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# Time Series: Homework 1

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## Exercise 1.1.

Suppose that  $X_t = Z_t + \theta Z_{t-1}$ ,  $t = 1, 2, \dots$ , where  $Z_0, Z_1, Z_2$  are independent random variables, each with moment generating function  $\mathbf{E} \exp(\lambda Z_i) = m(\lambda)$ .

- (a) Express the joint moment generating function  $\mathbf{E} \exp(\sum_{i=1}^n \lambda_i X_i)$  in terms of the function  $m(\cdot)$ .
- (b) Deduce from (a) that  $\{X_t\}$  is strictly stationary.

## Solution part (a)

Since  $\{Z_t\}$  are independent, for  $X_t = Z_t + \theta Z_{t-1}$ , the moment generating function:

$$\begin{aligned}\mathbf{E} \exp(\lambda X_t) &= \mathbf{E} \exp(\lambda(Z_t + \theta Z_{t-1})) \\ &= \mathbf{E} \exp(\lambda Z_t) \cdot \mathbf{E} \exp(\lambda \theta Z_{t-1}) \\ &= m(\lambda) \cdot m(\theta \lambda)\end{aligned}$$

On the other hand,

$$\begin{aligned}\sum_{i=1}^n \lambda_i X_i &= \sum_{i=1}^n \lambda_i Z_i + \sum_{i=1}^n \lambda_i \theta Z_{i-1} \\ &= \sum_{i=1}^n \lambda_i Z_i + \sum_{i=0}^{n-1} \lambda_{i+1} \theta Z_i \\ &= \lambda_n Z_n + \left[ \sum_{i=1}^{n-1} (\lambda_i + \theta \lambda_{i+1}) Z_i \right] + \lambda_1 \theta Z_0.\end{aligned}$$

Therefore, using a similar argument as before

$$\mathbf{E} \exp \left( \sum_{i=1}^n \lambda_i X_i \right) = m(\lambda_n) \cdot \left[ \prod_{i=1}^{n-1} m(\lambda_i + \theta \lambda_{i+1}) \right] \cdot m(\lambda_1 \theta)$$

### Solution part (b)

Let  $(X_1, \dots, X_n)'$  be a random vector in  $\mathbb{R}^k$ . The moment generating function of this vector is defined as follows,

$$M_{X_{1:n}}(\lambda_{1:n}) = \mathbf{E} \exp(\langle \lambda_{1:n}, X_{1:n} \rangle) = \mathbf{E} \exp \left( \sum_{i=1}^n \lambda_i X_i \right), \lambda_{1:n} \in \mathbb{R}^n.$$

So, we know from the previous part that

$$\begin{aligned} M_{X_{1:n}}(\lambda_{1:n}) &= m(\lambda_n) \cdot \left[ \prod_{i=1}^{n-1} m(\lambda_i + \theta \lambda_{i+1}) \right] \cdot m(\lambda_1 \theta) \\ &= \mathbf{E} \exp(\lambda_n Z_{n+h}) + \left[ \prod_{i=1}^{n-1} \mathbf{E} \exp((\lambda_i + \theta \lambda_{i+1}) Z_{i+h}) \right] + \mathbf{E} \exp(\lambda_1 \theta Z_h) \\ &= \mathbf{E} \exp \left( \sum_{i=1}^n \lambda_i X_{i+h} \right) \\ &= M_{X_{1+h:n+h}}(\lambda_{1:n}) \end{aligned}$$

Since the moment generating function of both  $(X_1, \dots, X_n)'$  and  $(X_{1+h}, \dots, X_{n+h})'$  coincide, they have the same joint distribution. Thus,  $\{X_t\}$  is strictly stationary.

### Exercise 1.4.

If  $m_t = \sum_{k=0}^p c_k t^k$ ,  $t = 0, \pm 1, \dots$ , show that  $\nabla m_t$  is a polynomial of degree  $(p-1)$  in  $t$  and hence that  $\nabla^{p+1} m_t = 0$ .

