

1. (a) In the n th year from now there are x_n customers of P_1 ; by assumption 85% of those remain with P_1 into the following year, contributing $(0.85)x_n$ to the value of x_{n+1} , and 8% of the y_n customers for P_2 switch to P_1 , contributing a further $(0.08)y_n$ to the value of x_{n+1} . The sum of these gives the stated formula for x_{n+1} .

We derive the formula for y_{n+1} the same way. Now the roles of 85% and 8% are played by 92% and 15%, and this leads to the stated formula for y_{n+1} . We can express all of this in matrix form using the Markov matrix

$$M = \begin{bmatrix} .85 & .08 \\ .15 & .92 \end{bmatrix}.$$

- (b) The matrix M^2 tells us about the behavior 2 years into the future from a given year, since $\mathbf{v}_{n+2} = M\mathbf{v}_{n+1} = M(M\mathbf{v}_n) = (MM)\mathbf{v}_n = M^2\mathbf{v}_n$. Thus, 73.45% of P_1 customers at the start of a calendar year are with P_1 and the end of two years' time (they may have moved to P_2 and then moved back in the meantime), and 14.16% of P_2 customers at the start of a calendar year have switched to P_1 by the end of two years' time.
- (c) In the long run, 34.78% of the population is with P_1 and 65.22% of the population is with P_2 . (Customers keep moving around, but the "total market share" for each company stabilizes at these proportions.) \diamond
2. (a) Since there is no movement other than the indicated one from I_1 to I_2 in the spring and from I_2 to I_1 in the fall, the spring transition matrix is $\begin{bmatrix} 0.9 & 0 \\ 0.1 & 1 \end{bmatrix}$, or B , and the fall transition matrix is $\begin{bmatrix} 1 & 0.1 \\ 0 & 0.9 \end{bmatrix}$, or D . Thus, $B\mathbf{p}$ is the population vector at the end of the spring, and D carries the spring population vector to the fall population vector. That is, the population vector at the end of the fall is $D(B\mathbf{p}) = (DB)\mathbf{p}$. Hence, we have $M = DB$. Multiplying these two matrices gives $M = \begin{bmatrix} .91 & .1 \\ .09 & .9 \end{bmatrix}$.
- (b) This part concerns the passage of two full years. Since the transition matrix for one full year is DB , for two full years it is $N = (DB)^2$. Indeed, beginning with a population vector \mathbf{p} at the start of a year, after one full year it has become $(DB)\mathbf{p}$. Thus, after one more full year (so two full years from the start) we arrive at the population vector $(DB)((DB)\mathbf{p}) = (DB)^2\mathbf{p}$ (using the associative law). Hence, $N = (DB)^2$, which is $\begin{bmatrix} .8371 & .181 \\ .1629 & .819 \end{bmatrix}$.
- (c) In the long run, 52.63% of the birds are on island I_1 and 47.37% of the birds are on island I_2 at the end of every year. Birds keep moving around every year, but the proportions on each at the end of the year stabilize at these values. \diamond

3. (a) The given rules tell us:

$$F_{i+1} = (0.9)F_i + (0.05)N_i, \quad T_{i+1} = (0.9)T_i + (0.1)N_i, \quad N_{i+1} = (0.1)F_i + (0.1)T_i + (0.85)N_i.$$

Hence, we use

$$A = \begin{bmatrix} 0.9 & 0 & 0.05 \\ 0 & 0.9 & 0.1 \\ 0.1 & 0.1 & 0.85 \end{bmatrix}.$$

- (b) We have $M = A^{10}$, by the same iteration arguments as given in the course text for many circumstances (bird migration, gambler's ruin, etc.). \diamond

4. (a) We calculate

$$\mathbf{v}_{n+1} = \begin{bmatrix} a_{n+2} \\ a_{n+1} \end{bmatrix} = \begin{bmatrix} a_{n+1} - a_n \\ a_{n+1} \end{bmatrix} = \begin{bmatrix} 1 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} a_{n+1} \\ a_n \end{bmatrix} = A\mathbf{v}_n.$$

Also by direct multiplication, $A^2 = \begin{bmatrix} 1 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix}$, so

$$A^3 = A^2 A = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}.$$

$$\text{Hence, } A^6 = A^3 A^3 = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}.$$

