular to w , then

$$(v - \lambda w)^T w = 0 \implies \lambda = \frac{v^T w}{w^T w} \implies P_{ww} v = \frac{v^T w}{w^T w} w$$

suppose a eigenvalue λ_i has several eigenvectors ξ_i, ξ_{i+1}, \dots , we could use Gram-Schmidt process to get the orthogonal vectors

$$\eta_i = \xi_i \quad (1)$$

$$\eta_{i+1} = \xi_{i+1} - \frac{\xi_{i+1}^T \eta_i}{\eta_i^T \eta_i} \eta_i \quad (2)$$

$$\eta_{i+2} = \xi_{i+2} - \frac{\xi_{i+2}^T \eta_i}{\eta_i^T \eta_i} \eta_i - \frac{\xi_{i+2}^T \eta_{i+1}}{\eta_{i+1}^T \eta_{i+1}} \eta_{i+1} \quad (3)$$

Orthogonal: just vertical **Orthonormal:** normalized and orthogonal
 if A is symmetric, for any x we have $x^T A x > 0$, we say A is positive definite, if $x^T A x \geq 0$, we say A is semi-positive definite

Properties:

1. A is Positive-definite iff all Eigenvalues are positive
2. A is Postive Semi-definite iff all eigenvalues are non-negative
3. If A is Positive-definite, then so is A^{-1}
4. If $B_{n \times m}$ is any matrix, then $B^T B, B B^T$ are postive semi-definite

Laplace Transform

$$F(s) = \mathcal{L}[f(t)] = \int_0^\infty f(t) e^{-st} dt$$

$$\mathcal{L}[1] = \int_0^\infty e^{-st} dt = \frac{1}{s}$$

$$\mathcal{L}[t^n] = \int_0^\infty e^{-st} t^n dt = \frac{n!}{s^{n+1}}$$

$$\mathcal{L}[e^{-at}] = \int_0^\infty e^{-st} e^{-at} dt = \frac{1}{s+a}$$

$$\mathcal{L}[e^{at} f(t)] = F(s-a)$$

$$\text{Linearity: } \mathcal{L}[\alpha f(t) + \beta g(t)] = \alpha F(s) + \beta G(s)$$

$$\text{Scaling: } \mathcal{L}[f(at)] = \frac{1}{|a|} F\left(\frac{s}{a}\right)$$

$$\text{Multiplication: } \mathcal{L}[e^{ct} f(t)] = F(s-c) \mathcal{L}[t^n f(t)] = (-1)^n F^{(n)}(s)$$

Differentiation:

$$\mathcal{L}[f'(t)] = sF(s) - f(0) \quad \mathcal{L}[f''(t)] = s^2 F(s) - sf(0) - f'(0)$$

$$\mathcal{L}[f^{(n)}(t)] = s^n F(s) - s^{n-1} f(0) - s^{n-2} f'(0) - \dots - sf^{(n-2)}(0) - f^{(n-1)}(0)$$

$$\text{Convolution: } \mathcal{L}\left[\int_0^t f(\tau) g(t-\tau) d\tau\right] = F(s)G(s)$$

$$\text{Example: } x' + x = e^{-2t}$$

$$sX(s) - X(0) + X(s) = \frac{1}{s+2}$$

$$X(s) = \frac{x(0)}{s+1} + \frac{1}{s+1} - \frac{1}{s+2}$$

$$x(t) = x(0)e^{-t} + e^{-t} - e^{-2t}$$

1st Order Exact Equation

$$\text{if } \frac{\partial M}{\partial y} = \frac{\partial N}{\partial x}$$

then the solution is $u(x, y) = \int M dx + f(y) = \int N dy + g(x)$

given $P(x, t)dx + Q(x, t)dt = 0$

if $\frac{1}{P(x, t)} \left(\frac{\partial Q}{\partial x} - \frac{\partial P}{\partial t} \right) = u(t)$, then $e^{\int u(t) dt}$ is an integrating factor

if $\frac{1}{Q(x, t)} \left(\frac{\partial P}{\partial t} - \frac{\partial Q}{\partial x} \right) = v(x)$, then $e^{\int v(x) dx}$ is an integrating factor

Example: solve $2xdx + (x^2 - t)dt = 0$

$$\frac{1}{P(x, t)} \left(\frac{\partial Q}{\partial x} - \frac{\partial P}{\partial t} \right) = \frac{1}{2x} \times 2x = 1, \text{ then } e^{\int 1 dt} = e^t$$

$$u(x, t) = \int e^t 2xdx + f(t) = e^t x^2 + f(t)$$

$$= \int e^t (x^2 - t) dt + g(x) = e^t x^2 - \int e^t t dt + g(x)$$

$$= e^t x^2 - e^t t + e^t + g(x)$$

2nd Order ODE

for $y'' + ay' + by = 0$, we solve $\lambda^2 + a\lambda + b = 0$ and get the value λ_1, λ_2

