

# VEKTÖR ANALİZİ

2/2/2021, DERS 2

İlker  
Bırbıl

# TENSÖRLER

$$A \in \mathbb{R}^{4 \times 2}$$



$$x \in \mathbb{R}^3$$



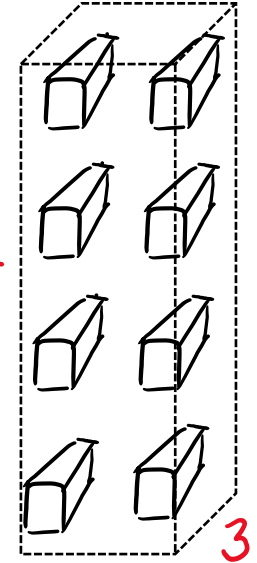
$$\frac{dA}{dx} \in \mathbb{R}^?$$

$$A = \begin{bmatrix} A_{11}(x) & A_{12}(x) \\ A_{21}(x) & A_{22}(x) \\ A_{31}(x) & A_{32}(x) \\ A_{41}(x) & A_{42}(x) \end{bmatrix}$$

$$\frac{dA}{dx} =$$

$$\begin{bmatrix} \nabla_x A_{11} & \nabla_x A_{12} \\ \nabla_x A_{21} & \nabla_x A_{22} \\ \nabla_x A_{31} & \nabla_x A_{32} \\ \nabla_x A_{41} & \nabla_x A_{42} \end{bmatrix}$$

= 4



2

3

TENSÖR

$$A_{ij}: \mathbb{R}^3 \rightarrow \mathbb{R}$$

$$\nabla_x A_{ij} = \left[ \frac{\partial A_{ij}}{\partial x_1} \quad \frac{\partial A_{ij}}{\partial x_2} \quad \frac{\partial A_{ij}}{\partial x_3} \right]$$



