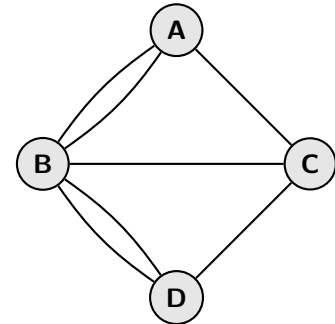


Example 174. We discussed the famous problem of the **Seven Bridges of Königsberg**.

https://en.wikipedia.org/wiki/Seven_Bridges_of_Königsberg

Historically, Euler's clever (and general!!) solution (1736) laid the foundation of graph theory and topology.

A key insight is to focus on the 4 landmasses instead of the 7 bridges. Let's call these landmasses A, B, C, D . They are connected via the bridges as in the graph to the right.



The problem at hand is to traverse the graph via a sequence of consecutive edges such that each edge is traversed exactly once.

Obviously (think about it!), if a (connected) graph is thus traversable, then

- either all nodes have even degree (the **degree** refers to the number of edges at that node), or
- two nodes (start and end of our route) have odd degree and all other nodes have even degree.

Neither of these conditions is true for our graph. Hence the 7 bridges problem has no solution.

More is true! However, more is true and this makes Euler's approach much more interesting. Namely, the two conditions above are true **if and only if** a (connected) graph is traversable.

Can you prove that claim? That is, for a graph satisfying the two conditions, you need to show that it can indeed be traversed. One way to do that is by induction on the number of edges (roughly: start with an edge attached to a node with odd degree (if not present, just take any edge); then, remove that edge and keep it in mind as the first step of your journey...).

10 Linear approximations to arbitrary functions

- **(Calculus 1)** Let $f(x)$ be any (nice; i.e. differentiable) function $\mathbb{R} \rightarrow \mathbb{R}$.

Suppose the best linear approximation to $f(x)$ at $x = a$ is $f(x) \approx f(a) + m(x - a)$.

Geometrically, $x \mapsto f(a) + m(x - a)$ describes the tangent line to $f(x)$ at $x = a$.

In Calculus 1, you learn that the crucial quantity m gives rise to the **derivative**, $m = f'(a)$, and you learn how to compute it, given a function $f(x)$.

- **(Calculus 3)** Let $f(x)$ be any (nice) function $\mathbb{R}^n \rightarrow \mathbb{R}^m$.

Now, the best linear approximation to $f(x)$ at $x = a$ is $f(x) \approx f(a) + M(x - a)$.

Here, M is a $m \times n$ matrix. (Why these dimensions?!)

In Calculus 3, you again learn how to compute the quantity $M = Df(a)$ (the **derivative** or **Jacobian matrix**), given a function $f(x)$. Indeed:

$$f(x) = \begin{bmatrix} f_1(x_1, \dots, x_n) \\ \vdots \\ f_m(x_1, \dots, x_n) \end{bmatrix}, \quad Df = \begin{bmatrix} \frac{\partial}{\partial x_1} f_1 & \cdots & \frac{\partial}{\partial x_n} f_1 \\ \vdots & & \vdots \\ \frac{\partial}{\partial x_1} f_m & \cdots & \frac{\partial}{\partial x_n} f_m \end{bmatrix}$$

Comment. One important point of this discussion is that, through our linear algebra glasses, the transition from Calculus 1 to Calculus 3 is exactly as expected.

Example 175. Determine the best linear approximation to $\mathbf{f}(x, y) = \begin{bmatrix} x^2 + y^2 \\ xy + 1 \end{bmatrix}$ at $\mathbf{x} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$.

Solution. The Jacobian matrix at $\mathbf{x} = \begin{bmatrix} x \\ y \end{bmatrix}$ is

$$D\mathbf{f}(x, y) = \begin{bmatrix} \frac{\partial}{\partial x}(x^2 + y^2) & \frac{\partial}{\partial y}(x^2 + y^2) \\ \frac{\partial}{\partial x}(xy + 1) & \frac{\partial}{\partial y}(xy + 1) \end{bmatrix} = \begin{bmatrix} 2x & 2y \\ y & x \end{bmatrix}, \quad D\mathbf{f}(2, 1) = \begin{bmatrix} 4 & 2 \\ 1 & 2 \end{bmatrix}.$$

Hence, the best linear approximation to $\mathbf{f}(x, y)$ at $\mathbf{x} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ is

$$\mathbf{f}(x, y) \approx \mathbf{f}(2, 1) + \begin{bmatrix} 4 & 2 \\ 1 & 2 \end{bmatrix} \left(\begin{bmatrix} x \\ y \end{bmatrix} - \begin{bmatrix} 2 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} 5 \\ 3 \end{bmatrix} + \begin{bmatrix} 4 & 2 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x - 2 \\ y - 1 \end{bmatrix}.$$

Comment. For most purposes, this is the nicest form to write the linear approximation. If we want to multiply out the matrix-vector product, we get

$$\mathbf{f}(x, y) \approx \begin{bmatrix} 4x + 2y - 5 \\ x + 2y - 1 \end{bmatrix}.$$

Note how we solved two independent problems at once: approximating $x^2 + y^2$ at $(x, y) = (2, 1)$, and approximating $xy + 1$ at $(x, y) = (2, 1)$.

Example 176. (homework) For $\mathbf{f}(r, \theta) = \begin{bmatrix} r \cos \theta \\ r \sin \theta \end{bmatrix}$, determine the Jacobian matrix (i.e. the derivative) and its determinant.

Solution. The Jacobian matrix is $\begin{bmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{bmatrix}$.

Its determinant is $\det \left(\begin{bmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{bmatrix} \right) = r \cos^2 \theta + r \sin^2 \theta = r$.

Example 177. If you have taken Calculus 3, you have learned about **substitution** in multiple integrals. To make the change of variables $\mathbf{x} = \mathbf{g}(\mathbf{u})$, that is, $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} g_1(u_1, u_2) \\ g_2(u_1, u_2) \end{bmatrix}$, we have

$$\iint_R f(\mathbf{x}) dx_2 dx_1 = \iint_G f(\mathbf{g}(\mathbf{u})) |D\mathbf{g}(\mathbf{u})| du_2 du_1$$

- where G is the region in the $u_1 u_2$ -plane corresponding to the region R in the $x_1 x_2$ -plane,

- and $|D\mathbf{g}(\mathbf{u})| = \det \begin{bmatrix} \frac{\partial}{\partial u_1} g_1 & \frac{\partial}{\partial u_2} g_1 \\ \frac{\partial}{\partial u_1} g_2 & \frac{\partial}{\partial u_2} g_2 \end{bmatrix}$ is the **Jacobian determinant**.

Comment. Have another look at Example 84. In that example, we observed that $\det(A)$ measures by much a little volume (or area) is scaled under $\mathbf{x} \mapsto A\mathbf{x}$. Hence, it makes perfect sense that the formula for substitution needs to take into account how much the function \mathbf{g} changes volumes. Locally, $\mathbf{u} \mapsto \mathbf{g}(\mathbf{u})$ is approximated (up to a shift) by multiplication with $D\mathbf{g}$, and so the change in volume is measured by the Jacobian determinant.

For instance. From your computation in the previous example, it follows that, using **polar coordinates** $x = r \cos \theta$, $y = r \sin \theta$, we have the familiar formula

$$\iint_R f(x, y) dy dx = \iint_G f(r \cos \theta, r \sin \theta) r dr d\theta.$$