- if $a^2 - 4b > 0$, then $y = C_1 e^{\lambda_1 x} + C_2 e^{\lambda_2 x}$

- if $a^2 - 4b = 0$, then $y = C_1 e^{\lambda x} + C_2 x e^{\lambda x}$

- if $a^2 - 4b < 0$, and $\lambda = \alpha \pm \beta i$, then $y = C_1 e^{\alpha x} \cos \beta x + C_2 e^{\alpha x} \sin \beta x$

Linear System

$$\frac{d}{dt} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \quad \mathbf{x} = \mathbf{P} \begin{bmatrix} C_1 e^{\lambda_1 t} \\ \vdots \\ C_n e^{\lambda_n t} \end{bmatrix}$$

we do diagonalize and the eigenvalue is $\Lambda = \text{diag}(\lambda_1, \dots, \lambda_n)$, then

$$\text{example } \frac{d}{dt} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$(A - 3I)u = 0 \Rightarrow \begin{bmatrix} -2 & 1 \\ 1 & -2 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$(A + I)u = 0 \Rightarrow \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \frac{1}{\sqrt{5}} \begin{bmatrix} -1 \\ 2 \end{bmatrix}$$

Formula

$$\sin A \cos B = \frac{1}{2} (\sin(A-B) + \sin(A+B))$$

$$\sin A \sin B = \frac{1}{2} (\cos(A-B) - \cos(A+B))$$

$$\cos A \cos B = \frac{1}{2} (\cos(A-B) + \cos(A+B))$$

$$\frac{d}{dx} a^x = a^x \ln a \quad \int \sec^2 x dx = \tan x + C$$

$$\frac{d}{dx} \ln |x| = \frac{1}{x} \quad \int \csc^2 x dx = -\cot x + C$$

$$\frac{d}{dx} \sin^{-1} x = \frac{1}{\sqrt{1-x^2}} \quad \int \sec x \tan x dx = \sec x + C$$

$$\frac{d}{dx} \cos^{-1} x = \frac{-1}{\sqrt{1-x^2}}$$

$$\frac{d}{dx} \tan^{-1} x = \frac{1}{x^2 + 1} \quad \int \csc x \cot x dx = -\csc x + C$$

$$\frac{d}{dx} \cot^{-1} x = \frac{-1}{x^2 + 1} \quad \int \frac{dx}{\sqrt{a^2 - x^2}} = \sin^{-1} \frac{x}{a} + C$$

$$\frac{d}{dx} \sec^{-1} x = \frac{1}{|x| \sqrt{x^2 - 1}} \quad \int \frac{dx}{a^2 + x^2} = \frac{1}{a} \tan^{-1} \frac{x}{a} + C$$

$$\frac{d}{dx} \csc^{-1} x = \frac{-1}{|x| \sqrt{x^2 - 1}} \quad \int \frac{dx}{x \sqrt{x^2 - a^2}} = \frac{1}{a} \sec^{-1} \frac{|x|}{a} + C$$

Calculus

Suppose all partial derivatives of the function $G(x_1, x_2, \dots, y)$ exists.
 and $G(x_1, x_2, \dots, y) = C$, then $\frac{\partial y}{\partial x_i} = -\frac{\partial G}{\partial x_i} / \frac{\partial G}{\partial y}$

The 2nd-order is

$$\frac{\partial}{\partial x_i} \left(\frac{\partial G}{\partial x_i} + \frac{\partial G}{\partial y} \frac{\partial y}{\partial x_i} \right) = \frac{\partial}{\partial x_i} \left(\frac{\partial G}{\partial x_i} \right) + \frac{\partial}{\partial x_i} \left(\frac{\partial G}{\partial y} \right) \frac{\partial y}{\partial x_i} + \frac{\partial G}{\partial y} \frac{\partial^2 y}{\partial x_i^2} = 0$$

and

$$\frac{\partial}{\partial x_i} \left(\frac{\partial G}{\partial x_i} \right) = \frac{\partial^2 G}{\partial x_i^2} + \frac{\partial^2 G}{\partial y \partial x_i} \frac{\partial y}{\partial x_i} \quad \frac{\partial}{\partial x_i} \left(\frac{\partial G}{\partial y} \right) = \frac{\partial^2 G}{\partial x_i \partial y} + \frac{\partial^2 G}{\partial y^2} \frac{\partial y}{\partial x_i}$$

spherical polar coordinates:

$$r = (x^2 + y^2 + z^2)^{1/2}, \quad \theta = \cos^{-1} \frac{z}{(x^2 + y^2 + z^2)^{1/2}}, \quad \phi = \tan^{-1} \frac{y}{x}$$

$$x = r \sin \theta \cos \phi, \quad y = r \sin \theta \sin \phi, \quad z = r \cos \theta$$