**Solution:**

$$\begin{aligned} m_{t-1} &= \sum_{k=0}^p c_k (t-1)^k \\ &= \sum_{k=0}^p c_k \sum_{j=0}^k \binom{k}{j} t^j (-1)^{k-j} \\ &= \sum_{j=0}^p t^j \sum_{k=j}^p \binom{k}{j} (-1)^{k-j} c_k \end{aligned}$$

The last line can be deduced from the following diagram

$$\begin{aligned}
m_{t-1} &= \sum_{k=0}^p c_k \sum_{j=0}^k \binom{k}{j} t^j (-1)^{k-j} = \\
& c_0 \binom{0}{0} t^0 (-1)^{0-0} \\
& c_1 \binom{1}{0} t^0 (-1)^{1-0} + c_1 \binom{1}{1} t^1 (-1)^{1-1} \\
& c_2 \binom{2}{0} t^0 (-1)^{2-0} + c_2 \binom{2}{1} t^1 (-1)^{2-1} + c_2 \binom{2}{2} t^2 (-1)^{2-2} \\
& \vdots \qquad \qquad \qquad \vdots \qquad \qquad \qquad \vdots \qquad \qquad \qquad \ddots \\
& c_p \binom{p}{0} t^0 (-1)^{p-0} + c_p \binom{p}{1} t^1 (-1)^{p-1} + c_p \binom{p}{2} t^2 (-1)^{p-2} + \dots + c_p \binom{p}{p} t^p (-1)^{p-p} \\
& = \qquad \qquad \qquad = \qquad \qquad \qquad = \qquad \qquad \qquad \dots \qquad \qquad = \\
& t^0 \sum_{k=0}^p c_k \binom{k}{0} (-1)^{k-0} + t^1 \sum_{k=1}^p c_k \binom{k}{1} (-1)^{k-1} + t^2 \sum_{k=2}^p c_k \binom{k}{2} (-1)^{k-2} + \dots + t^p \sum_{k=p}^p c_k \binom{k}{p} (-1)^{k-p}
\end{aligned}$$

Thus, for  $j = p$ , the coefficient that accompanies  $t^p$  is  $\binom{p}{p}(-1)^{p-p}c_p = c_p$ . So it follows that

$$\begin{aligned}
\nabla m_t &= \sum_{j=0}^p c_j t^j - \sum_{j=0}^p t^j \sum_{k=j}^p \binom{k}{j} (-1)^{k-j} c_k \\
&= c_p t^p + \sum_{j=0}^{p-1} c_j t^j - c_p t^p - \sum_{j=0}^{p-1} t^j \sum_{k=j}^p \binom{k}{j} (-1)^{k-j} c_k \\
&= \sum_{j=0}^{p-1} t^j \cdot \left[ c_j - \sum_{k=j}^p \binom{k}{j} (-1)^{k-j} c_k \right],
\end{aligned}$$

which is a  $(p-1)$ -degree polynomial. Now, note that

$$\nabla^n m_t = (I - B)^n(m_t).$$

One can inductively show that  $\nabla^n m_t$  has degree  $p-n$  for any polynomial  $m_t$  of degree  $p$ . We proved the base case previously, so assume that  $\nabla^{n-1} m_t$  has degree  $p-n+1$ . Then, define  $d_j = [\nabla^{n-1} m_t]_{t^j}$  as the coefficient that accompanies  $t^j$ .

Since we proved that  $(I - B)$  reduces by one the degree of any polynomial, it follows that  $(I - B)\nabla^{n-1} m_t$  has degree  $(p-n+1) - 1 = p-n$ . This can be verified with the following

calculation:

$$\begin{aligned}
\nabla^n m_t &= (I - B)(I - B)^{n-1} m_t \\
&= (I - B) \nabla^{n-1} m_t \\
&= \nabla \left( \sum_{k=0}^{p-n+1} d_k t^k \right) \\
&= \sum_{j=0}^{p-n} t^j \cdot \left[ d_j - \sum_{k=j}^{p-n+1} \binom{k}{j} (-1)^{k-j} d_k \right].
\end{aligned}$$

Finally,  $\nabla^p m_t$  is polynomial of degree 0, and thus, it's a constant function  $f_t = K$ . Therefore,

$$\begin{aligned}
\nabla^{p+1} m_t &= (I - B)(\nabla^p m_t) \\
&= (I - B)(K t^0) \\
&= K - BK \\
&= K - K = 0.
\end{aligned}$$

The backwards shift operator evaluated on a constant is the same constant since  $f_t = f_{t-1} = K$  for a constant function  $f_t$ .

