I would like this book to do more than help you pass calculus. (I hope it does that too.) After calculus you will have choices—Which mathematics course to take next?—and these pages aim to serve as a guide. Part of the answer depends on where you are going—toward engineering or management or teaching or science or another career where mathematics plays a part. The rest of the answer depends on where the courses are going. This chapter can be a useful reference, to give a clearer idea than course titles can do:

Linear Algebra  Differential Equations  Discrete Mathematics
Advanced Calculus (with Fourier Series)  Numerical Methods  Statistics

Pure mathematics is often divided into analysis and algebra and geometry. Those parts come together in the “mathematical way of thinking”—a mixture of logic and ideas. It is a deep and creative subject—here we make a start.

Two main courses after calculus are linear algebra and differential equations. I hope you can take both. To help you later, Sections 16.1 and 16.2 organize them by examples. First a few words to compare and contrast those two subjects.

Linear algebra is about systems of equations. There are $n$ variables to solve for. A change in one affects the others. They can be prices or velocities or currents or concentrations—outputs from any model with interconnected parts.

Linear algebra makes only one assumption—the model must be linear. A change in one variable produces proportional changes in all variables. Practically every subject begins that way. (When it becomes nonlinear, we solve by a sequence of linear equations. Linear programming is nonlinear because we require $x \geq 0$.) Elsewhere I wrote that “Linear algebra has become as basic and as applicable as calculus, and fortunately it is easier.” I recommend taking it.

A differential equation is continuous (from calculus), where a matrix equation is discrete (from algebra). The rate $dy/dt$ is determined by the present state $y$—which changes by following that rule. Section 16.2 solves $y' = cy + s(t)$ for economics and life sciences, and $y'' + by' + cy = f(t)$ for physics and engineering. Please keep it and refer to it.
A third key direction is **discrete mathematics**. Matrices are a part, networks and algorithms are a bigger part. Derivatives are not a part—this is closer to algebra. It is needed in computer science. Some people have a knack for counting the ways a computer can send ten messages in parallel—and for finding the fastest way.

**Typical question:** Can 25 states be matched with 25 neighbors, so one state in each pair has an even number of letters? New York can pair with New Jersey, Texas with Oklahoma, California with Arizona. We need rules for Hawaii and Alaska. This matching question doesn't sound mathematical, but it is.

Section 16.3 selects four topics from discrete mathematics, so you can decide if you want more.

Go back for a moment to calculus and differential equations. A completely realistic problem is seldom easy, but we can solve models. (Developing a good model is a skill in itself.) One method of solution involves complex numbers:

\[
\begin{align*}
\text{any function } & u(x + iy) \text{ solves } u_{xx} + u_{yy} = 0 \quad \text{(Laplace equation)} \\
\text{any function } & e^{ik(x + \alpha)} \text{ solves } u_{tt} - c^2 u_{xx} = 0 \quad \text{(wave equation)}.
\end{align*}
\]

From those building blocks we assemble solutions. For the wave equation, a signal starts at \( t = 0 \). It is a combination of pure oscillations \( e^{ikx} \). The coefficients in that combination make up the **Fourier transform**—to tell how much of each frequency is in the signal. A lot of engineers and scientists would rather know those Fourier coefficients than \( f(x) \).

A Fourier series breaks the signal into \( \Sigma a_k \cos kx \) or \( \Sigma b_k \sin kx \) or \( \Sigma c_k e^{ikx} \). These sums can be infinite (like power series). Instead of values of \( f(x) \), or derivatives at the basepoint, the function is described by \( a_k, b_k, c_k \). Everything is computed by the "Fast Fourier Transform." This is the greatest algorithm since Newton's method.

A radio signal is near one frequency. A step function has many frequencies. A delta function has every frequency in the same amount: \( \delta(x) = \Sigma \cos kx \). Channel 4 can't broadcast a perfect step function. You wouldn't want to hear a delta function.

We mentioned **computing**. For nonlinear equations this means Newton's method. For \( Ax = b \) it means elimination—**algorithms take the place of formulas**. Exact solutions are gone—speed and accuracy and stability become essential. It seems right to make scientific computing a part of applied mathematics, and teach the algorithms with the theory. My text *Introduction to Applied Mathematics* is one step in this direction, trying to present advanced calculus as it is actually used.

We cannot discuss applications and forget **statistics**. Our society produces oceans of data— somebody has to draw conclusions. To decide if a new drug works, and if oil spills are common or rare, and how often to have a checkup, we can't just guess. I am astounded that the connection between smoking and health was hidden for centuries. It was in the data! Eventually the statisticians uncovered it. Professionals can find patterns, and the rest of us can understand (with a little mathematics) what has been found.

One purpose in studying mathematics is to know more about your own life. Calculus lights up a key idea: **Functions**. Shapes and populations and heart signals and profits and growth rates, all are given by functions. They change in time. They have integrals and derivatives. To understand and use them is a challenge—mathematics takes effort. A lot of people have contributed, in whatever way they could—as you and I are doing. We may not be Newton or Leibniz or Gauss or Einstein, but we can share some part of what they created.
You have met the idea of a matrix. An m by n matrix A has m rows and n columns (it is square if m = n). It multiplies a vector x that has n components. The result is a vector Ax with m components. The central problem of linear algebra is to go backward: From Ax = b, find x. That is possible when A is square and invertible. Otherwise there is no solution x—or there are infinitely many.

The crucial property of matrix multiplication is linearity. If Ax = b and AX = B then A times x + X is b + B. Also A times 2x is 2b. In general A times cx is cb. In particular A times 0 is 0 (one vector has n zeros, the other vector has m zeros). The whole subject develops from linearity. Derivatives and integrals obey linearity too.

**Question 1** What are the solutions to Ax = 0? One solution is x = 0. There may be other solutions and they fill up the "nullspace":

\[
\begin{bmatrix}
1 & 2 \\
0 & 3
\end{bmatrix}
\begin{bmatrix}
x \\
y
\end{bmatrix} =
\begin{bmatrix}
0 \\
0
\end{bmatrix}
\]

requires \( x = 0 \) and \( y = 0 \).

\[
\begin{bmatrix}
x \\
y \\
z
\end{bmatrix} =
\begin{bmatrix}
1 & 2 & 0 \\
0 & 3 & 1
\end{bmatrix}
\begin{bmatrix}
x \\
y \\
z
\end{bmatrix}
\]

also allows \( y = -1 \) and \( z = 3 \).

*When there are more unknowns than equations*—when A has more columns than rows—the system Ax = 0 has many solutions. They are not scattered randomly around! Another solution is \( X = 4, Y = -2, Z = 6 \). This lies on the same line as \( (2, -1, 3) \) and \( (0, 0, 0) \). Always the solutions to Ax = 0 form a “space” of vectors—which brings us to a central idea of linear algebra.

**Note** These pages are not concentrating on the mechanics of multiplying or inverting matrices. Those are explained in all courses. My own teaching emphasizes that Ax is a combination of the columns of A. The solution \( x = A^{-1}b \) is computed by elimination. Here we explain the deeper idea of a vector space—and especially the particular spaces that control Ax = b. I cannot go into the same detail as in my book on Linear Algebra and Its Applications, where examples and exercises develop the new ideas. Still these pages can be a useful support.