Hessian Matrix: If all second partial derivatives are continuous, then Hessian Matrix is symmetric

$$[\nabla^2 f] = \frac{\partial^2 f}{\partial x \partial x^T} = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} & \dots & \frac{\partial^2 f}{\partial x_2 \partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \frac{\partial^2 f}{\partial x_n \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}$$

$$\frac{\partial \mathbf{x}^T \mathbf{A} \mathbf{x}}{\partial \mathbf{x}^T} = (\mathbf{A} + \mathbf{A}^T) \mathbf{x} = 2\mathbf{A} \mathbf{x}$$

Taylor Expansion: $f(x) = \sum_{k=0}^n \frac{f^{(k)}(x_0)}{k!} (x - x_0)^k + R_n(x)$

$$f(x) = f(x_0) + (x - x_0)^T \nabla f(x_0) + \frac{1}{2!} (x - x_0)^T [\nabla^2 f(x_0)] (x - x_0) + \dots$$

Notice: the quadratic term has coefficient $1/2!$, but this not apply to $(x - x_0)(y - y_0)$

if $(x, y, z) \rightarrow (u, v, w)$, then Jacob matrix is

$$\mathbf{J} = \begin{bmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} & \frac{\partial x}{\partial w} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} & \frac{\partial y}{\partial w} \\ \frac{\partial z}{\partial u} & \frac{\partial z}{\partial v} & \frac{\partial z}{\partial w} \end{bmatrix}$$

and

$$\int_R f(x, y, z) dV(x, y, z)$$

$$= \int_S f(x(u, v, w), y(u, v, w), z(u, v, w)) |det(J)| dV(u, v, w)$$

Example: $\iint_{x^2 + y^2 \leq 4} (x^2 + y^2 + 3) dA(x, y)$

$$\det(\mathbf{J}) = \begin{vmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{vmatrix} = r \cos^2 \theta - (-r \sin^2 \theta) = r$$

$$\int_0^{2\pi} \int_0^2 (r^2 + 3) r dr d\theta = \int_0^{2\pi} \left[\frac{r^4}{4} + \frac{3r^2}{2} \right]_{r=0}^{r=2} d\theta = 20\pi$$

for spherical polar coordinates:

$$\iiint_R f(x, y, z) dV = \iiint_S f(r \sin \theta \cos \phi, r \sin \theta \sin \phi, r \cos \theta) r^2 \sin \theta dr d\theta d\phi$$

Optimization

For $f(x)$, $x = a$ is a critical point if $\nabla f(a) = 0$

To determine the nature of critical points at $x = a$, we need to analyze the eigenvalues $\lambda_1, \dots, \lambda_n$ of the Hessian Matrix of f :

- $\nabla f(a) = 0, \lambda_i(a) < 0$: maximum
- $\nabla f(a) = 0, \lambda_i(a) > 0$: minimum
- $\nabla f(a) = 0$, mixed signs: saddle point

if Hessian Matrix A is positive semi-definite (i.e. $A \geq 0$), then local minimum is global minimum

Example

$$\frac{1}{2}v\frac{\partial^2 f}{\partial x^2} + \rho\sigma v\frac{\partial^2 f}{\partial x\partial v} + \frac{1}{2}\sigma^2 v\frac{\partial^2 f}{\partial v^2} + (r + \beta v)\frac{\partial f}{\partial x} + (a - bv)\frac{\partial f}{\partial v} + \frac{\partial f}{\partial t} = 0,$$

$$f(x, v, T) = e^{\phi x}$$

(a) with an ansatz $f(x, v, t) = \exp\{C(t) + D(t)v + \phi x\}$, Derive the equations driving $C(t)$ and $D(t)$

(b) Suppose that the function $D(t)$ for $t \in [0, T]$ is solvable. Solve $C(t)$