## Exercise 1.7.

Let  $Z_t, t = 0, \pm 1, \dots$ , be independent normal random variables each with mean 0 and variance  $\sigma^2$  and let  $a, b$  and  $c$  be constants. Which, if any, of the following processes are stationary? For each stationary process specify the mean and autocovariance function.

- (a)  $X_t = a + bZ_t + cZ_{t-1}$ ,
- (c)  $X_t = Z_1 \cos(ct) + Z_2 \sin(ct)$ ,
- (e)  $X_t = Z_t \cos(ct) + Z_{t-1} \sin(ct)$

**Note:** I assumed by mistake that  $\sigma^2 = 1$ . However, in all of the equations on the following solution, the  $\sigma^2$  term can always be factorized without altering the truth value of the propositions.

### Solution part (a)

Using the linearity of the expected value and the variance ( $Z_t$ 's are independent)

$$\mathbf{E}X_t = a + b\mathbf{E}Z_t + c\mathbf{E}Z_{t-1} = a$$

$$\begin{aligned}\mathbf{Var}(X_t) &= b^2\mathbf{Var}(Z_t) + c^2\mathbf{Var}(Z_{t-1}) = b^2 + c^2 \\ \implies \mathbf{E}|X_t|^2 &= \mathbf{Var}(X_t) + (\mathbf{E}X_t)^2 = a^2 + b^2 + c^2 < \infty\end{aligned}$$

Now for the autocovariance function,

$$\begin{aligned}\gamma_X(r, s) &= \mathbf{E}[(X_r - a)(X_s - a)] \\ &= \mathbf{E}[(bZ_r + cZ_{r-1})]\mathbf{E}[(bZ_s + cZ_{s-1})] \\ &= b^2\mathbf{E}Z_rZ_s + bc\mathbf{E}Z_rZ_{s-1} + bc\mathbf{E}Z_{r-1}Z_s + c^2\mathbf{E}Z_{r-1}Z_{s-1}.\end{aligned}$$

There are two cases where  $\gamma_X$  is not zero, and that's because  $\mathbf{E}Z_rZ_s = 1 \iff r = s$ :

$$\begin{aligned}\gamma_X(t, t) &= b^2\mathbf{E}Z_tZ_t + 2bc\mathbf{E}Z_tZ_{t-1} + c^2\mathbf{E}Z_{t-1}Z_{t-1} \\ &= b^2\mathbf{E}Z_t^2 + c^2\mathbf{E}Z_{t-1}^2 \\ &= b^2 + c^2,\end{aligned}$$

and then, by symmetry of  $\gamma$ ,

$$\begin{aligned}\gamma_X(t, t+1) &= \gamma_X(t, t-1) = b^2\mathbf{E}Z_tZ_{t-1} + bc\mathbf{E}Z_tZ_{t-2} + bc\mathbf{E}Z_{t-1}Z_{t-1} + c^2\mathbf{E}Z_{t-1}Z_{t-2} \\ &= bc\mathbf{E}Z_{t-1}^2 \\ &= bc.\end{aligned}$$

On the other hand, for  $|h| > 1$ ,

$$t \neq t+h, \quad t \neq t+h-1, \quad t-1 \neq t+h, \quad t-1 \neq t+h-1$$

$$\begin{aligned}\implies \gamma_X(t, t+h) &= b^2\mathbf{E}Z_tZ_{t+h} + bc\mathbf{E}Z_tZ_{t+h-1} + bc\mathbf{E}Z_{t-1}Z_{t+h} + c^2\mathbf{E}Z_{t-1}Z_{t+h-1} \\ &= 0.\end{aligned}$$

Finally, note that  $\gamma_X$  is only dependent on the difference  $r - s$ , and thus,  $X_t$  is a stationary process with autocovariance function

$$\gamma(h) = \begin{cases} b^2 + c^2 & h = 0, \\ bc & h = 1, \\ 0 & \text{otherwise.} \end{cases}$$

### Solution part (b)

Again, using the linearity of expectation and variance,

$$\mathbf{E}X_t = \cos(ct)\mathbf{E}Z_1 + \sin(ct)\mathbf{E}Z_2 = 0,$$

$$\begin{aligned} \mathbf{E}|X_t|^2 &= \mathbf{Var}(X_t) = \cos^2(ct)\mathbf{Var}Z_1 + \sin^2(ct)\mathbf{Var}Z_2 \\ &= \cos^2(ct) + \sin^2(ct) \\ &= 1. \end{aligned}$$