All vectors with n components lie in n-dimensional space. You can add them and subtract them and multiply them by any c. (Don’t multiply two vectors and never write \( 1/x \) or \( 1/A \)). The results \( x + X \) and \( x - X \) and \( cx \) are still vectors in the space. Here is the important point:

The line of solutions to Ax = 0 is a “subspace”—a vector space in its own right.

The sum \( x + X \) has components 6, -3, 9—which is another solution. The difference \( x - X \) is a solution, and so is 4x. These operations leave us in the subspace.

The nullspace consists of all solutions to Ax = 0. It may contain only the zero vector (as in the first example). It may contain a line of vectors (as in the second example). It may contain a whole plane of vectors (Problem 5). In every case \( x + X \) and \( x - X \) and \( cx \) are also in the nullspace. We are assigning a new word to an old idea—the equation \( x - 2y = 0 \) has always been represented by a line (its nullspace). Now we have 6-dimensional subspaces of an 8-dimensional vector space.

Notice that \( x^2 - y = 0 \) does not produce a subspace (a parabola instead). Even the x and y axes together, from \( xy = 0 \), do not form a subspace. We go off the axes when we add \( (1, 0) \) to \( (0, 1) \). You might expect the straight line \( x - 2y = 1 \) to be a subspace, but again it is not so. When x and y are doubled, we have \( X - 2Y = 2 \). Then \( (X, Y) \) is on a different line. Only Ax = 0 is guaranteed to produce a subspace.
Figure 16.1 shows the nullspace and "row space." Check dot products (both zero).

\[ A = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 3 & 1 \end{bmatrix}, \quad A^T = \begin{bmatrix} 1 & 0 \\ 2 & 3 \\ 0 & 1 \end{bmatrix} \]

**Fig. 16.1** The nullspace is perpendicular to the rows of \( A \) (the columns of \( A^T \)).

**Question 2** When \( A \) multiplies a vector \( x \), what subspace does \( Ax \) lie in? The product \( Ax \) is a combination of the columns of \( A \)—hence the name "column space":

\[
\begin{bmatrix} 2 & 1 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \end{bmatrix} = 3(\text{column 1}) + 2(\text{column 2}) = \begin{bmatrix} 8 \\ 2 \\ 0 \end{bmatrix}
\]

This is in the column space; \( \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \) is not.

No choice of \( x \) can produce \( Ax = (0, 0, 1) \). For this \( A \), all combinations of the columns end in a \( \boxed{\_} \). The column space is like the xy plane within xyz space. It is a subspace of \( m \)-dimensional space, containing every vector \( b \) that is a combination of the columns:

**The system** \( Ax = b \** has a solution exactly when \( b \) is in the column space.**

When \( A \) has an inverse, the column space is the whole \( n \)-dimensional space. The nullspace contains only \( x = 0 \). There is exactly one solution \( x = A^{-1}b \). This is the good case—and we outline four more key topics in linear algebra.

1. **Basis and dimension of a subspace.** A one-dimensional subspace is a line. A plane has dimension two. The nullspace above contained all multiples of \( (2, -1, 3) \)—by knowing that "basis vector" we know the whole line. The column space was a plane containing column 1 and column 2. Again those vectors are a "basis"—by knowing the columns we know the whole column space.

   Our 2 by 3 matrix has three columns: \((1, 0)\) and \((2, 3)\) and \((0, 1)\). Those are *not* a basis for the column space! This space is only a plane, and three vectors are too many. The dimension is two. By combining \((1, 0)\) and \((0, 1)\) we can produce the other vector \((2, 3)\). There are only two *independent* columns, and they form a basis for this column space.

   In general: When a subspace contains \( r \) independent vectors, and no more, those vectors are a basis and the dimension is \( r \). "Independent" means that no vector is a combination of the others. In the example, \((1, 0)\) and \((2, 3)\) are also a basis. A subspace has many bases, just as a plane has many axes.

2. **Least squares.** If \( Ax = b \) has no solution, we look for the \( x \) that comes closest. Section 11.4 found the straight line nearest to a set of points. We make the length of \( Ax - b \) as small as possible, when zero length is not possible. No vector solves
16.1 Vector Spaces and Linear Algebra

$Ax = b$, when $b$ is not in the column space. So $b$ is projected onto that space. This leads to the "normal equations" that produce the best $x$:

$$A^T Ax = A^T b. \tag{1}$$

When a rectangular matrix appears in applications, its transpose generally comes too. The columns of $A$ are the rows of $A^T$. The rows of $A$ are the columns of $A^T$. Then $A^T A$ is square and symmetric—equal to its transpose and vital for applied mathematics.

3. Eigenvalues (for square matrices only). Normally $Ax$ points in a direction different from $x$. For certain special eigenvectors, $Ax$ is parallel to $x$. Here is a 2 by 2 matrix with two eigenvectors—in one case $Ax = 5x$ and in the other $Ax = 2x$:

$$Ax = \lambda x: \begin{bmatrix} 3 & 2 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 5 \\ 5 \end{bmatrix} = 5 \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} 3 & 2 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} 2 \\ -1 \end{bmatrix} = \begin{bmatrix} 4 \\ -2 \end{bmatrix} = 2 \begin{bmatrix} 2 \\ -1 \end{bmatrix}.$$

The multipliers 5 and 2 are the eigenvalues of $A$. An 8 by 8 matrix has eight eigenvalues, which tell what the matrix is doing (to the eigenvectors). The eigenvectors are uncoupled, and they go their own way. A system of equations $dy/dt = Ay$ acts like one equation—when $y$ is an eigenvector:

$$dy_1/dt = 3y_1 + 2y_2 \quad \text{has the solution} \quad y_1 = e^{5t}, \quad y_2 = e^{3t}.$$

The eigenvector is $(1, 1)$. The eigenvalue $\lambda = 5$ is in the exponent. When you substitute $y_1$ and $y_2$ the differential equations become $5e^{5t} = 5e^{5t}$. The fundamental principle for $dy/dt = cy$ still works for the system $dy/dt = Ay$: Look for pure exponential solutions. The eigenvalue "lamba" is the growth rate in the exponent.

I have to add: Find the eigenvectors also. The second eigenvector $(2, -1)$ has eigenvalue $\lambda = 2$. A second solution is $y_1 = 2e^{2t}, y_2 = -e^{2t}$. Substitute those into the equation—they are even better at displaying the general rule:

**If** $Ax = \lambda x$ **then** $d/dt(e^{\lambda t}x) = A(e^{\lambda t}x)$. **The pure exponentials are** $y = e^{\lambda t}x$.

The four entries of $A$ pull together for the eigenvector. So do the 64 entries of an 8 by 8 matrix—again $e^{\lambda t}x$ solves the equation. Growth or decay is decided by $\lambda > 0$ or $\lambda < 0$. When $\lambda = k + ia$ is a complex number, growth and oscillation combine in $e^{\lambda t} = e^{kt}e^{iat} = e^{kt}(\cos at + i \sin at)$.

Subspaces govern static problems $Ax = b$. Eigenvalues and eigenvectors govern dynamic problems $dy/dt = Ay$. Look for exponentials $y = e^{\lambda t}x$.

4. Determinants and inverse matrices. A 2 by 2 matrix has determinant $D = ad - bc$. This matrix has no inverse if $D = 0$. Reason: $A^{-1}$ divides by $D$:

$$A^{-1} = \frac{1}{D} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \quad \text{times} \quad A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \quad \text{equals} \quad I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}.$$

This pattern extends to $n$ by $n$ matrices, but $D$ and $A^{-1}$ become more complicated.