(c) suppose $\sigma = 0$, solve for $D(t)$

$$\begin{aligned} \frac{\phi^2}{2}vf + \rho\sigma\phi D(t)vf + \frac{1}{2}\sigma^2 D^2(t)vf + (r + \beta v)\phi f + (a - bv)D(t)f \\ + \left(\frac{dC}{dt} + \frac{dD}{dt}v\right)f = 0 \end{aligned}$$

with boundary condition $C(T) = 0, D(T) = 0$, since $f \neq 0$, we can organize the PDE in the order of v as follows:

$$\left[\frac{dD}{dt} + \frac{\sigma^2}{2}D^2(t) + (\rho\sigma\phi - b)D(t) + \frac{\phi^2}{2} + \beta\phi\right]v + \left[\frac{dC}{dt} + aD(t) + r\right] = 0$$

Note that the equation holds for all v . Hence, we have

$$\frac{dC}{dt} + aD(t) + r = 0, \quad C(T) = 0; \quad (4)$$

$$\frac{dD}{dt} + \frac{\sigma^2}{2}D^2(t) + (\rho\sigma\phi - b)D(t) + \frac{\phi^2}{2} + \beta\phi = 0, \quad D(T) = 0. \quad (5)$$

(b) Integrating both sides of the 1st equation from t to T , we have

$$C(t) = r(T - t) + a \int_t^T D(s)ds$$

(c) In this case, we have

$$\frac{dD}{dt} - bD(t) + \frac{\phi^2}{2} + \beta\phi = 0 \iff \frac{d(e^{-bt}D(t))}{dt} + e^{-bt}\left(\frac{\phi^2}{2} + \beta\phi\right) = 0$$

Integrating from t to T

$$e^{-bt}D(t) = \left(\frac{\phi^2}{2} + \beta\phi\right) \int_t^T e^{-bs}ds \implies D(t) = \frac{\phi(\phi + 2\beta)}{2b}(1 - e^{-b(T-t)})$$

Example

$$\frac{1}{x+z} + \frac{\partial z}{\partial x} \left(\frac{1}{x+z} + e^y + x^2 \right) + 2xz = 0 \implies \frac{\partial z}{\partial x} \Big|_{(x,y,z)=(0,0,1)} = -\frac{1}{2}$$

$$\frac{\partial z}{\partial y} \left(\frac{1}{x+z} + e^y + x^2 \right) + ze^y = 0 \implies \frac{\partial z}{\partial y} \Big|_{(x,y,z)=(0,0,1)} = -\frac{1}{2}$$

$$\frac{\partial^2 z}{\partial x \partial y} \left(\frac{1}{x+z} + e^y + x^2 \right) + \frac{\partial z}{\partial y} \left(-\frac{1}{(x+z)^2} \left(1 + \frac{\partial z}{\partial x} \right) + 2x \right) + \frac{\partial z}{\partial x} e^y = 0$$

$$\implies \frac{\partial^2 z}{\partial x \partial y} \Big|_{(x,y,z)=(0,0,1)} = \frac{1}{8}$$

$$\frac{\partial^2 z}{\partial y^2} \left(\frac{1}{x+z} + e^y + x^2 \right) + \frac{\partial z}{\partial y} \left(-\frac{1}{(x+z)^2} \frac{\partial z}{\partial y} + e^y \right) + \frac{\partial z}{\partial y} e^y + ze^y = 0$$

$$\implies \frac{\partial^2 z}{\partial y^2} \Big|_{(x,y,z)=(0,0,1)} = \frac{1}{8}$$

$$\begin{aligned} -\frac{1}{(x+z)^2} \left(1 + \frac{\partial z}{\partial x} \right) + \frac{\partial^2 z}{\partial x^2} \left(\frac{1}{x+z} + e^y + x^2 \right) \\ + \frac{\partial z}{\partial x} \left(-\frac{1}{(x+z)^2} \left(1 + \frac{\partial z}{\partial x} \right) + 2x \right) + 2z + 2x \frac{\partial z}{\partial x} = 0 \end{aligned}$$

$$\implies \frac{\partial^2 z}{\partial x^2} \Big|_{(x,y,z)=(0,0,1)} = -\frac{7}{8}$$

Hence

$$z(x, y) \approx 1 - \frac{1}{2}x - \frac{1}{2}y - \frac{7}{16}x^2 + \frac{1}{8}xy + \frac{1}{16}y^2$$

Example

$$\min_w \sigma(w) := \sqrt{w^\top \Sigma w} \text{ subject to } r_f + w^\top (\mu - r_f \mathbf{1}_p) = r^*$$

The problem is equivalent to minimizing $\sigma^2(w) = w^\top \Sigma w$ because square function is an increasing function. Define Lagrangian as follows

$$L(w; \lambda) = w^\top \Sigma w + \lambda(r_f + w^\top (\mu - r_f \mathbf{1}_p) - r^*)$$