For the autocovariance function,

$$\begin{aligned} \gamma_X(r, s) &= \mathbf{E}[(\cos(cr)Z_1 + \sin(cr)Z_2)(\cos(cs)Z_1 + \sin(cs)Z_2)] \\ &= \cos(cr)\cos(cs)\mathbf{E}Z_1^2 + \cos(cr)\sin(cs)\mathbf{E}Z_1Z_2 \\ &\quad + \sin(cr)\cos(cs)\mathbf{E}Z_2Z_1 + \sin(cr)\sin(cs)\mathbf{E}Z_2^2 \\ &= \cos(cr)\cos(cs) + \sin(cr)\sin(cs) \\ &= \cos(c(r - s)), \end{aligned}$$

which is only dependent of the value  $r - s$ , and thus,  $\{X_t\}$  is stationary. The autocovariance function can then be defined as

$$\gamma(h) = \cos(c(h))$$

### Solution part (c)

$$\mathbf{E}X_t = \cos(ct)\mathbf{E}Z_t + \sin(ct)\mathbf{E}Z_t = 0,$$

$$\begin{aligned} \mathbf{E}|X_t|^2 &= \mathbf{Var}(X_t) = \cos^2(ct)\mathbf{Var}Z_t + \sin^2(ct)\mathbf{Var}Z_{t-1} \\ &= \cos^2(ct) + \sin^2(ct) \\ &= 1. \end{aligned}$$

Now, we can prove that  $\{X_t\}$  is not stationary by taking the case when  $r - s = 1$ ,

$$\begin{aligned}
\gamma_X(t, t-1) &= \mathbf{E}[(\cos(ct)Z_t + \sin(ct)Z_{t-1})(\cos(c(t-1))Z_{t-1} + \sin(c(t-1))Z_{t-2})] \\
&= \cos(ct)\cos(c(t-1))\mathbf{E}Z_t\mathbf{E}Z_{t-1} + \cos(ct)\sin(c(t-1))\mathbf{E}Z_tZ_{t-2} \\
&\quad + \sin(ct)\cos(c(t-1))\mathbf{E}Z_{t-1}Z_{t-1} + \sin(ct)\sin(c(t-1))\mathbf{E}Z_{t-1}Z_{t-2} \\
&= \sin(ct)\cos(c(t-1))
\end{aligned}$$

This case depends on the value of  $t$  (unless  $c$  is a multiple of  $\pi$ ). For example, if  $c = \pi/2$ , then

$$\begin{aligned}
\gamma_X(1, 0) &= \sin(\pi/2)\cos(0) = 1, \\
\gamma_X(2, 1) &= \sin(\pi)\cos(\pi/2) = 0.
\end{aligned}$$

Therefore,  $\{X_t\}$  is not stationary.

## Exercise 1.

Let  $f : \mathbb{R}^n \mapsto \mathbb{R}$  be a function such that  $|f(x)| \leq \|x\|^2$  for every  $x \in \mathbb{R}^n$ . Show that  $f$  is Fréchet differentiable in 0.

**Solution:** By hypothesis,  $f(\vec{0}) = 0$

$$\begin{aligned}
\lim_{\|x\| \rightarrow 0} \frac{f(x) - f(\vec{0})}{\|x\|} &\leq \lim_{\|x\| \rightarrow 0} \frac{\|x\|^2}{\|x\|} \\
&= 0.
\end{aligned}$$

Thus, the linear function  $\ell(x) = 0$  is the Fréchet derivative of  $f$  at 0.

## Exercise 2.

Let  $f : \mathbb{R}^2 \mapsto \mathbb{R}$  given by

$$f(x, y) = \frac{xy^2}{x^2 + y^4}$$

if  $(x, y) \neq (0, 0)$ , otherwise 0.

- Show that  $f$  has directional derivatives at the origin for every direction  $d$ .
- Show that  $f$  is not Gâteaux differentiable at the origin.



c) Show that  $f$  is not continuous at the origin.

