Dynamic similarity, the dimensionless science

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Dimensional analysis, a framework for drawing physical parallels between systems of disparate scale, affords key insights into natural phenomena too expansive and too energetic to replicate in the lab.

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Many experiments seem daunting at first glance, owing to the sheer number of physical variables they involve. To design an apparatus that circulates fluid, for instance, one must know how the flow is affected by pressure, by the apparatus’s dimensions, and by the fluid’s density and viscosity. Complicating matters, those parameters may be temperature and pressure dependent. Understanding the role of each parameter in such a high-dimensional space can be elusive or prohibitively time consuming.

Dimensional analysis, a concept historically rooted in the field of fluid mechanics, can help to simplify such problems by reducing the number of system parameters. For example, in a fluid apparatus in which the flow is isothermal and incompressible, the number of relevant parameters can often be reduced to one. The rewards of such a reduction can be spectacular: It may allow a model the size of a children’s toy to yield insight into the dynamics of a jet airplane, or a fluid-filled cylinder the size of a garbage can to elucidate the behavior of a stellar interior. (See box 1 for a brief history of dimensional analysis.)

**Dimensional reasoning**

Dimensional analysis comes in many forms. One of its simplest uses is to check the plausibility of theoretical results. For example, the displacement \( x(t) \) of a falling body having initial displacement \( x_0 \) and initial velocity \( u_0 \) is

\[
x = x_0 + u_0 t + \frac{1}{2} g t^2,
\]

where \( g \) is its acceleration due to gravity. According to the principle of dimensional homogeneity, if the left- and right-hand sides of the equation are truly equal, they must share the same dimensions. Indeed, each term in the equation has dimensions of length. Despite the modesty of the dimensional-homogeneity requirement, it is violated by a number of equations often used in the hydraulics literature, such as the Manning formula for flow in an open channel and the Hazen–Williams formula, which describes flows of water through pipes.

Dimensional analysis can also help to supply a theoretical result. Consider the ray of light illustrated in figure 1, which, in accordance with general relativity, is deflected as it passes through the gravitational field of the Sun. Assuming the Sun can be treated as a point of mass \( m \) and that the ray of light passes the mass with a distance of closest approach \( r \), dimensional reasoning can help predict the deflection angle \( \theta \).

Expressed in terms of mass \( M \), length \( L \), and time \( T \), the variables’ dimensions—denoted with square brackets—are

\[
[M] = [L]^1 [T]^{-2}
\]

\[
[L] = [L]^1
\]

\[
[T] = [T]^1
\]

Box 1. A brief history of dimensional analysis

Going back more than 300 years, discussions of dimensional analysis have appeared in scores of texts, often with different slants:

**1687.** Isaac Newton publishes the *Principia*, which, in book II, section 7, contains perhaps the earliest documented discussion of dimensional analysis.

**1765.** Leonhard Euler writes extensively about units and dimensional reasoning in *Theoria motus corporum solidorum seu rigidorum*, a comprehensive treatment of the mechanics of rigid bodies.

**1822.** Joseph Fourier employs concepts of dimensional analysis in his *Analytical Theory of Heat*.

**1877.** Lord Rayleigh outlines a “method of dimensions” in his *Theory of Sound*.

**1908.** At the 4th International Congress of Mathematicians in Rome, Arnold Sommerfeld introduces a dimensionless number that he calls the Reynolds number, in tribute to Osborne Reynolds. The Reynolds number, which appeared in what’s now known as the Orr–Sommerfeld equation, is among the most famous of all dimensionless numbers.

**1914.** In what is generally regarded as the big breakthrough in dimensional analysis, physicist Edgar Buckingham introduces the theorem now known as the Buckingham Pi theorem. It is one of several methods of reducing a number of dimensional variables to a smaller number of dimensionless groups.