For 3 by 3 matrices $D$ has six terms. Section 11.5 identified $D$ as a triple product $a \cdot (b \times c)$ of the columns. Three events come together in the singular case: $D$ is zero and $A$ has no inverse and the columns lie in a plane. The opposite events produce the "nonsingular" case: $D$ is nonzero and $A^{-1}$ exists. Then $Ax = b$ is solved by $x = A^{-1}b$. 

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### Notes

- **Ax = b**, $b$ not in column space → projected onto column space → "normal equations" $A^T Ax = A^T b$.
- Rectangular matrix transpose $A^T$ comes with $A$. Columns $A$ are rows $A^T$. Rows $A$ are columns $A^T$.
- $A^T A$ square symmetric vital for applied mathematics.
- **Eigenvalues** for square matrices: $Ax$ in $x$, for special eigenvectors $Ax$ parallel to $x$.
- Example: $3 \times 2$ matrix $A$.
  - $Ax = \lambda x$:
    $$\begin{bmatrix} 3 & 2 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 5 \\ 5 \end{bmatrix} = 5 \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$
  - $Ax = 5x, 2x$.
- **Differential equation**: $dy/dt = Ay$.
  - Pure exponential solution: $e^{\lambda t}$.
  - General rule: $d/dt(e^{\lambda t}x) = A(e^{\lambda t}x)$.
  - **Determinants and inverse**: $A^{-1}$ exists if $D = ad - bc \neq 0$.
  - **Singular case**: $D = 0$, columns lie in a plane, no inverse.
  - **Nonsingular case**: $D \neq 0$, inverse exists.
D is also the product of the pivots and the product of the eigenvalues. The pivots arise in elimination—the practical way to solve $Ax = b$ without $A^{-1}$. To find eigenvalues we turn $Ax = \lambda x$ into $(A - \lambda I)x = 0$. By a nice twist of fate, this matrix $A - \lambda I$ has $D = 0$. Go back to the example:

$$
\begin{bmatrix}
3 & 2 \\
1 & 4 \\
\end{bmatrix} - \lambda
\begin{bmatrix}
1 & 0 \\
0 & 1 \\
\end{bmatrix}
= \begin{bmatrix}
3 - \lambda & 2 \\
1 & 4 - \lambda \\
\end{bmatrix}
$$

has $D = (3 - \lambda)(4 - \lambda) - 2 = \lambda^2 - 7\lambda + 10$.

The equation $\lambda^2 - 7\lambda + 10 = 0$ gives $\lambda = 5$ and $\lambda = 2$. The eigenvalues come first, to make $D = 0$. Then $(A - 5I)x = 0$ and $(A - 2I)x = 0$ yield the eigenvectors. These $x$'s go into $y = e^{\mu x}$ to solve differential equations—which come next.

### 16.1 Exercises

**Read-through questions**

If $Ax = b$ and $AX = B$, then $A$ times $2x + 3X$ equals \_\_\_.

If $Ax = 0$ and $AX = 0$ then $A$ times $2x + 3X$ equals \_\_\_.

In this case $x$ and $X$ are in the \_\_\_ of $A$, and so is the combination \_\_\_. The nullspace contains all solutions to \_\_\_. It is a subspace, which means \_\_\_. If $x = (1, 1, 1)$ is in the nullspace then the columns add to \_\_\_, so they are (independent)(dependent).

Another subspace is the \_\_\_ space of $A$, containing all combinations of the columns. The system $Ax = b$ can be solved when $b$ is \_\_. Otherwise the best solution comes from $A^TAx = _\. \_$. Here $A^T$ is the \_\_\_ matrix, whose rows are \_\_\_. The nullspace of $A^T$ contains all solutions to \_\_\_. The \_\_\_ space of $A^T$ (row space of $A$) is the fourth fundamental subspace. Each subspace has a basis containing as many \_\_\_ vectors as possible. The number of vectors in the basis is the \_\_\_ of the subspace.

When $Ax = \lambda x$, the number $\lambda$ is an \_\_\_ and $x$ is an \_\_\_. The equation $dy/dt = Ay$ has the exponential solution $y = e^\mu$. A 7 by 7 matrix has \_\_\_ eigenvalues, whose product is the _\_\_ of $A$. If $D$ is nonzero the matrix $A$ has an \_\_\_. Then $Ax = b$ is solved by $x = w$. The formula for $D$ contains 7! = 5040 terms, so $x$ is better computed by \_\_. On the other hand $Ax = \lambda x$ means that $A - \lambda I$ has determinant \_\_. The eigenvalue is computed before the \_\_.

**Find the nullspace in 1–6. Along with $x$ go all $cx$.**

1. $A = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$ (solve $Ax = 0$) \[ B = \begin{bmatrix} 12 & -6 \\ -6 & 3 \end{bmatrix} \]

2. $C = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{bmatrix}$ (solve $Cx = 0$) \[ D = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 2 \end{bmatrix} \]

3. $E = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$ \[ F = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \]

5. Change Problem 1 to $Ax = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$ and find all solutions.

6. Change Problem 1 to $Ax = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ and find all solutions.

7. Change Problem 1 to $Ax = \begin{bmatrix} 3 \\ 6 \end{bmatrix}$ (a) Find any particular solution $x_p$. (b) Add any $x_0$ from the nullspace and show that $x_p + x_0$ is also a solution.

8. Change Problem 1 to $Ax = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ and find all solutions.

Graph the lines $x_1 + 2x_2 = 1$ and $2x_1 + 4x_2 = 0$ in a plane.

9. Suppose $Ax_p = b$ and $Ax_0 = 0$. Then by linearity $A(x_p + x_0) = \_\_. \_\_. Conclusion: The sum of a particular solution $x_p$ and any nullvector $x_0$ is \_\_.

10. Suppose $Ax = b$ and $Ax_p = b$. Then by linearity $A(x - x_p) = \_\_. \_\_. The difference between solutions is a vector in \_\_. Conclusion: Every solution has the form $x = x_p + x_0$, one particular solution plus a vector in the nullspace.

11. Find three vectors $b$ in the column space of $E$. Find all vectors $b$ for which $Ex = b$ can be solved.

12. If $Ax = 0$ then the rows of $A$ are perpendicular to $x$. Draw the row space and nullspace (lines in a plane) for $A$ above.

13. Compute $CC^T$ and $C^TC$. Why not $C^T$?

14. Show that $Cx = b$ has no solution, if $b = (-1, 1, 1)$. Find the best solution from $C^TCx = C^Tb$.

15. $C^T$ has three columns. How many are independent? Which ones?

16. Find two independent vectors that are in the column space of $C$ but are not columns of $C$.

17. For which of the matrices $ABCDEF$ are the columns a basis for the column space?
18. Explain the reasoning: If the columns of a matrix $A$ are independent, the only solution to $Ax = 0$ is $x = 0$.

19. Which of the matrices $ABCEF$ have nonzero determinants?

20. Find a basis for the full three-dimensional space using only vectors with positive components.

21. Find the matrix $F^{-1}$ for which $FF^{-1} = I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$.

22. Verify that $(\text{determinant of } F)^2 = (\text{determinant of } F^2)$.

23. (Important) Write down $F - \lambda I$ and compute its determinant. Find the two numbers $\lambda$ that make this determinant zero. For those two numbers find eigenvectors $x$ such that $Fx = \lambda x$.