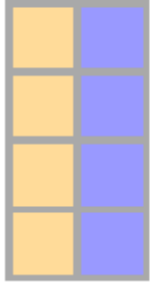
3

1

$$\frac{dA}{dx} \in \mathbb{R}^{4 \times 2 \times 3}$$

# TENSÖRLER

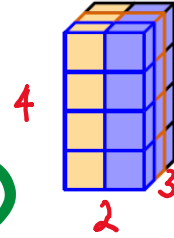
$$A \in \mathbb{R}^{4 \times 2}$$



$$x \in \mathbb{R}^3$$

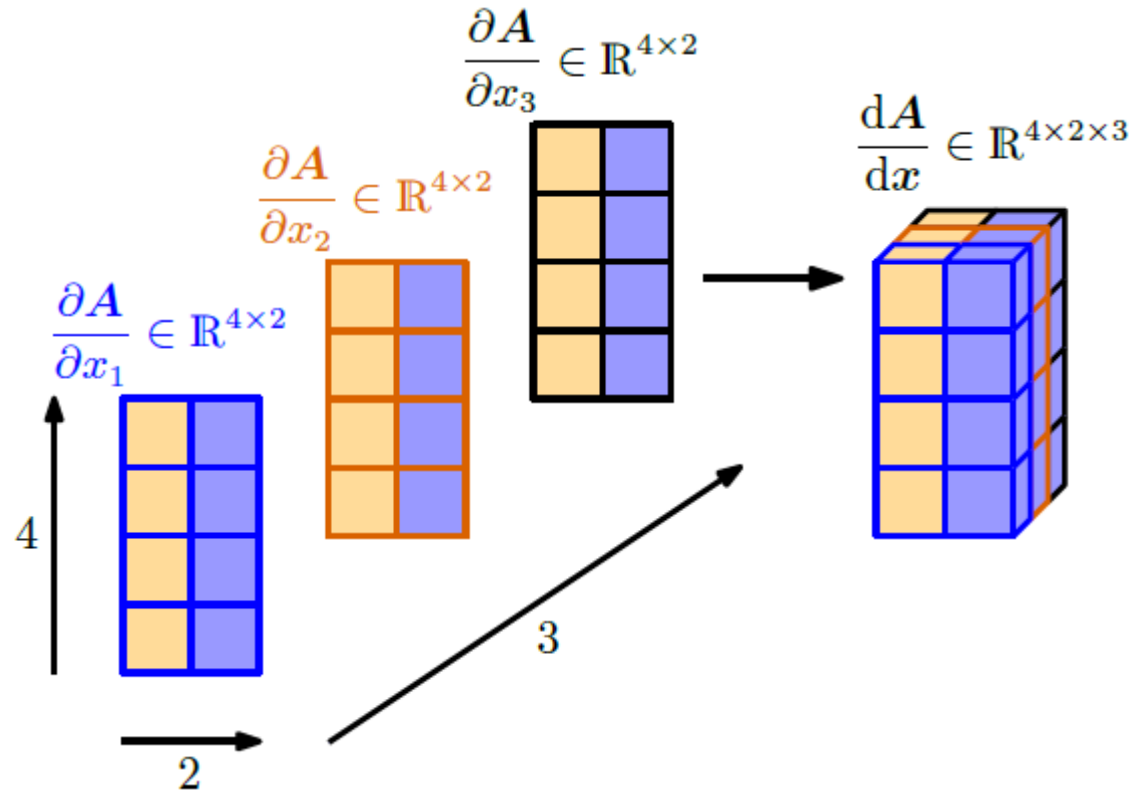


$$\frac{dA}{dx} \in \mathbb{R}^{4 \times 2 \times 3}$$



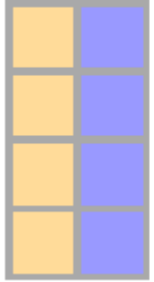
TENSÖR

1. YÖL



# TENSÖRLER

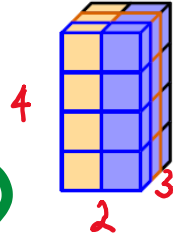
$$A \in \mathbb{R}^{4 \times 2}$$



$$x \in \mathbb{R}^3$$



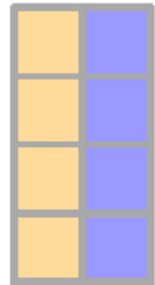
$$\frac{dA}{dx} \in \mathbb{R}^{4 \times 2 \times 3}$$



TENSÖR

2.YOL

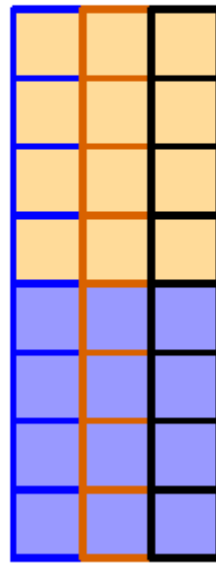
$$A \in \mathbb{R}^{4 \times 2}$$



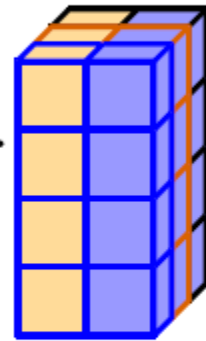
$$\tilde{A} \in \mathbb{R}^8$$



$$\frac{d\tilde{A}}{dx} \in \mathbb{R}^{8 \times 3}$$



$$\frac{dA}{dx} \in \mathbb{R}^{4 \times 2 \times 3}$$



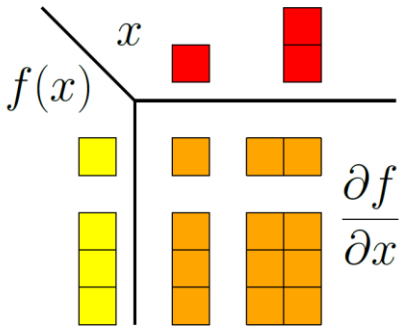
# TENSÖRLER

$$f = Ax, \quad f \in \mathbb{R}^M, \quad A \in \mathbb{R}^{M \times N}, \quad x \in \mathbb{R}^N$$

$$\frac{df}{dA} \in \mathbb{R}^?$$

$$f: \mathbb{R}^{M \times N} \rightarrow \mathbb{R}^M, \quad \frac{df}{dA} = \begin{matrix} M \times N \\ M \end{matrix} \left[ \begin{matrix} M \times N \end{matrix} \right] \in \mathbb{R}^{M \times (M \times N)}$$

TENSÖR



$$f(R) = R^T R =: K \in \mathbb{R}^{N \times N}, \quad R \in \mathbb{R}^{M \times N}$$

$$\frac{dK}{dR} \in \mathbb{R}^?$$

$$f: \mathbb{R}^{M \times N} \rightarrow \mathbb{R}^{N \times N}, \quad \frac{dK}{dR} = \begin{matrix} N \times N \\ M \times N \end{matrix} \left[ \begin{matrix} M \times N \end{matrix} \right] \in \mathbb{R}^{(N \times N) \times (M \times N)}$$