**1922.** In his influential book *Dimensional Analysis*, Percy Bridgman outlines a general theory of the subject.

**1953.** In his George Darwin lecture before the Royal Astronomical Society, Subrahmanyan Chandrasekhar names the Rayleigh number, a dimensionless temperature difference central to thermal convection.
[\theta] = L\theta T^2 M^0, [r] = L T^0 M^0, and [m] = L T^0 M^1. It follows that no combination of powers of \(r\) and \(m\) can be dimensionally homogeneous with \(\theta\).

Adding the gravitational constant \(G\), which has dimensions \(L T^{-2} M^{-1}\), and the speed of light \(c\), which has dimensions \(L T^{-1} M^0\), seems sensible for a problem concerning gravity and light. Of the potential expressions containing \(m\), \(r\), \(c\), and \(G\), algebraic calculations reveal that dimensional homogeneity is achieved only with solutions of the form \(m^k r^\kappa c^{2\kappa} G^\mu\), where \(\kappa\) is any real integer. (See reference 1 for a step-by-step treatment.)

If the light ray skims just across the Sun’s surface, then \(r = 6.96 \times 10^8\) m, \(m = 1.99 \times 10^{30}\) kg, and the quantity \(m^k r^\kappa c^{2\kappa} G^\mu\) will be small—on the order of \(10^8\) when \(\kappa = 1\). The \(k = 1\) term will give the largest effect, and higher-order terms can be neglected. Arguments based purely on dimensional reasoning suggest, then, that

\[
\theta = a \left( \frac{Gm}{c^2 r} \right),
\]

where \(a\) is an unknown constant. When Isaac Newton considered the problem more than 300 years ago, he arrived at an identical expression, with \(a = 2\). General relativity predicts \(a = 4\), and the latest experiments agree with that result to within 0.02%.

**Prototypes, models, and similitude**

In fluid dynamics, dimensional analysis is used to reduce a large number of parameters to a small number of dimensionless groups, often in spectacular fashion. In addition to easing analysis, that reduction of variables gives rise to new classes of similarity.

Consider the simple example of flow around a prototype airfoil, \(p\), and a much smaller model, \(m\), as illustrated in figure 2. The model and prototype are geometrically similar if all of their corresponding length scales, including surface roughness, are proportionate. Likewise, flows in the two systems are kinematically similar if the velocity ratios \(u_p/u_m\) are the same for all pairs of corresponding, or homologous, points.

Depending on what is to be learned from the model, kinematic similarity may be too lax a requirement. The stricter standard of dynamic similarity exists if the ratios of all forces acting on homologous fluid particles and boundary surfaces in the two systems are constant. Dynamically similar systems are by definition both geometrically and kinematically similar.

An important conclusion of fluid mechanics is that incompressible, isothermal flows in or around geometrically similar bodies are considered dynamically similar if they have the same Reynolds number \(Re\), where \(Re\) is the ratio of inertial to viscous forces (see box 2). Consider the example of a typical attack submarine, 110 m long and capable of moving at 20 knots, or about 10 m/s. In water, that corresponds to \(Re = 1.13 \times 10^9\). If all design tests must be conducted on a 6-m-long scale model that can be towed at a top speed of 10 knots, the highest achievable \(Re\) would be about \(3 \times 10^7\), about 1/36 that of an actual submarine. The model would be a poor descriptor of large-\(Re\) effects such as turbulence.

If the same model were placed in a very large wind tunnel blowing air at, say, half the speed of sound, \(Re\) would be about \(7 \times 10^7\)—closer to, but still well short of, true submarine conditions. Further options are to cool the air, thus lowering its viscosity and increasing its density—and thereby reducing the kinematic viscosity—or to operate at high pressures, increasing density more still. Adopting just that strategy, engineers at the National Transonic Facility at NASA’s Langley Research Center in Virginia reached \(Re\) of about \(1 \times 10^9\) in a cryogenic wind tunnel.

Cryogenic tunnels are currently among the most advanced test facilities available. Remarkably, one of the smallest such tunnels, just 1.4 cm in diameter, is capable of reaching \(Re\) as high as 1.5 million.\(^2\)

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**Figure 1.** A light beam is deflected as it passes through the gravitational field of a star. Here, \(m\) is the star’s mass and \(r\) is the distance of closest approach. Even without knowing the underlying physics, one can use dimensional reasoning to predict that the deflection angle \(\theta\) scales as \(Gmr^{-1}c^{-2}\), where \(G\) is the gravitational constant and \(c\) is the speed of light.