24. Compute $G = F^2$. Find the determinant of $G - \lambda I$ and the two $\lambda$'s that make it zero. For those two $\lambda$'s find eigenvectors $x$ such that $Gx = \lambda x$. Conclusion: if $Fx = \lambda x$ then $F^2x = \lambda^2 x$.

25. From Problem 23 find two exponential solutions to the equation $dy/dt = Fy$. Then find a combination of those solutions that starts from $y_0 = (1, 0)$ at $t = 0$.

26. From Problem 24 find two solutions to $dy/dt = Gy$. Then find the solution that starts from $y_0 = (2, 1)$.

27. Compute the determinant of $E - \lambda I$. Find all $\lambda$'s that make this determinant zero. Which eigenvalue is repeated?

28. Which previous problem found eigenvectors for $Ex = 0x$? Find an eigenvector for $Ex = 3x$.

29. Find the eigenvalues and eigenvectors of $A$.

30. Explain the reasoning: A matrix has a zero eigenvalue if and only if its determinant is zero.

31. Find the matrix $H$ whose eigenvalues are 0 and 4 with eigenvectors $(1, 1)$ and $(1, -1)$.

32. If $Fx = \lambda x$ then multiplying both sides by $F^{-1}$ gives $F^{-1}x = \lambda^{-1}x$. If $F$ has eigenvalues 1 and 3 then $F^{-1}$ has eigenvalues ______. The determinants of $F$ and $F^{-1}$ are ______.

33. True or false, with a reason or an example.

(a) The solutions to $Ax = b$ form a subspace.

(b) $\begin{bmatrix} 0 & 2 \\ 0 & 0 \end{bmatrix}$ has $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ in its nullspace and column space.

(c) $A^T A$ has the same entry in its upper right and lower left corners.

(d) If $Ax = \lambda x$ then $y = e^{\lambda t}$ solves $dy/dt = Ay$.

(e) If the columns of $A$ are not independent, their combinations still form a subspace.

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16.2 Differential Equations

We just solved differential equations by linear algebra. Those were special systems $dy/dt = Ay$, linear with constant coefficients. The solutions were exponentials, involving $e^{\lambda t}$. The eigenvalues of $A$ were the "growth factors" $\lambda$. This section solves other equations—by no means all. We concentrate on a few that have important applications.

Return for a moment to the beginning—when direct integration was king:

1. $dy/dt = s(t)$
2. $dy/dt = cy$
3. $dy/dt = u(y)c(t)$

In 1, $y(t)$ is the integral of $s(t)$. In 2, $y(t)$ is the integral of $cy(t)$. That sounds circular—it only made sense after the discovery of $y = e^{\lambda t}$. This exponential has the correct derivative $cy$. To find it by integration instead of inventing it, separate $y$ from $t$:

Separation and integration also solve 3: $\int dy/u(y) = \int c(t)dt$. The model logistic equation has $u = y - y^2$ = quadratic. Equation 2 has $u = y$ = linear. Equation 1 is also a special case with $u = 1$ = constant. But 2 and 1 are very different, for the following reason.

The compound interest equation $y' = cy$ is growing from inside. The equation $y' = s(t)$ is growing from outside. Where $c$ is a "growth rate," $s$ is a "source." They don't have the same meaning, and they don't have the same units. The combination $y' = cy + s$ was solved in Chapter 6, provided $c$ and $s$ are constant—but applications force us to go further.
In three examples we introduce non-constant source terms.

**EXAMPLE 1** Solve \( \frac{dy}{dt} = cy + s \) with the new source term \( s = e^{kt} \).

**Method** Substitute \( y = Be^{kt} \), with an "undetermined coefficient" \( B \) to make it right:

\[
kBe^{kt} = cBe^{kt} + e^{kt} \quad \text{yields} \quad B = 1/(k - c).
\]

The source \( e^{kt} \) is the **driving term**. The solution \( Be^{kt} \) is the **response**. The exponent is the same! The key idea is to expect \( e^{kt} \) in the response.

**Initial condition** To match \( y_0 \) at \( t = 0 \), the solution needs another exponential.

It is the **free response** \( Ae^{kt} \), which satisfies \( \frac{dy}{dt} = cy \) with no source. To make \( y = Ae^{kt} + Be^{kt} \) agree with \( y_0 \), choose \( A = y_0 - B \):

**Final solution** \( y = (y_0 - B)e^{kt} + Be^{kt} = y_0e^{kt} + (e^{kt} - e^{kt})(k - c). \) (1)

**Exceptional case** \( B = 1/(k - c) \) grows larger as \( k \) approaches \( c \). When \( k = c \) the method breaks down—the response \( Be^{kt} \) is no longer correct. The solution (1) approaches 010, and in the limit we get a derivative. It has an extra factor \( t \):

\[
\frac{e^{kt} - e^{kt}}{k - c} = \frac{\text{change in } e^{kt}}{\text{change in } c} \rightarrow \frac{d}{dc}(e^{ct}) = te^{kt}.
\] (2)

The correct response is \( te^{kt} \) when \( k = c \). This is the form to substitute, when the driving rate \( k \) equals the natural rate \( c \) (called **resonance**).

Add the free response \( y_0e^{kt} \) to match the initial condition.

**EXAMPLE 2** Solve \( \frac{dy}{dt} = cy + s \) with the new source term \( s = \cos kt \).

Substitute \( y = B \sin kt + D \cos kt \). This has two undetermined coefficients \( B \) and \( D \):

\[
kB \cos kt - kD \sin kt = c(B \sin kt + D \cos kt) + \cos kt.
\] (3)

Matching cosines gives \( kB = cD + 1 \). The sines give \( -kD = cB \). Algebra gives \( B, D; y \):

\[
B = \frac{c}{k^2 + c^2} \quad D = \frac{k}{k^2 + c^2} \quad y = \frac{c \sin kt + k \cos kt}{k^2 + c^2}.
\] (4)

**Question** Why do we need both \( B \sin kt \) and \( D \cos kt \) in the response to \( \cos kt \)?

**First Answer** Equation (3) is impossible if we leave out \( B \) or \( D \).

**Second Answer** \( \cos kt \) is \( \frac{1}{2}e^{ikt} + \frac{1}{2}e^{-ikt} \). So \( e^{ikt} \) and \( e^{-ikt} \) are both in the response.

**EXAMPLE 3** Solve \( \frac{dy}{dt} = cy + s \) with the new source term \( s = te^{kt} \).

**Method** Look for \( y = Be^{kt} + Dte^{kt} \). Problem 13 determines \( B \) and \( D \). Add \( Ae^{kt} \) as needed, to match the initial value \( y_0 \).

**SECOND-ORDER EQUATIONS**

The equation \( \frac{dy}{dt} = cy \) is **first-order**. The equation \( \frac{d^2y}{dt^2} = -cy \) is **second-order**.

The first is typical of problems in life sciences and economics—the rate \( \frac{dy}{dt} \) depends on the present situation \( y \). The second is typical of engineering and physical sciences—the acceleration \( \frac{d^2y}{dt^2} \) enters the equation.

If you put money in a bank, it starts growing immediately. If you turn the wheels of a car, it changes direction **gradually**. The path is a curve, not a sharp corner.