TENSÖR



KISA YOLLAR

$$\frac{\partial}{\partial X} f(X)^\top = \left( \frac{\partial f(X)}{\partial X} \right)^\top$$

$$\frac{\partial}{\partial X} \text{tr}(f(X)) = \text{tr} \left( \frac{\partial f(X)}{\partial X} \right)$$

$$\frac{\partial}{\partial X} \det(f(X)) = \det(f(X)) \text{tr} \left( f(X)^{-1} \frac{\partial f(X)}{\partial X} \right)$$

$$\frac{\partial}{\partial X} f(X)^{-1} = -f(X)^{-1} \frac{\partial f(X)}{\partial X} f(X)^{-1}$$

$$\frac{\partial a^\top X^{-1} b}{\partial X} = -(X^{-1})^\top a b^\top (X^{-1})^\top$$

$$\frac{\partial x^\top a}{\partial x} = a^\top$$

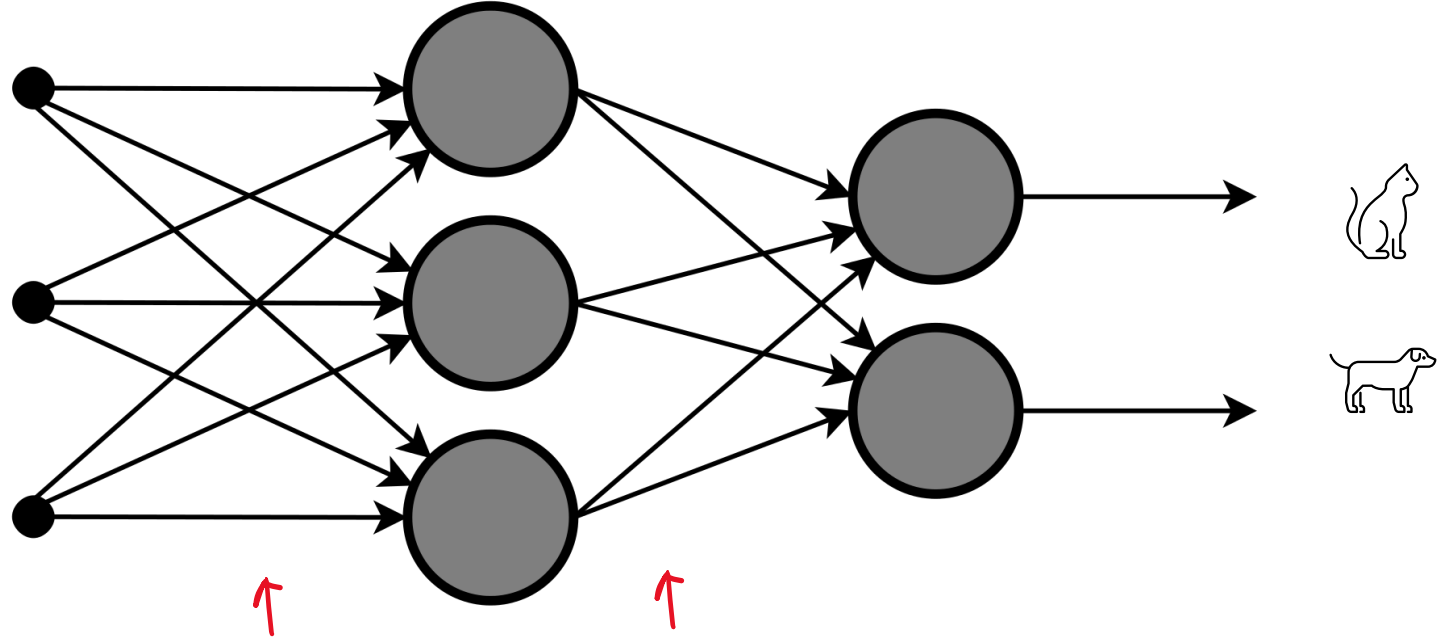
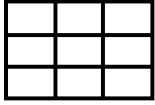
$$\frac{\partial a^\top x}{\partial x} = a^\top$$

$$\frac{\partial a^\top X b}{\partial X} = a b^\top$$

$$\frac{\partial x^\top B x}{\partial x} = x^\top (B + B^\top)$$

$$\frac{\partial}{\partial s} (x - A s)^\top W (x - A s) = -2(x - A s)^\top W A \quad (W^\top = W)$$