(b) Since $\mathbf{v}_{n+1} = A\mathbf{v}_n$, feeding this into itself more times gives $\mathbf{v}_{n+2} = A\mathbf{v}_{n+1} = A(A\mathbf{v}_n) = A^2\mathbf{v}_n$ and more generally $\mathbf{v}_{n+k} = A^k\mathbf{v}_n$ for any $k \geq 1$. Setting $k = 6$ gives $\mathbf{v}_{n+6} = A^6\mathbf{v}_n = I_2\mathbf{v}_n = \mathbf{v}_n$ for all n . Comparing bottom entries in these vectors gives $a_{n+6} = a_n$ for all n . We can also verify this final equality from the recurrence definition:

$$\begin{aligned} a_{n+6} &= a_{n+5} - a_{n+4} = (a_{n+4} - a_{n+3}) - a_{n+4} = -a_{n+3} = -(a_{n+2} - a_{n+1}) \\ &= -a_{n+2} + a_{n+1} \\ &= -(a_{n+1} - a_n) + a_{n+1} \\ &= -a_{n+1} + a_n + a_{n+1} \\ &= a_n. \end{aligned}$$

◇

5. (a) First consider city C . Next year's urban population consists of 93% of last year's x_U (namely, those not in the 7% of x_U that moved to the suburbs) and 3% of last year's x_S , so the new value of x_U at the end of this year will be $(0.93)x_U + (0.03)x_S$, using last year's x_U and x_S . Likewise, the new value of x_S at the end of this year will be $(0.07)x_U + (0.97)x_S$ since it has a contribution from the 7% of the urban residents who move to the suburbs during this year and the 97% of suburban residents not in the 3% who move to the urban part of the city. Putting it all together,

$$M = \begin{bmatrix} .93 & .03 \\ .07 & .97 \end{bmatrix}.$$

We can apply the same reasoning for C' , with the roles of 7% and 3% being replaced by 5% and 4% respectively, yielding

$$M' = \begin{bmatrix} .95 & .04 \\ .05 & .96 \end{bmatrix}.$$

(b) For M , the meaning is that in the long run, 30% of the population of C is urban and 70% of the population of C is suburban. Likewise, for M' the meaning is that in the long run, 44.4% of its population is urban and 55.6% of its population is suburban. It is indeed surprising that a change of just a 1% or 2% in the proportions moving in either direction has an impact on the long-term population distribution on the order of around 15% each! ◇

6. As a prelude to using the Chain Rule, we first calculate the derivative matrices

$$(Df)(x, y) = \begin{bmatrix} \cos(x) & 0 \\ 2xye^{yx^2} & x^2e^{yx^2} \end{bmatrix}, \quad (Dg)(r, s, t) = \begin{bmatrix} st & rt & rs \end{bmatrix}, \quad (Dh)(v) = \begin{bmatrix} 1 + \cos(v) \\ 1 - \sin(v) \end{bmatrix}.$$

(a) This is the matrix product

$$\begin{bmatrix} 1 + \cos(rst) \\ 1 - \sin(rst) \end{bmatrix} \begin{bmatrix} st & rt & rs \end{bmatrix} = \begin{bmatrix} st + st \cos(rst) & rt + rt \cos(rst) & rs + rs \cos(rst) \\ st - st \sin(rst) & rt - rt \sin(rst) & rs - rs \sin(rst) \end{bmatrix}.$$

(b) Since $h(0) = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$, we are computing the matrix product

$$(Df)(0, 1) (Dh)(0) = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \end{bmatrix}.$$

(c) Explicitly, $g(f(x, y), x) = x \sin(x) e^{yx^2}$, and its x -partial is computed by using the product rule and Chain Rule of single-variable calculus treating y as constant:

$$\begin{aligned} ((x \sin(x)) e^{yx^2})' &= (x \sin(x))' e^{yx^2} + (x \sin(x)) (e^{yx^2})' \\ &= (\sin(x) + x \cos(x)) e^{yx^2} + 2xy (x \sin(x)) e^{yx^2} \\ &= e^{yx^2} (\sin(x) + x \cos(x) + 2x^2 y \sin(x)). \end{aligned}$$