Newton's law is \( F = ma \), not \( F = mv \).
A mathematician compares a straight line to a parabola. The straight line crosses the x axis no more than once. The parabola can cross twice. The equation \( ax^2 + bx + c = 0 \) has two solutions, provided we allow them to be complex or equal. These are exactly the possibilities we face below: two real solutions, two complex solutions, or one solution that counts twice. The quadratic could be \( x^2 - 1 \) or \( x^2 + 1 \) or \( x^2 \). The roots are \( 1 \) and \( -1 \), \( i \) and \( -i \), \( 0 \) and \( 0 \).

In solving differential equations the roots appear in the exponent, and are called \( \lambda \).

**EXAMPLE 4** \( y'' + y = 0 \) solutions \( y = e^t \) and \( y = e^{-t} \) \( \lambda = 1, -1 \)

**EXAMPLE 5** \( y'' + y = 0 \) solutions \( y = 1 \) and \( y = t \) \( \lambda = 0, 0 \)

**EXAMPLE 6** \( y'' - y = 0 \) solutions \( y = \cos t \) and \( y = \sin t \) \( \lambda = i, -i \)

Where are the complex solutions? They are hidden in Example 6, which could be written \( y = e^{it} \) and \( y = e^{-it} \). These satisfy \( y'' = -y \) since \( i^2 = -1 \). The use of sines and cosines avoids the imaginary number \( i \), but it breaks the pattern of \( e^{\lambda t} \).

Example 5 also seems to break the pattern—again \( e^{\lambda t} \) is hidden. The solution \( y = 1 \) is \( e^0 \). The other solution \( y = t \) is \( te^0 \). The zero exponent is repeated—another exceptional case that needs an extra factor \( t \).

Exponentials solve every equation with constant coefficients and zero right hand side: To solve \( ay'' + by' + cy = 0 \) substitute \( y = e^{\lambda t} \) and find \( \lambda \).

This method has three steps, leading to the right exponents \( \lambda = r \) and \( \lambda = s \):

1. With \( y = e^{\lambda t} \) the equation is \( a\lambda^2 e^{\lambda t} + b\lambda e^{\lambda t} + ce^{\lambda t} = 0 \). Cancel \( e^{\lambda t} \).
2. Solve \( a\lambda^2 + b\lambda + c = 0 \). Factor or use the formula \( \lambda = (-b \pm \sqrt{b^2 - 4ac})/2a \).
3. Call those roots \( \lambda = r \) and \( \lambda = s \). The complete solution is \( y = Ae^{rt} + Be^{st} \).

The pure exponentials are \( y = e^{\lambda t} \) and \( y = e^{-\lambda t} \). Depending on \( r \) and \( s \), they grow or decay or oscillate. They are combined with constants \( A \) and \( B \) to match the two conditions at \( t = 0 \). The initial state \( y_0 \) equals \( A + B \). The initial velocity \( y'_0 \) equals \( rA + sB \) (the derivative at \( t = 0 \)).

**EXAMPLE 7** Solve \( y'' - 3y' + 2y = 0 \) with \( y_0 = 5 \) and \( y'_0 = 4 \).

Step 1 substitutes \( y = e^{\lambda t} \). The equation becomes \( \lambda^2 e^{\lambda t} - 3\lambda e^{\lambda t} + 2e^{\lambda t} = 0 \). Cancel \( e^{\lambda t} \).
Step 2 solves \( \lambda^2 - 3\lambda + 2 = 0 \). Factor into \( (\lambda - 1)(\lambda - 2) = 0 \). The exponents \( r \), \( s \) are 1, 2.
Step 3 produces \( y = Ae^t + Be^{2t} \). The initial conditions give \( A + B = 5 \) and \( 1A + 2B = 4 \).
The constants are \( A = 6 \) and \( B = -1 \). The solution is \( y = 6e^t - e^{2t} \).

This solution grows because there is a positive \( \lambda \). The equation is "unstable." It becomes stable when the middle term \( -3y' \) is changed to \( +3y' \). When the damping is positive the solution decays. The \( \lambda \)'s are negative:

**EXAMPLE 8** \( (\lambda^2 + 3\lambda + 2) \) factors into \( (\lambda + 1)(\lambda + 2) \). The exponents are \(-1\) and \(-2\).
The solution is \( y = Ae^{-t} + Be^{-2t} \). It decays to zero for any initial condition.

**EXAMPLES 9-10** Solve \( y'' + 2y' + 2y = 0 \) and \( y'' + 2y' + y = 0 \). How do they differ?

Key difference \( \lambda^2 + 2\lambda + 2 \) has complex roots, \( \lambda^2 + 2\lambda + 1 \) has a repeated root:

\[ \lambda^2 + 2\lambda + 2 = 0 \] gives \( \lambda = -1 \pm i \)  \( (\lambda + 1)^2 = 0 \) gives \( \lambda = -1, -1 \).

The \(-1\) in all these \( \lambda \)'s means decay. The \( i \) means oscillation. The first exponential is \( e^{-\lambda t} \), which splits into \( e^{-t} \) (decay) times \( e^{it} \) (oscillation). Even better, change \( e^{it} \)
and $e^{-it}$ into cosines and sines:

$$y = Ae^{-t + i\lambda} + Be^{-t - i\lambda} = e^{-t}(a \cos \lambda + b \sin \lambda).$$  \hspace{1cm} (5)

At $t = 0$ this produces $y_0 = a$. Then matching $y_b$ leads to $b$.

Example 10 has $r = s = -1$ (repeated root). One solution is $e^{-t}$ as usual. The second solution cannot be another $e^{-t}$. Problem 21 shows that it is $te^{-t}$—again the exceptional case multiplies by $t$! The general solution is $y = Ae^{-t} + Bte^{-t}$.

Without the damping term $2y'$, these examples are $y'' + 2y = 0$ or $y'' + y = 0$—pure oscillation. A small amount of damping mixes oscillation and decay. Large damping gives pure decay. The borderline is when $\lambda$ is repeated ($r = s$). That occurs when $b^2 - 4ac$ in the square root is zero. The **borderline between two real roots and two complex roots is two repeated roots**.

The method of solution comes down to one idea: Substitute $y = e^{\lambda t}$. The equations apply to mechanical vibrations and electrical circuits (also other things, but those two are of prime importance). While describing these applications I will collect the information that comes from $\lambda$.

**SPRINGS AND CIRCUITS: MECHANICAL AND ELECTRICAL ENGINEERING**

A mass is hanging from a spring. We pull it down an extra distance $y_0$ and give it a starting velocity $y_b$. The mass moves up or down, obeying Newton's law: **mass times acceleration equals spring force plus damping force**:

$$my'' = -kx - dy' \quad \text{or} \quad my'' + dy' + ky = 0. \hspace{1cm} (6)$$

This is free oscillation. The spring force $-kx$ is proportional to the stretching $x$ (Hooke's law). The damping acts like a shock absorber or air resistance—it takes out energy. Whether the system goes directly toward zero or swings back and forth is decided by the three numbers $m, d, k$. They were previously called $a, b, c$.