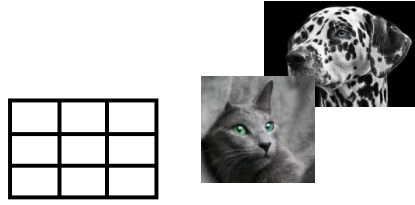
# GERİ YAYILIM



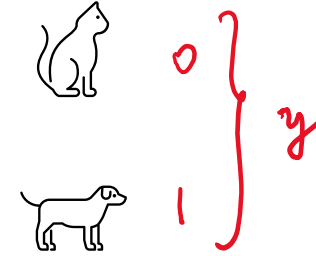
(kaynak)

PARAMETRELER

# GERİ YAYILIM

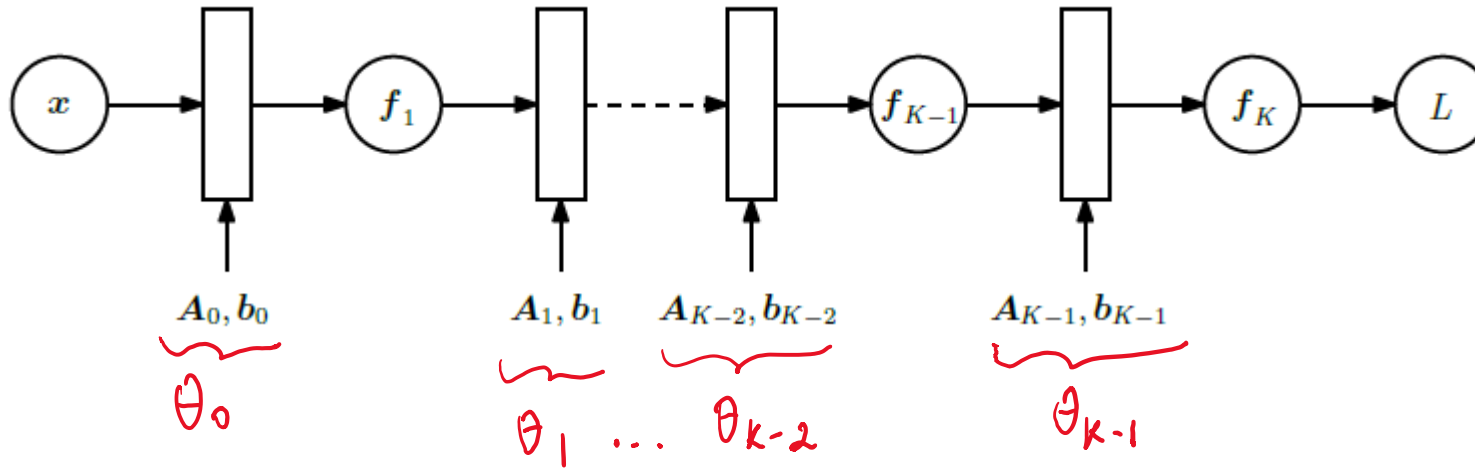


$$(f_K \circ f_{K-1} \circ \dots \circ f_1)(x) = f_K(f_{K-1}(\dots(f_1(x))\dots))$$