Setting its derivative with respect to w equal to 0 yields that

$$2\Sigma w + \lambda(\mu - r_f \mathbf{1}_p) = 0 \implies w^*(\lambda) = -\frac{\lambda}{2}\Sigma^{-1}(\mu - r_f \mathbf{1}_p)$$

Plug it into constraint

$$r_f - \frac{\lambda}{2}(\mu - r_f \mathbf{1}_p)^\top \Sigma^{-1}(\mu - r_f \mathbf{1}_p) = r^*$$

$$\implies \lambda^* = \frac{-2(r^* - r_f)}{(\mu - r_f \mathbf{1}_p)^\top \Sigma^{-1}(\mu - r_f \mathbf{1}_p)}$$

Hence

$$w^* = \frac{r^* - r_f}{(\mu - r_f \mathbf{1}_p)^\top \Sigma^{-1}(\mu - r_f \mathbf{1}_p)} \Sigma^{-1}(\mu - r_f \mathbf{1}_p)$$

$$\sigma^* = \sigma(w^*) = \frac{r^* - r_f}{\sqrt{(\mu - r_f \mathbf{1}_p)^\top \Sigma^{-1}(\mu - r_f \mathbf{1}_p)}}$$

Example

$$V_t + rxV_x - \frac{\theta}{2} \frac{V_x^2}{V_{xx}} = 0 \quad \text{with} \quad V(T, x) = \frac{x^{1-\gamma}}{1-\gamma}$$

with an ansatz $V(t, x) = g(t) \frac{x^{1-\gamma}}{1-\gamma}$ derive the equation driving $g(t)$ and solve for it

$$V_t = \frac{dg}{dt} \frac{x^{1-\gamma}}{1-\gamma}, V_x = g(t)x^{-\gamma}, V_{xx} = -\gamma g(t)x^{-\gamma-1}$$

Hence, PDE becomes

$$\frac{dg}{dt} \frac{x^{1-\gamma}}{1-\gamma} + rg(t)x^{1-\gamma} + \frac{\theta}{2} \frac{g^2(t)x^{-2\gamma}}{\gamma g(t)x^{-\gamma-1}} = 0, \quad g(T) = 1$$

$$\Leftrightarrow \frac{dg}{dt} + r(1-\gamma)g(t) + \frac{\theta(1-\gamma)}{2\gamma}g(t) = 0, \quad g(T) = 1$$

$$\frac{d(\ln g)}{dt} = -r(1-\gamma) - \frac{\theta(1-\gamma)}{2\gamma}, \quad g(T) = 1$$

Integrating from t to T

$$-\ln g(t) = \ln g(T) - \ln g(t) = -\int_t^T \left[r(1-\gamma) + \frac{\theta(1-\gamma)}{2\gamma} \right] ds$$

$$\Leftrightarrow g(t) = \exp \left(\left[r(1-\gamma) + \frac{\theta(1-\gamma)}{2\gamma} \right] (T-t) \right)$$

Example

let

$$\bar{x}_n = \frac{1}{n} \sum_{i=1}^n x_i, \quad e_n = x_n - \bar{x}_n, \quad Q_n = \sum_{i=1}^n (x_i - \bar{x}_n)(x_i - \bar{x}_n)^\top$$

Prove

$$\bar{x}_{n-1} = \bar{x}_n - \frac{e_n}{n-1}, \quad Q_{n-1} = Q_n - \left(1 + \frac{1}{n-1} \right) e_n e_n^\top$$

$$Q_{n-1} = \sum_{i=1}^{n-1} (x_i - \bar{x}_{n-1})(x_i - \bar{x}_{n-1})^\top$$

$$= \sum_{i=1}^{n-1} \left(x_i - \bar{x}_n + \frac{e_n}{n-1} \right) \left(x_i - \bar{x}_n + \frac{e_n}{n-1} \right)^\top$$

$$= \sum_{i=1}^{n-1} [(x_i - \bar{x}_n)(x_i - \bar{x}_n)^\top + (x_i - \bar{x}_n) \frac{e_n^\top}{n-1} +$$

$$\frac{e_n}{n-1} (x_i - \bar{x}_n)^\top + \frac{e_n e_n^\top}{(n-1)^2}]$$

$$= Q_n - e_n e_n^\top - \frac{e_n e_n^\top}{n-1} - \frac{e_n e_n^\top}{n-1} + \frac{e_n e_n^\top}{n-1}$$

$$= Q_n - \left(1 + \frac{1}{n-1} \right) e_n e_n^\top$$