**16A** The solutions $e^{\lambda t}$ to $my'' + dy' + ky = 0$ are controlled by the roots of $m\lambda^2 + d\lambda + k = 0$. With $d > 0$ there is damping and decay. From $\sqrt{d^2 - 4mk}$ there may be oscillation:

- **overdamping**: $d^2 > 4mk$ gives real roots and pure decay (Example 8)
- **underdamping**: $d^2 < 4mk$ gives complex roots and oscillation (Example 9)
- **critical damping**: $d^2 = 4mk$ gives a real repeated root $-d/2m$ (Example 10)

We are using letters when the examples had numbers, but the results are the same:

$$m\lambda^2 + d\lambda + k = 0 \quad \text{has roots} \quad r, s = -\frac{d}{2m} \pm \frac{1}{2m} \sqrt{d^2 - 4mk}.$$  

Overdamping has no imaginary parts or oscillations: $y = Ae^{\lambda t} + Be^{\mu t}$. Critical damping has $r = s$ and an exceptional solution with an extra $t$: $y = Ae^{\lambda t} + Bte^{\mu t}$. (This is only a solution when $r = s$.) Underdamping has decay from $-d/2m$ and oscillation from the imaginary part. An undamped spring ($d = 0$) has pure oscillation at the natural frequency $\omega_0 = \sqrt{k/m}$.

**All these possibilities are in Figure 16.2**, created by Alar Toomre. At the top is pure oscillation ($d = 0$ and $y = \cos 2t$). The equation is $y'' + 4y = 0$ and $d$ starts to grow. When $d$ reaches 4, the quadratic is $\lambda^2 + 4\lambda + 4$ or $(\lambda + 2)^2$. The repeated root
yields $e^{-2t}$ and $te^{-2t}$. After that the oscillation is gone. There is a smooth transition from one case to the next—as complex roots join in the repeated root and split into real roots.

At the bottom right, the final value $y(2\pi)$ increases with large damping. This was a surprise. At $d = 5$ the roots are $-1$ and $-4$. At $d = 8.5$ the roots are $-\frac{1}{2}$ and $-8$. The small root gives slow decay (like molasses). As $d \to \infty$ the solution approaches $y = 1$.

If we are serious about using mathematics, we should take advantage of anything that helps. For second-order equations, the formulas look clumsy but the examples are quite neat. The idea of $e^{at}$ is absolutely basic. The good thing is that electrical circuits satisfy the same equation. There is a beautiful analogy between springs and circuits:

$$
\begin{align*}
\text{mass } m & \leftrightarrow \text{inductance } L \\
\text{damping constant } d & \leftrightarrow \text{resistance } R \\
\text{elastic constant } k & \leftrightarrow 1/(\text{capacitance } C)
\end{align*}
$$

The resistor takes out energy as the shock absorber did—converting into heat by friction. Without resistance we have pure oscillation. Electric charge is stored in the capacitor (like potential energy). It is released as current (like kinetic energy). It is stored up again (like a stretched spring). This continues at a frequency $\omega_0 = 1/\sqrt{LC}$ (like the spring's natural frequency $\sqrt{k/m}$). These analogies turn mechanical engineers into electrical engineers and vice versa.

The equation for the current $y(t)$ now includes a driving term on the right:

$$
L \frac{dy}{dt} + R y + \frac{1}{C} \int y \, dt = \text{applied voltage} = V \sin \omega t. 
$$

To match networks with springs, differentiate both sides of (7):

$$
Ly'' + Ry' + y/C = V\omega \cos \omega t. 
$$

The oscillations are free when $V = 0$ and forced when $V \neq 0$. The free oscillations $e^{at}$ are controlled by $L\lambda^2 + R\lambda + 1/C = 0$. Notice the undamped case $R = 0$ when
\[ \lambda = \pm i\sqrt{LC} \]. This shows the natural frequency \( \omega_0 = 1/\sqrt{LC} \). Damped free oscillations are in the exercises—what is new and important is the forcing from the right hand side. Our last step is to solve equation (8).

**PARTICULAR SOLUTIONS—THE METHOD OF UNDETERMINED COEFFICIENTS**

The forcing term is a multiple of \( \cos \omega t \). The "particular solution" is a multiple of \( \cos \omega t \) plus a multiple of \( \sin \omega t \). To discover the undetermined coefficients in \( y = a \cos \omega t + b \sin \omega t \), substitute into the differential equation (8):

\[
- L\omega^2(a \cos \omega t + b \sin \omega t) + R\omega(-a \sin \omega t + b \cos \omega t) + (a \cos \omega t + b \sin \omega t)/C = V\omega \cos \omega t.
\]

The terms in \( \cos \omega t \) and the terms in \( \sin \omega t \) give two equations for \( a \) and \( b \):

\[
-aw^2 + bR\omega + a/C = V\omega \quad \text{and} \quad -bw^2 - aR\omega + b/C = 0. \tag{9}
\]

**EXAMPLE 11** Solve \( y'' + y = \cos \omega t \). The oscillations are forced at frequency \( \omega \). The oscillations are free \( (y'' + y = 0) \) at frequency 1. The solution contains both.

**Particular solution** Set \( y = a \cos \omega t + b \sin \omega t \) at the driving frequency \( \omega \), and (9) becomes

\[
-aw^2 + bR\omega + a/C = V\omega \quad \text{and} \quad -bw^2 - aR\omega + b/C = 0.
\]

The second equation gives \( b = 0 \). No sines are needed because the problem has no \( dy/dt \). The first equation gives \( a = 1/(1 - \omega^2) \), which multiplies the cosine:

\[ y = (\cos \omega t)/(1 - \omega^2) \] solves \( y'' + y = \cos \omega t \). \tag{10}

**General solution** Add to this particular solution any solution to \( y'' + y = 0 \):

\[ y = y_{\text{particular}} + y_{\text{homogeneous}} = (\cos \omega t)/(1 - \omega^2) + C \cos \omega t + D \sin \omega t. \tag{11} \]

**Problem of resonance** When the driving frequency is \( \omega = 1 \), the solution (11) becomes meaningless—its denominator is zero. **Reason:** The natural frequency in \( \cos t \) and \( \sin t \) is also 1. A new particular solution comes from \( t \cos t \) and \( t \sin t \).

The key to success is to know the form for \( y \). The table displays four right hand sides and the correct \( y \)'s for any constant-coefficient equation:

<table>
<thead>
<tr>
<th>Right hand side</th>
<th>Particular solution</th>
</tr>
</thead>
<tbody>
<tr>
<td>( e^{\lambda t} )</td>
<td>( y = Be^{\lambda t} ) (same exponent)</td>
</tr>
<tr>
<td>( \cos \omega t ) or ( \sin \omega t )</td>
<td>( y = a \cos \omega t + b \sin \omega t ) (include both)</td>
</tr>
<tr>
<td>polynomial in ( t )</td>
<td>( y = \text{polynomial of the same degree} )</td>
</tr>
<tr>
<td>( e^{\lambda t} \cos \omega t ) or ( e^{\lambda t} \sin \omega t )</td>
<td>( y = ae^{\lambda t} \cos \omega t + be^{\lambda t} \sin \omega t )</td>
</tr>
</tbody>
</table>

**Exception** If one of the roots \( \lambda \) for free oscillation equals \( k \) or \( i\omega \) or 0 or \( k + i\omega \), the corresponding \( y \) in the table is wrong. The proposed solution would give zero on the right hand side. The correct form for \( y \) includes an extra \( t \). All particular solutions are computed by substituting into the differential equation.

**Apology** Constant-coefficient equations hardly use calculus (only \( e^{\lambda t} \)). They reduce directly to algebra (substitute \( y \), solve for \( \lambda \) and \( a \) and \( b \)). I find the S-curve from the logistic equation much more remarkable. The nonlinearity of epidemics or heartbeats or earthquakes demands all the calculus we know. The solution is not so predictable. The extreme of unpredictability came when Lorenz studied weather prediction and discovered chaos.
NUMERICAL METHODS

Those four pages explained how to solve linear equations with constant coefficients: 
Substitute \( y = e^{\lambda t} \). The list of special solutions becomes longer in a course on 
differential equations. But for most nonlinear problems we enter another world—
where solutions are numerical and approximate, not exact.