$$f_0 := x$$

$$f_i := \sigma_i(A_{i-1}f_{i-1} + b_{i-1}), \quad i = 1, \dots, K$$



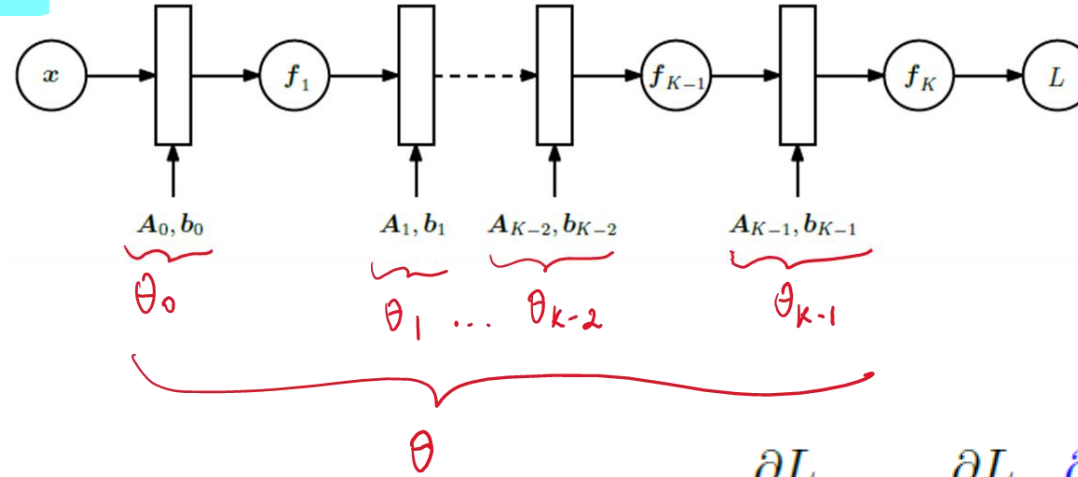
$$\min_{\theta} \{ \underbrace{\|y - f_K(\theta)\|^2}_{L(\theta)} \}$$

$\theta$  PARAMETRELER

$$\frac{\partial L}{\partial \theta} = ?$$



# GERİ YAYILIM

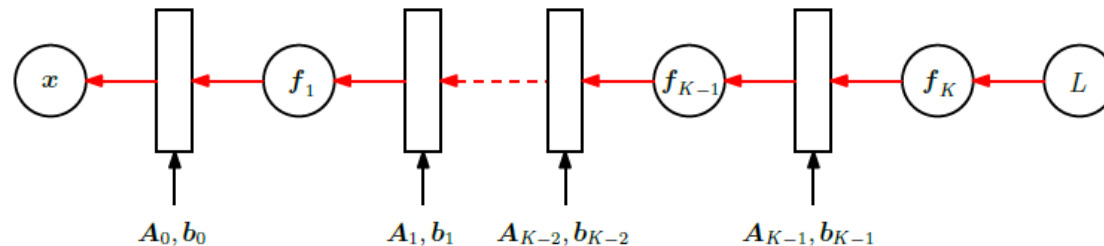


$$\min_{\theta} \{ \underbrace{\|y - f_K(\theta)\|^2}_{L(\theta)} \}$$

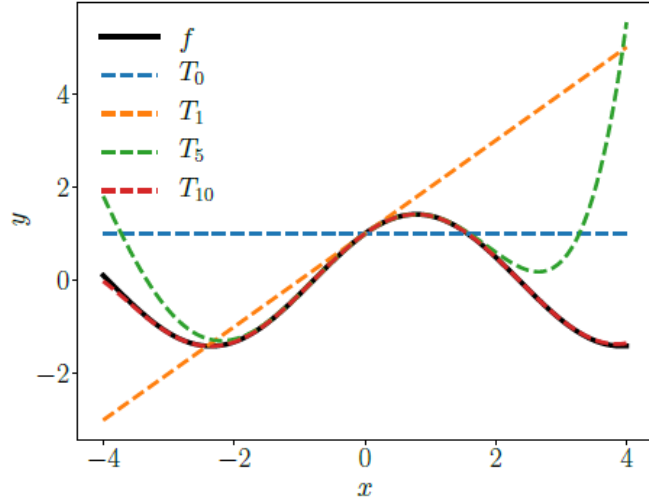
$$\frac{\partial L}{\partial \theta_{K-1}} = \frac{\partial L}{\partial f_K} \frac{\partial f_K}{\partial \theta_{K-1}}$$

$$\frac{\partial L}{\partial \theta_{K-2}} = \frac{\partial L}{\partial f_K} \left[ \frac{\partial f_K}{\partial f_{K-1}} \frac{\partial f_{K-1}}{\partial \theta_{K-2}} \right]$$

$$\frac{\partial L}{\partial \theta_{K-3}} = \frac{\partial L}{\partial f_K} \frac{\partial f_K}{\partial f_{K-1}} \left[ \frac{\partial f_{K-1}}{\partial f_{K-2}} \frac{\partial f_{K-2}}{\partial \theta_{K-3}} \right]$$