For the second way, the x -partial is the first entry in the 1×3 matrix $D(g \circ F)(x, y)$ where $F(x, y) = (\sin(x), e^{yx^2}, x)$. By the Chain Rule, this derivative matrix is

$$\begin{aligned} (Dg)(F(x, y)) (DF)(x, y) &= \begin{bmatrix} st & rt & rs \end{bmatrix} \Big|_{F(x, y)} \begin{bmatrix} \cos(x) & 0 \\ 2xy e^{yx^2} & x^2 e^{yx^2} \\ 1 & 0 \end{bmatrix} \\ &= \begin{bmatrix} e^{yx^2} x & x \sin(x) & e^{yx^2} \sin(x) \end{bmatrix} \begin{bmatrix} \cos(x) & 0 \\ 2xy e^{yx^2} & x^2 e^{yx^2} \\ 1 & 0 \end{bmatrix} \\ &= \begin{bmatrix} e^{yx^2} x \cos(x) + 2x^2 y \sin(x) e^{yx^2} + e^{yx^2} \sin(x) & x^3 \sin(x) e^{yx^2} \end{bmatrix}. \end{aligned}$$

Hence, looking at the first entry gives the desired x -partial as $e^{yx^2} (x \cos(x) + (2x^2 y + 1) \sin(x))$. This agrees with what we got the first way, but by an entirely different-looking procedure. \diamond

7. (a) We can view $f(x(r, s), y(r, s))$ as just a function of r (by holding s fixed). We calculate $\partial x / \partial r = 2r$ and $\partial y / \partial r = s$. Formula (17.1.6) applies and we get

$$\frac{\partial f}{\partial r} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial r} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial r} = \frac{\partial f}{\partial x} \cdot (2r) + \frac{\partial f}{\partial y} \cdot s.$$

(b) We iteratively apply the above formula

$$\begin{aligned} \frac{\partial f}{\partial v} &= \frac{\partial f}{\partial x} \frac{\partial x}{\partial v} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial v} \\ &= \frac{\partial f}{\partial x} \left(\frac{\partial x}{\partial r} \frac{\partial r}{\partial v} + \frac{\partial x}{\partial s} \frac{\partial s}{\partial v} \right) + \frac{\partial f}{\partial y} \left(\frac{\partial y}{\partial r} \frac{\partial r}{\partial v} + \frac{\partial y}{\partial s} \frac{\partial s}{\partial v} \right) \\ &= \frac{\partial f}{\partial x} \cdot (2r + 4sv) + \frac{\partial f}{\partial y} \cdot (s + 2rv) \end{aligned}$$

(c) We calculate

$$\begin{aligned}\frac{\partial f}{\partial t} &= \frac{\partial f}{\partial x} \frac{\partial x}{\partial t} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial t} \\ &= \frac{\partial f}{\partial x} \left(\frac{\partial x}{\partial r} \frac{\partial r}{\partial t} + \frac{\partial x}{\partial s} \frac{\partial s}{\partial t} \right) + \frac{\partial f}{\partial y} \left(\frac{\partial y}{\partial r} \frac{\partial r}{\partial t} + \frac{\partial y}{\partial s} \frac{\partial s}{\partial t} \right) \\ &= \frac{\partial f}{\partial x} \left(\frac{\partial x}{\partial r} \left(\frac{\partial r}{\partial v} \frac{\partial v}{\partial t} \right) + \frac{\partial x}{\partial s} \left(\frac{\partial s}{\partial v} \frac{\partial v}{\partial t} \right) \right) + \frac{\partial f}{\partial y} \left(\frac{\partial y}{\partial r} \left(\frac{\partial r}{\partial v} \frac{\partial v}{\partial t} \right) + \frac{\partial y}{\partial s} \left(\frac{\partial s}{\partial v} \frac{\partial v}{\partial t} \right) \right),\end{aligned}$$

where in the final equality there are no terms involving $\partial w/\partial t$ since this vanishes. Evaluating all of the partial derivatives apart from the unknown partials of f with respect to x and y now yields

$$\frac{\partial f}{\partial t} = \frac{\partial f}{\partial x} \cdot (4rt + 8svt) + \frac{\partial f}{\partial y} \cdot (2st + 4rvt). \quad \diamond$$

8. (a) The Chain Rule applied to the function $F(x, y(x)) = (F \circ h)(x) = c$ that is constant gives that its derivative $0 = c' = (F \circ h)'(x)$ is the entry in the 1×1 matrix

$$(DF)(x, y(x)) (Dh)(x) = \begin{bmatrix} F_x & F_y \end{bmatrix} \begin{bmatrix} 1 \\ y'(x) \end{bmatrix} = \begin{bmatrix} F_x + F_y y'(x) \end{bmatrix}.$$

This says $0 = F_x + F_y y'(x)$, so $y'(x) = -F_x/F_y$ wherever F_y is non-vanishing.