### Solution Part (a)

The directional derivate of  $f$  at a direction  $d = (x, y)$  at the origin is defined as

$$\begin{aligned} f'(\vec{0}; d) &= \lim_{h \rightarrow 0^+} \frac{f(hd) - f(\vec{0})}{h} \\ &= \lim_{h \rightarrow 0^+} \frac{h^3 xy^2}{h^3 x^2 + h^5 y^4} \\ &= \lim_{h \rightarrow 0^+} \frac{xy^2}{x^2 + h^2 y^4} \\ &\stackrel{(\text{L'Hospital})}{=} \frac{y^2}{x} \end{aligned}$$

### Solution Part (b)

$$\begin{aligned} f'(\vec{0}; e_1) &= \lim_{h \rightarrow 0^+} \frac{f(h e_1) - f(\vec{0})}{h} \\ &= \lim_{h \rightarrow 0^+} \frac{h \cdot 0^2}{h^3 + 0^4} \\ &= 0, \end{aligned}$$

$$\begin{aligned} f'(\vec{0}; e_2) &= \lim_{h \rightarrow 0^+} \frac{f(h e_2) - f(\vec{0})}{h} \\ &= \lim_{h \rightarrow 0^+} \frac{0 \cdot h^2}{0^2 + h^5} \\ &= 0. \end{aligned}$$

If we take the direction  $d = (1, 1)$ .

$$f'(\vec{0}; d) = 1 \neq 0 = f'(\vec{0}; e_1) + f'(\vec{0}; e_2).$$

So  $f'$  is not a linear function, and thus,  $f$  is not Gateaux differentiable.

### Solution Part (c)

In order to prove that the function is not continuous, take the curve

$$x(t) = t^2, \quad y(t) = t, \quad t \in (0, 1).$$

Then, if the function is continuous, then the following limit should be unique,

$$\begin{aligned} 0 &= \lim_{\|(x,y)\| \rightarrow 0} f(x,y) = \lim_{t \rightarrow 0^+} f(x(t), y(t)) \\ &= \lim_{t \rightarrow 0^+} \frac{t^2 \cdot t^2}{t^4 + t^4} \\ &= \lim_{t \rightarrow 0^+} \frac{t^4}{2t^4} \\ &= \frac{1}{2}. \end{aligned}$$

But that would lead to a contradiction.

### Exercise 3.

Let  $f : \mathbb{R}^2 \mapsto \mathbb{R}^2$  given by

$$f(x, y) = \frac{2y \exp(-x^{-2})}{y^2 + \exp(-2x^{-2})}$$

if  $(x, y) \neq (0, 0)$  and 0 otherwise. Show that  $f$  is Gâteaux differentiable at the origin but it's not continuous.

**Solution:**

Let  $d = (x, y)$  be the direction where we are differentiating. We want to show that  $f'(\vec{0}, d)$  exists for any  $d = xe_1 + ye_2 \in \mathbb{R}^2$  and it's linear

$$\begin{aligned} f'(\vec{0}; d) &= \lim_{h \rightarrow 0^+} \frac{f(\vec{0} + hd) - f(\vec{0})}{h} \\ &= \lim_{h \rightarrow 0^+} \frac{2hy \exp(-(hx)^{-2})}{h(h^2y^2 + \exp(-2(hx)^{-2}))} \\ &= \lim_{h \rightarrow 0^+} \frac{\exp(-(hx)^{-2})}{h^2y + \exp(-(hx)^{-2})} \end{aligned}$$

$$(\text{L'Hospital}) = 0.$$

However, note that the function is not continuous. For instance, take the curve

$$x(t) = \sqrt{\frac{-1}{\ln t}}, \quad y(t) = t, \quad t \in (0, 1)$$

Note that if we assume that  $f$  is continuous, then the following limit should be unique

$$\begin{aligned} 0 &= \lim_{\|(x,y)\| \rightarrow 0} f(x,y) = \lim_{t \rightarrow 0^+} f(x(t), y(t)) \\ &= \lim_{t \rightarrow 0^+} \frac{2te^{\ln t}}{t^2 + e^{2\ln t}} \\ &= \lim_{t \rightarrow 0^+} \frac{2t^2}{2t^2} \\ &= 1. \end{aligned}$$

But this would lead to a contradiction.

#### Exercise 4.

Let  $f : \mathbb{R}^2 \mapsto \mathbb{R}^2$  given by

$$f(x, y) = \frac{x^3 y}{x^4 + y^2}$$

if  $(x, y) \neq (0, 0)$  and 0 otherwise. Show that  $f$  is Gâteaux differentiable at the origin but it's not Fréchet differentiable.