In actual practice, numerical methods for \( \frac{dy}{dt} = f(t, y) \) divide in two groups:

1. Single-step methods like Euler and Runge-Kutta
2. Multistep methods like Adams-Bashforth

The unknown \( y \) and the right side \( f \) can be vectors with \( n \) components. The notation 
stays the same: the step is \( \Delta t = h \), the time \( t_n \) is \( nh \), and \( y_n \) is the approximation to 
the true \( y \) at that time. We test the first step, to find \( y_1 \) from \( y_0 = 1 \). The equation is 
\( dy/dt = y \), so the right side is \( f = y \) and the true solution is \( y = e^t \).

Notice how the first value \( f \) (in this case 1) is used inside the second \( f \):

**Improved Euler** 
\[
y_{n+1} = y_n + \frac{1}{2} h \left[ f(t_n, y_n) + f(t_{n+1}, y_n + hf(y_n, t_n)) \right]
\]

TEST \[
y_1 = 1 + \frac{1}{2} h [1 + (1 + h)] = 1 + h + \frac{1}{2} h^2
\]

At time \( h \) the true solution equals \( e^h \). Its infinite series is correct through \( h^2 \) 
for Improved Euler (a second-order method). The ordinary Euler method 
\( y_{n+1} = y_n + hf(t_n, y_n) \) is first-order. TEST: \( y_1 = 1 + h \). Now try Runge-Kutta 
(a fourth-order method):

**Runge-Kutta** 
\[
y_{n+1} = y_n + \frac{1}{6} \left[ k_1 + 2k_2 + 2k_3 + k_4 \right] \quad \text{with} \quad k_1 = hf(t_n, y_n)
\]
\[
k_2 = hf(t_n + \frac{1}{2} h, y_n + \frac{1}{2} k_1) \quad k_3 = hf(t_n + \frac{1}{2} h, y_n + \frac{1}{2} k_2) \quad k_4 = hf(t_n + h, y_n + k_3)
\]

Now the first value \( f \) is used in the second (for \( k_2 \)), the second is used in the third, 
and then \( k_3 \) is used in \( k_4 \). The programming is easy. Check the accuracy with another 
test on \( dy/dt = y \):

TEST \[
y_1 = 1 + \frac{1}{6} \left[ h + 2h \left( 1 + \frac{h}{2} \right) + 2h \left( 1 + h \left( 1 + \frac{h}{2} \right) \right) + h \left( 1 + h \left( 1 + \frac{h}{2} \left( 1 + \frac{h}{2} \right) \right) \right) \right]
\]
\[
= 1 + h + \frac{h^2}{2} + \frac{h^3}{6} + \frac{h^4}{24}. \quad \text{This answer agrees with} \ e^h \text{through} \ h^4.
\]

These formulas are included in the book so that you can apply them directly—
for example to see the S-shape from the logistic equation with \( f = cy - by^2 \).

Multistep formulas are simpler and quicker, but they need a single-step method to 
get started. Here is \( y_4 \) in a fourth-order formula that needs \( y_0, y_1, y_2, y_3 \). Just shift 
all indices for \( y_3, y_2, \) and \( y_{n+1} \):

**Multistep** 
\[
y_4 = y_3 + \frac{h}{24} \left[ 55y_3 - 59y_2 + 37y_1 - 9y_0 \right].
\]

The advantage is that each step needs only one new evaluation of \( y_n = f(t_n, y_n) \). 
Runge-Kutta needs four evaluations for the same accuracy.

**Stability** is the key requirement for any method. Now the good test is \( y' = -y \). The 
solution should decay and not blow up. Section 6.6 showed how a large time step 
makes Euler's method unstable—the same will happen for more accurate formulas. 
The price of total stability is an "implicit method" like \( y_1 = y_0 + \frac{h}{2}(y_0 + y_1) \), 
where the unknown \( y_1 \) appears also in \( y_1 \). There is an equation to be solved at every 
step. Calculus is ending as it started—with the methods of Isaac Newton.
Read-through questions

The solution to \( y' - 5y = 10 \) is \( y = Ae^{5t} + B \). The homogeneous part \( Ae^{5t} \) satisfies \( y' - 5y = 0 \). The particular solution \( B \) equals \( \frac{10}{5} \). The initial condition \( y_0 \) is matched by \( A = y_0 - B \). For \( y' - 5y = e^t \) the right form is \( y = Ae^{5t} + e^t \). For \( y' - 5y = \cos t \) the form is \( y = Ae^{5t} + \cos t \). The equation \( y'' + 4y' + 5y = 0 \) is second-order because \( \frac{d^2}{dx^2} \). The pure exponential solutions come from the roots of \( r^2 + 5r + 6 = 0 \), which are \( r = -2 \) and \( r = -3 \). The general solution is \( y = Ae^{-2t} + Be^{-3t} \). Changing \( 4y' \) to \( 4r \) yields pure oscillation. Changing to \( 2y' \) yields \( \lambda = -1 \pm 2i \), when the solutions become \( y = e^{-t} (A \cos 2t + B \sin 2t) \). This oscillation is (over)(under) (critically) damped. A spring with \( m = 1 \), \( d = 2 \), \( k = 5 \) goes (back and forth) (directly to zero). An electrical network with \( L = 1 \), \( R = 2 \), \( C = \frac{1}{10} \) also \( \frac{d}{dt} \).

One particular solution of \( y'' + 4y' + 5y = 0 \) is \( e^t \) times \( e^t \). If the right side is \( \cos t \), the form of \( y_p = \cos t \). If the right side is \( \sin t \), we have resonance and \( y_p \) contains an extra factor \( e^t \).

Problems 1–14 are about first-order linear equations.

1. Substitute \( y - Be^{3t} \) into \( y' - 5y = 8e^{3t} \) to find a particular solution.
2. Substitute \( y = a \cos 2t + b \sin 2t \) into \( y' + y = 4 \sin 2t \) to find a particular solution.
3. Substitute \( y = a + b t + c t^2 \) into \( y' + y = 1 + t^2 \) to find a particular solution.
4. Substitute \( y = ad \cos t + be \sin t \) into \( y' = 2ad \cos t + 2be \sin t \) to find a particular solution.
5. In Problem 1 we can add \( Ae^t \) because this solves the equation \( y' - 5y = 0 \). Choose \( A \) so that \( y(0) = 7 \).
6. In Problem 2 we can add \( Ae^{-j} \), which solves \( 2y' + y = 0 \). Choose \( A \) to match \( y(0) = 0 \).
7. In Problem 3 we add \( Ae^{-j} \) to match \( y(0) = 2 \).
8. In Problem 4 we can add \( y = A \). Why?
9. Starting from \( y_0 = 0 \) solve \( y' = e^t \) and also solve \( y' = 1 \). Show that the first solution approaches the second as \( k \to 0 \).
10. Solve \( y' = y = e^t \) starting from \( y_0 = 0 \). What happens to your formula as \( k \to 1 \)? By l'Hôpital's rule show that \( y \) approaches \( e^t \) as \( k \to 1 \).
11. Solve \( y' - y = e^t + \cos t \). What form do you assume for \( y \) with two terms on the right side?
12. Solve \( y' + y = e^t + t \). What form do you assume for \( y \)?
13. Solve \( y' = cy + te^t \) following Example 3 \( c \neq 1 \).