**Figure 2.** A prototype airfoil (left) can be tested with a much smaller model (right), provided the objects are geometrically similar and that the flows around them are dynamically similar. Dynamic similarity is achieved if the characteristic flow velocities \(U_p\) and \(U_m\) are such that the forces at all homologous points—such as the two marked by asterisks—are proportionate. (Adapted from ref. 15.)
The friction factor

One consequence of dynamic similarity in pipe flows is that the so-called friction factor \( \lambda \)—the dimensionless shear stress exerted by the fluid on the pipe—and vice versa—is a function of \( Re \) only, provided entrance effects, surface roughness, and temperature variations are small. The friction factor is especially significant in engineering. The standard \( \lambda(Re) \) plot, compiled from the results of eight papers published between 1914 and 1933, is reproduced in nearly all fluid dynamics texts and spans \( Re \) from \( 1 \times 10^3 \) to \( 3 \times 10^6 \). Two devices exploiting two different strategies have charted new territory.

In 1998 the Princeton University team of Mark Zagarola and Alexander Smits published the first results from their “superpipe,” a closed-loop, 34-m-long pipe with a nominal diameter of 12 cm. Using room-temperature air compressed as high as 187 atmospheres, the pair measured \( \lambda(Re) \) for \( Re \) up to \( 3.6 \times 10^6 \).

Later, a University of Oregon group led by one of us (Donnelly) designed a device consisting of a 28-cm-long pipe, roughly a half-centimeter in diameter, housed in a tabletop helium cryostat. Several room-temperature gases—helium, oxygen, nitrogen, carbon dioxide, and sulfur hexafluoride—were used to measure \( \lambda(Re) \) for relatively small \( Re \); liquid He was used to attain the highest \( Re \), up to \( 1.1 \times 10^6 \).

Figure 3 shows datasets from both experiments. Combined, the data span \( Re \) ranging from 11 to 37 million. Despite a dramatic difference in scale—the Princeton superpipe weighs about 25 tons, the Oregon tube about an ounce—the overlapping data sets agree to within about 2%. It is a testament to the power of dynamic similarity.

Rayleigh–Bénard convection

Thermally driven convection is a conceptually simple but experimentally challenging problem. In the lab it’s typically carried out in a Bénard cell like that sketched in figure 4a, a container of fluid heated from below and cooled from above. The temperature difference \( \Delta T \) gives rise to a density gradient; for typical fluids having a positive thermal expansion coefficient, the denser fluid will lie above the less dense fluid. If the den-
sity gradients are large enough, the configuration destabilizes, leading to circulating flow known as Rayleigh–Bénard convection. That convection enhances the heat transfer from the hot lower boundary to the cool upper one.

As detailed in box 2, the character of a Rayleigh–Bénard flow can be described wholly in terms of dimensionless parameters. The Nusselt number $Nu$, the dimensionless heat-transfer rate, depends on the Rayleigh number $Ra$, the dimensionless temperature difference, and the Prandtl number $Pr$, the ratio of the diffusivity of vorticity to thermal diffusivity. (We’ve assumed that the system’s geometry is fixed.) For small $Ra$, the fluid layer remains at rest, heat transfer is entirely convective, and $Nu$ is relatively small. But as $Ra$ grows, the fluid begins to convect and $Nu$ increases. A series of complicated flow transitions ensues, until eventually—roughly, around $Ra = 10^6$—the flow becomes turbulent. In that turbulent regime, dimensional arguments suggest that $Nu \propto Ra^{1/2}$, where heuristic arguments suggest that $\gamma$ should vary from around 2/7, or 1/3, to an asymptotic value of 1/2. (Other expressions for $Nu(Ra)$ have been hypothesized; see, for example, the article by Leo Kadanoff, Physics Today, August 2001, page 34.)