#### Solution:

The directional derivative is:

$$\begin{aligned} f'(\vec{0}; d) &= \lim_{h \rightarrow 0^+} \frac{f(hd) - f(\vec{0})}{h} \\ &= \lim_{h \rightarrow 0^+} \frac{h^4}{h^5 + h^3} \\ &= \lim_{h \rightarrow 0^+} \frac{h}{h^2 + 1} = 0. \end{aligned}$$

And the partial derivatives evaluated at the origin,

$$\begin{aligned}
f'(\vec{0}; e_1) &= \lim_{h \rightarrow 0^+} \frac{f(h e_1) - f(\vec{0})}{h} \\
&= \lim_{h \rightarrow 0^+} \frac{h^3 \cdot 0}{h^5 + 0^2} \\
&= 0,
\end{aligned}$$

$$\begin{aligned}
f'(\vec{0}; e_2) &= \lim_{h \rightarrow 0^+} \frac{f(h e_2) - f(\vec{0})}{h} \\
&= \lim_{h \rightarrow 0^+} \frac{0 \cdot h}{0^4 + h^3} \\
&= 0.
\end{aligned}$$

Thus, it follows that  $f'$  is linear and the Gâteaux derivative exists. On the other hand, in order to prove that  $f$  is not Fréchet differentiable at the origin, assume for the sake of contradiction that it is. Then, we know that since the limit is unique,

$$\lim_{\|v\| \rightarrow 0} \frac{f(v) - f(\vec{0}) - \ell(v)}{\|v\|} = 0,$$

where  $\ell(v) = f'(\vec{0}; v) = 0$ . However, by taking the curve

$$x(t) = t, \quad y(t) = t^2.$$

Then, using the uniform convergence provided by the Fréchet differentiability,

$$\begin{aligned}
0 &= \lim_{\|v\| \rightarrow 0} \frac{f(v) - f(\vec{0}) - \ell(v)}{\|v\|} = \lim_{t \rightarrow 0} \frac{f(x(t), y(t)) - f(\vec{0}) - \ell(x(t), y(t))}{\|(x(t), y(t))\|} \\
&= \lim_{t \rightarrow 0} \frac{\frac{t^3 t^2}{t^4 + t^4}}{\sqrt{t^2 + t^4}} \\
&= \lim_{t \rightarrow 0} \frac{t^5}{2t^5 \sqrt{1 + t^2}} \\
&= \frac{1}{2} \lim_{t \rightarrow 0} \frac{1}{\sqrt{1 + t^2}} \\
&= \frac{1}{2}.
\end{aligned}$$

Therefore, a contradiction.

## Exercise 5.

Calculate the first 2 derivatives of the function  $f(X) = \det(X)$  for a  $n \times n$  symmetric matrix  $X$ .

**Solution:**

We know that for  $f : \text{Sym}_{++}^n \mapsto \mathbb{R}$ ,  $f(X) = \ln(\det(X))$ ,  $Df(X) = X^{-1}$ . Also, the inner product here is defined as

$$\langle A, B \rangle = \text{tr}(A^* B)$$

Using the chain rule,

$$D(g \circ f) = (Dg \circ f) \circ Df$$

Thus, by defining  $h(X) = \det(X) = \exp(\ln(\det(X)))$ , we obtain

$$\begin{aligned} Dh(X) &= (D \exp)(\ln \det(X)) \cdot (D \ln \det)(X) \\ &= \exp(\ln \det(X)) \cdot X^{-1} \\ &= e^{f(X)} \cdot Df(X). \end{aligned}$$

For the second derivative, we know that  $Hf = D(Df)(X)(Y) = -X^{-1}YX^{-1}$ . Therefore, the Taylor approximation is

$$f(Z) = f(X) + \langle X^{-1}, (Z - X) \rangle - \frac{1}{2} \langle X^{-1}(Z - X)X^{-1}, (Z - X) \rangle + o(\|Z - X\|^2).$$

With this definition, we can use the chain rule on  $D(Dh)$ .

$$\begin{aligned} D(Dh)(X)(Y) &= D(e^{f(X)} \cdot Df(X))(Y) \\ &= [Df(X) \cdot e^{f(X)} \cdot Df(X)](Y) + e^{f(X)} \cdot D(Df)(X)(Y) \\ &= e^{f(X)}(X^{-2}Y - X^{-1}YX^{-1}) \end{aligned}$$