14. Solve \( y' = y + t \) following Example 3 \( c = 1 \) and \( k = 0 \).

Problems 15–28 are about second-order linear equations.

15. Substitute \( y = e^{j} \) into \( y'' + 6y' + 5y = 0 \). (a) Find all \( \lambda \) and \( \mu \). (b) The solution decays because \( \mu \). (c) The general solution with constants \( A \) and \( B \) is \( y = Ae^{\lambda t} + Be^{\mu t} \).
16. Substitute \( y = e^{j} \) into \( y'' + 9y = 0 \). (a) Find all \( \lambda \). (b) The solution oscillates because \( 2\lambda = 0 \). (c) The general solution with constants \( a \) and \( b \) is \( y = a \cos \lambda t + b \sin \lambda t \).
17. Substitute \( y = e^{j} \) into \( y'' + 2y' + 3y = 0 \). Find both \( \lambda \). (a) The solution oscillates as it decays because \( \lambda \). (b) The general solution with \( e^{\lambda t} \) and \( e^{\mu t} \) is \( y = a \cos \lambda t + b \sin \lambda t \).
18. Substitute \( y = e^{j} \) into \( y'' + 6y' + 9y = 0 \). (a) Find all \( \lambda \). (b) The general solution with \( e^{\lambda t} \) and \( e^{\mu t} \) is \( y = a \cos \lambda t + b \sin \lambda t \).
19. For \( y'' + dy' + y = 0 \) find the type of damping at \( d = 0, 1, 2, 3 \).
20. For \( y'' + 2y' + ky = 0 \) find the type of damping at \( k = 0, 1, 2 \).
21. If \( \lambda^2 + b \lambda + c = 0 \) has a repeated root prove it is \( \lambda = -b/2 \). In this case compute \( y'' + by' + cy \) when \( y = te^t \).
22. \( \lambda^2 + 2\lambda + 2 = 0 \) has roots \( -1 \) and \( -2 \) (not repeated). Show that \( te^{-t} \) does not solve \( y'' + 3y' + 2y = 0 \).
23. Find \( y = a \cos t + b \sin t \) to solve \( y'' + y' + y = \cos t \).
24. Find \( y = a \cos \omega t + b \sin \omega t \) to solve \( y'' + y' + y = \sin \omega t \).
25. Solve \( y'' + 9y = \cos 3t \) with \( y_0 = 0 \) and \( y'_0 = 0 \). The solution contains \( \cos 3t \) and \( \cos 5t \).
26. The difference \( \cos 5t - \cos 3t \) equals \( 2 \sin 4t \). Graph it to see fast oscillations inside slow oscillations (beats).
27. The solution to \( y'' + \omega^2 y = \cos \omega t \) with \( y_0 = 0 \) and \( y'_0 = 0 \) is what multiple of \( \cos \omega t - \cos \omega t \)? The formula breaks down when \( \omega = \) 
28. Substitute \( y = Ae^{jkt} \) into the circuit equation \( Ly' + Ry + f y dt/C = Ve^{jkt} \). Cancel \( e^{jkt} \) to find \( A \). Its denominator is the impedance.

Problems 29–32 have the four right sides in the table (end of section). Find \( y_{\text{particular}} \) by using the correct form.

29. \( y'' + 3y = e^t \)
30. \( y'' + 3y = \sin t \)
31. \( y'' + 2y = 1 + t \)
32. \( y'' + 2y = e' \cos t \).
33. Find the coefficients of \( y \) in Problems 29–31 for which the forms in the table are wrong. Why are they wrong? What new forms are correct?
34 The magic factor $t$ entered equation (2). The series for $e^{kt} - e^{ct}$ starts with $1 + kt + \frac{1}{2}kt^2$ minus $1 + ct + \frac{1}{2}ct^2$. Divide by $k - c$ and set $k = c$ to start the series for $te^t$.

35 Find four exponentials $y = e^t$ for $d^4y/dt^4 - y = 0$.

36 Find a particular solution to $d^4y/dt^4 + y = e^t$.

37 The solution is $y = Ae^{-2t} + Be^{-3t}$ when $d = 4$ in Figure 16.2. Choose $A$ and $B$ to match $y_0 = 1$ and $y_0' = 0$. How large is $y(2t)$?

38 When $d$ reaches 5 the quadratic for Figure 16.2 is $k^2 + 5k + 4 = (k + 1)(k + 4)$. Match $y = Ae^{-x} + Be^{-4x}$ to $y_0 = 1$ and $y_0' = 0$. How large is $y(2n)$?

39 When the quadratic for Figure 16.2 has roots $-r$ and $-4/r$, the solution is $y = Ae^{-x} + Be^{-4x}e^{x}$.

   (a) Match the initial conditions $y_0 = 1$ and $y_0' = 0$.

   (b) Show that $y$ approaches 1 as $r \to 0$.

40 In one sentence tell why $y' = 6y$ has exponential solutions but $y'' = 6y^2$ does not. What power $y = x^n$ solves this equation?

41 The solution to $dy/dt = f(t)$, with no $y$ on the right side, is $y = \int f(t) dt$. Show that the Runge-Kutta method becomes Simpson's Rule.

42 Test all methods on the logistic equation $y' = y - y^2$ to see which gives $y_m = 1$ most accurately. Start at the inflection point $y_0 = \frac{1}{r}$ with $h = r$. Begin the multistep method with exact values of $y = (1 + e^{-t})^{-1}$.

43 Extend the tests of Improved Euler and Runge-Kutta to $y' = -y$ with $y_0 = 1$. They are stable if $|y_1| \leq 1$. How large can $h$ be?

44 Apply Runge-Kutta to $y' = -100y + 100\sin t$ with $y_0 = 0$ and $h = .02$. Increase $h$ to .03 to see that instability is no joke.

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16.3 Discrete Mathematics: Algorithms

Discrete mathematics is not like calculus. Everything is finite. I can start with the 50 states of the U.S. I ask if Maine is connected to California, by a path through neighboring states. You say yes. I ask for the shortest path (fewest states on the way). You get a map and try all possibilities (not really all—but your answer is right). Then I close all boundaries between states like Illinois and Indiana, because one has an even number of letters and the other has an odd number. Is New York still connected to Washington? You ask what kind of game this is—but I hope you will read on.

Far from being dumb, or easy, or useless, discrete mathematics asks good questions. It is important to know the fastest way across the country. It is more important to know the fastest way through a phone network. When you call long distance, a quick connection has to be found. Some lines are tied up, like Illinois to Indiana, and there is no way to try every route.

The example connects New York to New Jersey (7 letters and 9). Washington is connected to Oregon (10 letters and 6). As you read those words, your mind jumps to this fact—there is no path from New York with 7 letters to Washington with 10. Somewhere you must get stuck. There might be a path between all states with an odd number of letters—I doubt it. Graph theory gives a way to find out.

**GRAPHS**

A model for a large part of finite mathematics is a graph. It is not the graph of $y = f(x)$. The word "graph" is used in a totally different way, for a collection of nodes and edges. The nodes are like the 50 states. The edges go between two nodes—the neighboring states. A network of computers fits this model. So do the airline connections between cities. A pair of cities may or may not have an edge between them—depending on flight schedules. The model is determined by $V$ and $E$. 