The experimental challenge is to explore $Nu$ in the highly turbulent regimes of large $Ra$. Dynamic similarity affords multiple ways to do so. One way to boost $Ra$—defined as $galΔT/κν$, where $α$, $κ$, and $ν$ are the thermal expansion coefficient, the vorticity diffusivity, and the thermal diffusivity, respectively—is to create large temperature gradients. A drawback of that approach, however, is that it can yield large density variations, which complicate theoretical modeling. When $ΔT$ is large, a central assumption of the Boussinesq equations that describe thermal convection—namely, that density depends linearly on temperature—no longer holds true (see box 2).

Another strategy for obtaining large $Ra$ is to use a thick fluid layer. Because $Ra$ scales as $L^3$, modest increases in $L$ can produce substantial gains in $Ra$. Alternatively, one can choose a fluid having large $α/κν$. By those measures, low-temperature He is, to our knowledge, the most ideal fluid available. Extreme increases in $α/κν$, however, can lead to undesirably large changes in $Pr$.

Several experiments adopt some combination of the above strategies to explore heat transfer at the higher reaches of $Ra$. A University of Oregon team led by Donnelly used low-temperature He to explore $Ra$ spanning 11 orders of magnitude. Strikingly, the data, shown in red in figure 4b, are described cleanly by a single power law, with $γ = 0.31$.

In 2001, about the same time as the Oregon data appeared, researchers at Joseph Fourier University in Grenoble, France, used a nearly identical cryogenic cell to obtain the results shown in green in figure 4b. As $Ra$ nears $2 × 10^{11}$, the Grenoble data switch from a power law described by $γ = 0.31$ to one described by $γ = 0.39$. One potentially exciting interpretation is that the switch marks the transition to the “ultimate state,” a state predicted by Robert Kraichnan for asymptotically large $Ra$, in which the viscous boundary layers at the ends of the Bénard cell become turbulent.

The discrepancy between the Oregon and Grenoble data...
as plotted in figure 4b seems quite small. Is it of any real consequence? In geophysical and astrophysical fluid dynamics, the answer is yes. Natural phenomena such as mantle convection in Earth’s outer core, atmospheric and oceanic winds, and flows in gas giants and stars are estimated to have $Ra$ ranging from $10^{20}$ to $10^{30}$, perhaps larger in stellar systems. Extrapolated to such geophysical and astrophysical proportions, a slight difference in scaling relationships could yield order-of-magnitude differences in $Nu$.

The ideas of dynamic similarity can help resolve the discrepancy. Guenter Ahlers of the University of California, Santa Barbara and colleagues at the University of Göttingen in Germany explored the range of $Ra$ between $10^6$ and $10^{13}$ using He, $N_2$, and SF$_6$ at ambient temperatures and pressures up to 15 atmospheres. The high pressure, a feature absent from the Oregon and Grenoble experiments, limits changes in $Pr$.

Ahlers and company’s data, shown in blue in figure 4b, agree closely with the Oregon results and contradict the ultimate-state interpretation of the Grenoble data. But as the figure shows, there is still no consensus for large-$Ra$ behavior, and the story continues to unfold. It remains unclear whether Kraichnan’s ultimate state exists, and if so, where it begins.

Rayleigh–Bénard convection, with rotation

Most convection systems of geophysical and astrophysical interest also involve rotation. The influence of rotation is studied in the lab by spinning a Bénard cell about its axis with some angular velocity $\Omega$. Again, the flow behavior can be described in dimensionless terms. Except now, in addition to $Ra$ and $Pr$, two new dimensionless numbers are also important.

The first is the convective Rossby number $Ro = (g \alpha \Delta T / 4 \Omega L)^{1/2}$, the ratio of temperature-induced buoyant forces to rotation-induced Coriolis forces. One might anticipate that the transition between rotationally dominated and nonrotationally dominated flow should occur somewhere near $Ro = 1$.