- (b) Let $F(x, y) = 2x^3y - y^5x$. According to (a) we have

$$y'(x) = -\frac{F_x(x, y)}{F_y(x, y)} = -\frac{6x^2y - y^5}{2x^3 - 5y^4x}.$$

Hence the slope $(1, 1)$ is $-\frac{6-1}{2-5} = \frac{5}{3}$. \diamond

9. (a) Since $f(x, y, z(x, y)) = c$ is constant, its y -partial derivative vanishes. But the Chain Rule computes the y -partial of $f(x, y, z(x, y)) = (f \circ g)(x, y)$ for $g(x, y) = (x, y, z(x, y))$ to be the second entry in the matrix

$$\begin{aligned}D(f \circ g)(x, y) &= (Df)(g(x, y)) (Dg)(x, y) = \begin{bmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} & \frac{\partial f}{\partial z} \end{bmatrix} \Big|_{g(x, y)} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ \partial z/\partial x & \partial z/\partial y \end{bmatrix} \\ &= \begin{bmatrix} \frac{\partial f}{\partial x} + \frac{\partial f}{\partial z} \frac{\partial z}{\partial x} & \frac{\partial f}{\partial y} + \frac{\partial f}{\partial z} \frac{\partial z}{\partial y} \end{bmatrix}.\end{aligned}$$

In other words, we have $0 = \frac{\partial f}{\partial y} + \frac{\partial f}{\partial z} \frac{\partial z}{\partial y}$, so

$$\frac{\partial z}{\partial y} = -\frac{\partial f/\partial y}{\partial f/\partial z}$$

as desired.

The exact same reasoning, but treating y as a function of (x, z) on the level surface $f = c$ and computing its x -partial yields $\partial y/\partial x = -f_x/f_y$. Likewise, treating x as a function of (y, z) on the level surface $f = c$ and computing its z -partial yields $\partial x/\partial z = -f_z/f_x$; doing similarly with the y -partial yields $\partial x/\partial y = -f_y/f_x$, which is the reciprocal of $-f_x/f_y = \partial y/\partial x$ as desired.

(b) Putting it all together,

$$\frac{\partial z}{\partial y} \frac{\partial y}{\partial x} \frac{\partial x}{\partial z} = \left(-\frac{f_y}{f_z}\right) \left(-\frac{f_x}{f_y}\right) \left(-\frac{f_z}{f_x}\right) = (-1)^3 \frac{f_y}{f_z} \cdot \frac{f_x}{f_y} \cdot \frac{f_z}{f_x} = -1$$

(by cancellation of numerators and denominators in genuine fractions!).

(c) Since $P = cT/V$, we have $\partial P/\partial T = c/V$. Since $T = PV/c$, we have $\partial T/\partial V = P/c$. Since $V = cT/P$, we have $\partial V/\partial P = -cT/P^2$. Hence, the product of all three of these is $(c/V)(P/c)(-cT/P^2) = -cT/(VP) = -1$. \diamond

10. Let $H : \mathbf{R}^n \rightarrow \mathbf{R}^{2m}$ be given by $H(\mathbf{x}) = (f(\mathbf{x}), g(\mathbf{x}))$ and $M : \mathbf{R}^{2m} \rightarrow \mathbf{R}$ be given by $M(\mathbf{u}, \mathbf{v}) = \mathbf{u} \cdot \mathbf{v}$. Then we have $h = M \circ H$. To calculate the derivative we calculate

$$(DH)(\mathbf{x}, \mathbf{y}) = \begin{bmatrix} (Df)(\mathbf{x}) \\ (Dg)(\mathbf{x}) \end{bmatrix}$$

and

$$(DM)(\mathbf{u}, \mathbf{v}) = [v_1 \ \cdots \ v_m \ u_1 \ \cdots \ u_m].$$

Thus,

$$\begin{aligned} D(M \circ H)(\mathbf{x}) &= (DM)(f(\mathbf{x}), g(\mathbf{x})) \begin{bmatrix} (Df)(\mathbf{x}) \\ (Dg)(\mathbf{x}) \end{bmatrix} \\ &= [g_1(\mathbf{x}) \ \cdots \ g_m(\mathbf{x}) \ f_1(\mathbf{x}) \ \cdots \ f_m(\mathbf{x})] \begin{bmatrix} (Df)(\mathbf{x}) \\ (Dg)(\mathbf{x}) \end{bmatrix} \\ &= g(\mathbf{x})^\top (Df)(\mathbf{x}) + f(\mathbf{x})^\top (Dg)(\mathbf{x}) \end{aligned}$$