# İKİNCİ DERECE



$$f: \mathbb{R}^n \rightarrow \mathbb{R} \quad \nabla f = \left[ \frac{\partial f}{\partial x_1} \quad \frac{\partial f}{\partial x_2} \quad \dots \quad \frac{\partial f}{\partial x_n} \right]$$

$$\nabla^2 f = H = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \frac{\partial^2 f}{\partial x_n \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix} \quad (H = H^T)$$

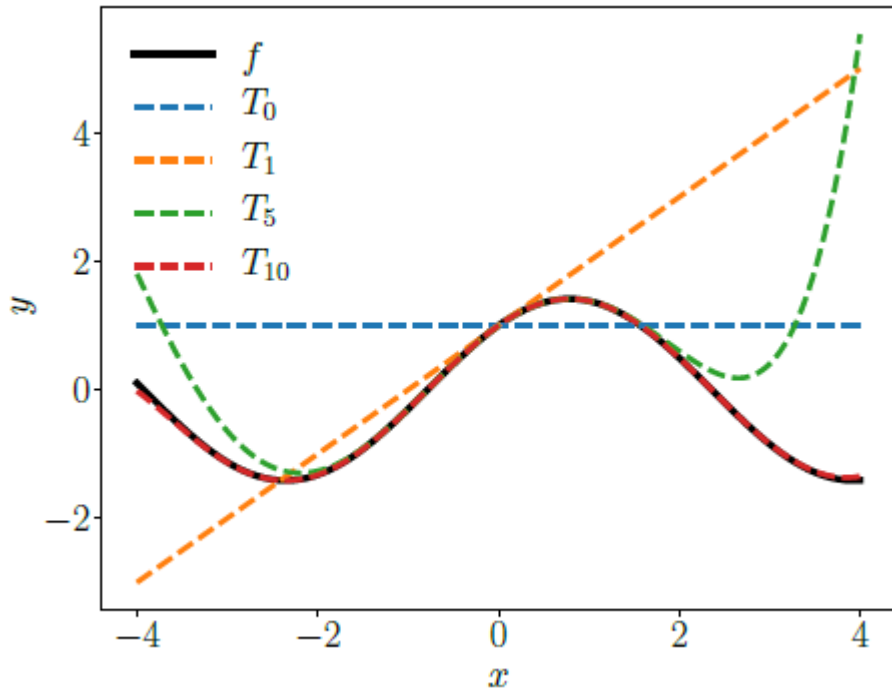
HESSEN  
MATRİSİ

$$f(x) \approx f(x_0) + \nabla f(x_0)(x - x_0) + \frac{1}{2} (x - x_0)^T \nabla^2 f(x_0) (x - x_0)$$

# İKİNCİ DEĞER

$x \in \mathbb{R}^n$

$$f(x) \approx f(x_0) + \nabla f(x_0)(x-x_0) + \frac{1}{2} (x-x_0)^T \nabla^2 f(x_0) (x-x_0)$$



$$T_{\infty}(x) = \sum_{k=0}^{\infty} \frac{f^{(k)}(x_0)}{k!} (x - x_0)^k$$

$$f(x) \approx f(x_0) + f^{(1)}(x_0)(x-x_0) + \frac{1}{2} f^{(2)}(x_0)(x-x_0)^2$$

$x \in \mathbb{R}$

$$\frac{1}{2} (x-x_0) f^{(2)}(x_0) (x-x_0)$$

# ÖDEV

5.4 Compute the Taylor polynomials  $T_n$ ,  $n = 0, \dots, 5$  of  $f(x) = \sin(x) + \cos(x)$

5.6 Differentiate  $f$  with respect to  $t$  and  $g$  with respect to  $\mathbf{X}$ , where

5.9 We define

$$g(\mathbf{z}, \boldsymbol{\nu}) := \log p(\mathbf{x}, \mathbf{z}) - \log q(\mathbf{z}, \boldsymbol{\nu})$$

$$\mathbf{z} := t(\boldsymbol{\epsilon}, \boldsymbol{\nu})$$

for differentiable functions  $p, q, t$ , and  $\mathbf{x} \in \mathbb{R}^D, \mathbf{z} \in \mathbb{R}^E, \boldsymbol{\nu} \in \mathbb{R}^F, \boldsymbol{\epsilon} \in \mathbb{R}^G$ . By using the chain rule, compute the gradient

$$\frac{d}{d\boldsymbol{\nu}} g(\mathbf{z}, \boldsymbol{\nu}) .$$