But there is also the Ekman number $E = \nu (2 \Omega L)^{-1}$, the ratio of viscous to Coriolis forces. Coriolis forces tend to sweep away the viscous boundary layer that exists near the container walls, and so the thickness $\delta_C$ of that boundary layer scales as $E^{-1/2}$. A competing length scale is the thickness $\delta_e$ of the thermal boundary layer, which scales as $Nu^{-1} L$. In general, communication between the container and the bulk fluid will be limited by the thinner of the two boundary layers. One might anticipate that the transition between rotationally dominated and nonrotationally dominated convection should occur when $\delta_e = \delta_C$.

As shown in figure 4c, data from experiments at UCLA’s simulated planetary interiors laboratory confirm that the condition $\delta_e = \delta_C$, not $Ro = 1$, governs the transition from rotationally dominated to nonrotationally dominated convection.$^{10}$ When $\delta_e < \delta_C$, rotation acts to prevent convection, and heat transfer is less efficient than in a nonrotating system. When $\delta_e > \delta_C$, rotation effects are negligible, and $Nu$ scales as it does in the nonrotating case. With that crucial observation, the UCLA researchers were able to estimate the temperature gradients in Earth’s liquid-metal outer core as corresponding to $Ra = 7 \times 10^{24}$. Of course, the extrapolation of carefully controlled laboratory experiments to geophysical fluid mechanics carries caveats, several of which are detailed in reference 10.

A real-world pendulum

Among the first problems posed to undergraduate physics students is that of a simple pendulum: a point mass suspended in a vacuum, oscillating with small amplitude. A real pendulum oscillating in a viscous fluid, however, presents a greater challenge. Wilfried Schoepe and colleagues at the University of Regensburg in Germany studied the problem$^{11}$ using a 100-μm sphere immersed in liquid He. Their data, shown in figure 5, indicate a deviation from laminar flow at a critical $Re$ near 700.

At the University of Oregon we duplicated the experiment with a 1-inch steel bob oscillating in water. The bob was 256 times as large as and 37 million times heavier than the Regensburg group’s sphere. Yet our experiment yielded a nearly identical relationship between dimensionless drag and $Re$ and showed a similar deviation from laminar flow at large $Re$. Photos revealed that the steel bob starts to shed vortex rings when $Re$ surpasses the critical value.$^{12}$

Beyond fluids

A quick check with an internet search engine reveals the ubiquity of dynamic similarity. Steven Vogel of Duke University has helped pioneer the use of dimensional analysis in biophysics.$^{13}$ He has used the concepts to highlight bounds on certain forms of physical behavior, such as the maximum
height of a tree if getting sap to the leaves is the crucial factor (see PHYSICS TODAY, November 1998, page 22).

Principles of similarity also underlie key economic models, such as the debt-to-income ratio. For a long time now, economists have had a good understanding of what that ratio should be, regardless of total annual income, if the debt is to be manageable. A recent article, “Dimensions and Economics: Some Problems,” suggests that many commonly used models are not dimensionally homogeneous, which could result in problems during application and analysis.

As with all tools, it is important to be aware of the potential limitations of dynamic similarity. The principle of similarity could be crudely construed as follows: Two systems can be considered completely similar when all dimensionless numbers are the same. In practice, complete similarity is impossible to achieve unless the two systems are exactly the same.

For example, in the submarine problem, we ignored flow-compressibility effects, which become pronounced when flow speeds approach the speed of sound. Sound travels much more slowly in air than in water, so one must be cautious. It is typically assumed that as long as the flow speed is less than half that of sound, compressibility can be neglected. Similarly, Rayleigh–Bénard experiments hint that Pr, often assumed to be negligible, may play a more important role in heat transfer than once thought.

History demonstrates, however, that it is certainly possible to use principles of similarity to draw valuable parallels between systems that aren’t entirely similar. Currently, dynamic similarity and dimensional analysis are topics that engineering students learn as part of their fluid mechanics course, typically in the second or third undergraduate year. We believe that emphasis on dimensional reasoning would be useful to students in many branches of physics as well.

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References