(where the final equality is seen by considering how the matrix product works out entry by entry; think through the case $m = n = 2$ to see what is going on). \diamond

11. (a) By definition, $(D_{\mathbf{v}}f)(\mathbf{a})$ is the entry in the 1×1 matrix

$$D(f \circ g)(0) = (Df)(g(0)) (Dg)(0) = (Df)(\mathbf{a}) (Dg)(0),$$

and the $n \times 1$ matrix (a “column vector”) $(Dg)(0)$ is exactly \mathbf{v} since we calculate its i th entry to be the derivative $g'_i(0)$ of the i th component function $g_i(t) = a_i + tv_i$ (and the t -derivative g'_i is the constant function with value v_i). In other words, $(D_{\mathbf{v}}f)(\mathbf{a}) = (Df)(\mathbf{a})\mathbf{v}$ when viewing the column vector \mathbf{v} as a an $n \times 1$ matrix, and this in turn is exactly the matrix-vector product.

(b) For the specific f we have, $(Df)(x, y) = [\pi y \cos(\pi xy) \ \pi x \cos(\pi xy)]$. For $\mathbf{v} = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$, the hilliness when moving directly northeast amounts to looking at the height function f on the line $\mathbf{a} + t\mathbf{v}$. Hence, the slope is given by the t -derivative of this height at $t = 0$, which is

$$(D_{\mathbf{v}}f)(1, 2) = ((Df)(1, 2)) \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = [2\pi \ \pi] \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = 3\pi/\sqrt{2}. \quad \diamond$$

12. (a) We have $h = f \circ F$ for $F(r, \theta) = (r \cos \theta, r \sin \theta)$, so F has component functions $F_1(r, \theta) = r \cos \theta$ and $F_2(r, \theta) = r \sin \theta$. Hence, by the Chain Rule

$$(Dh)(r, \theta) = (Df)(F(r, \theta))(DF)(r, \theta) = (Df)(x, y) \begin{bmatrix} (F_1)_r & (F_1)_\theta \\ (F_2)_r & (F_2)_\theta \end{bmatrix} = [f_x \ f_y] \begin{bmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{bmatrix}.$$

Multiplying the matrices gives

$$\begin{bmatrix} h_r & h_\theta \end{bmatrix} = [\cos \theta f_x + \sin \theta f_y \quad -r \sin \theta f_x + r \cos \theta f_y].$$

Equating entries on the two sides gives the desired formulas for h_r and h_θ .

Alternatively, via (17.1.6) with $x(r, \theta) = r \cos \theta$ and $y(r, \theta) = r \sin \theta$ we have

$$\frac{\partial f}{\partial r} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial r} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial r} = \cos \theta \frac{\partial f}{\partial x} + \sin \theta \frac{\partial f}{\partial y}$$

and

$$\frac{\partial f}{\partial \theta} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial \theta} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial \theta} = -r \sin \theta \frac{\partial f}{\partial x} + r \cos \theta \frac{\partial f}{\partial y}.$$

In such equations, the left sides are informal notation for what are really $\partial h / \partial r$ and $\partial h / \partial \theta$ respectively, so we again get the result we wanted.

(b) From (a) we have the pair of equations

$$\begin{aligned} \frac{\partial h}{\partial \theta} &= -r \sin \theta \frac{\partial f}{\partial x} + r \cos \theta \frac{\partial f}{\partial y} \\ \frac{\partial h}{\partial r} &= \cos \theta \frac{\partial f}{\partial x} + \sin \theta \frac{\partial f}{\partial y}. \end{aligned}$$

We want to unwind this to express the x and y partials of f in terms of the partials of h .

Let's adapt the method used to solve "two (linear) equations in two unknowns": multiply by suitable expressions to then combine the equations and make f_x disappear (since we want to describe f_y). To make f_x disappear, let's multiply the first equation by $\cos \theta$, the second by $r \sin \theta$, and then add the equations. This gives

$$\cos \theta h_\theta + r \sin \theta h_r = (r \cos^2 \theta + r \sin^2 \theta) f_y = r f_y.$$

Dividing by r (as we may do since we're assuming $r > 0$) gives

$$f_y = \frac{\cos \theta}{r} h_\theta + (\sin \theta) h_r.$$

So $g_1(r, \theta) = (\cos \theta)/r$ and $g_2(r, \theta) = \sin \theta$. ◇