

PELEG BAR SAPIR

PHYSICS SIMULATIONS (LECTURE NOTES)

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1.4.1 *Radians*

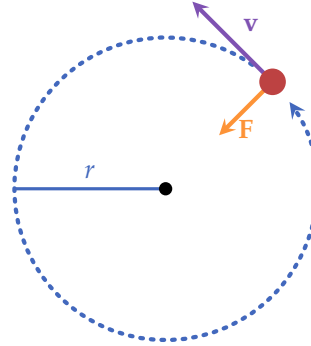
In the context of trigonometry in mathematics and physics, we most commonly use **radians** in place of **degrees**. This sometimes causes some confusion with newer students, so I find it appropriate to briefly discuss the what and why of radians and their usage.

Degrees are defined as a measurement of **angles**. That is, for historical and practical reasons we define a full rotation as 360° . The main problem with using angles is that they are units by themselves, i.e. they differ from lengths. This can cause some consistency issues with regards to units. Consider for example the following equation for a **centripetal force**:

$$F = m \frac{v^2}{r}, \quad (1.1)$$

where m is the mass of the rotating object, v^2 is the square of the linear velocity the object experiences in each time instance, and r is the distance of the rotating object from the center of rotation (Figure 1.1).

Figure 1.1: Object with mass m in perfect circular motion with distance r around a central point. The force acting on the object is directed towards the center and has value $F = m \frac{v^2}{r}$.



In the standard SI scheme, F has units of $[\text{N}]$ ("Newtons"), i.e. $[\text{kg m s}^{-1}]$. We know that m has units of $[\text{kg}]$, v of $[\text{m s}^{-1}]$ and r units of $[\text{m}]$ - so both sides of the equation have the same units.

However, nothing prevents us from measuring the rotation associated with $\frac{v^2}{r}$ directly using angular velocity ω - which can be defined as the amount of degrees of rotation per time T if the rotation is uniform:

$$\omega = \frac{\theta}{T}, \quad (1.2)$$

or more generally as the time derivative of the angle,

$$\omega = \frac{d\theta}{dt} = \dot{\theta}. \quad (1.3)$$

If we measure the angle θ in degrees, we get that the units of angular velocity is $[\text{° s}^{-1}]$. If we then want to use ω in the centripetal force equation, we must somehow cancel the degrees unit to get the proper units of $[\text{N}]$. This obviously leads to a somewhat cumbersome equation.

A much better approach is to measure the *ratio* between the length of an arc created by our angle at a radius R and the same radius itself (Figure 1.2). This gives a unitless measure which we call *radians*, and can be either ignored in the unit calculation (since it is unitless), or simply denoted $[\text{rad}]$.

Another way to view radians is that they measure an arc length in units of the radius (not necessarily a unit radius). If we scale the radius by any scalar, an arc length of the same angle will scale by exactly the same scalar, and so the amount of radians we measure for the arc length won't change (the ratio stays the same, as both the numerator and denominator are scaled by the same number).

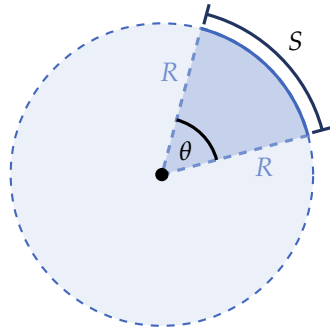


Figure 1.2: Measuring angles using radians. We express the length of S in units of R , e.g. if $S = 1$ then the arc length is equal to R , if $S = \pi$ the arc length is half that of the entire circumference of the circle, etc. The ratio S/R is constant for any real positive value of R , and thus can be used to uniquely describe the angle θ .

We can easily replace any measurement of angles in degrees by a corresponding measurement in radians, since any amount of degrees has a 1-to-1 correspondence with a single value in radians. Let us find this correspondence: since there are 2π radii in a full circle, and a full angle is 360° , we see that each radii length of an arc corresponds to $\frac{360^\circ}{2\pi} = \frac{180^\circ}{\pi} \approx 57.2958^\circ$.

Now, using the radian unit to measure an angle θ , we get that ω has units of $[\text{rad s}^{-1}]$, or simply $[\text{s}^{-1}]$. In turn, if we raise ω to the 2nd power and multiply it by the distance r between the rotating object and center of rotation, and the mass m of the object, we get the units $[\text{kg m s}^{-1}]$, which is exactly the explicit form of Newtons. Indeed, using this perspective, Equation 1.1 can be written as

$$F = m\omega^2 r. \quad (1.4)$$

In my mind, this is simply beautiful.

Let's finish with some example correspondences between degrees and radians, which can be seen in Table 1.1 below.

$[\circ]$	$[\text{rad}]$
0	0
30	$\frac{\pi}{6}$
45	$\frac{\pi}{4}$
60	$\frac{\pi}{3}$
90	$\frac{\pi}{2}$
120	$\frac{2\pi}{3}$
180	π
270	$\frac{3\pi}{2}$
360	2π

Table 1.1: Example correspondences between measurements in degrees and radians.

1.4.2 Taylor Series

For some functions, it is rather easy to calculate their values at some point x_0 : for example, given a polynomial function

$$P(x) = a_0 + a_1x + a_2x^2 + a_3x^3 + \cdots + a_nx^n, \quad (1.5)$$

it is rather easy to calculate its value at any real point x : all the operations that we need to use are addition and multiplication of real numbers, and raising real numbers to an integer power (which in principle can be implemented as repeated multiplications). Modern computers calculate such operations at the rate of billions of times a second.

Note 1.1 Floating point representation

For the sake of simplicity, I'm ignoring the entire topic of floating point numbers and relevant questions of precision.



Example 1.1 Calculating a value of a simple polynomial

Given the polynomial $P(x) = 3x^2 - 2x + 7$, we can easily calculate its value at, say, $x = 5$:

$$P(x = 5) = 3 \cdot 5^2 - 2 \cdot 5 + 7 = 3 \cdot 25 - 10 + 7 = 72.$$



For other functions, on the other hand, it is not that easy to calculate their values at most points. For example, consider the exponential function $f(x) = e^x$. We know precisely exactly one real value of the function: at $x = 0$, the function returns 1. But for any other value of x , we don't really know the value of the function. That is, we know all values *symbolically* (for example, $f(1) = e$, $f(2) = e^2$, etc.) - but not in explicit form.

However, given that the function behaves nicely enough (we'll discuss in a moment what that means), we can *approximate* its value to whatever precision we wish, using a method called the **Taylor series** of the function. The "price" we pay for greater precision is simply performing more calculations.

The basic idea of a Taylor series of a function $f(x)$ is that we approximate the function by adding higher and higher derivatives of the same function, at some point a for which we know precisely the value of the function and its derivatives to any order. This might sound a bit abstract, so let's explore this process using an example function - again, the exponential $f(x) = e^x$. As mentioned, we only really know one value of the function precisely, namely at $a = 0$: $e^0 = 1$.

We can therefore start approximating e^x simply as e^0 . This is obviously a very imprecise approximation, but the important thing

is that if we look at a very close neighborhood of $x = 0$, the approximation is not *that* bad: consider, for example $x_0 = 0.0001$. Our approximation gives $e^{x_0} \approx 1$, which is not far from the more precise value $e^{x_0} = 1.000100005$ (the value here is shown up to the ninth decimal). Of course, the closer we get to $a = 0$, the better our approximation gets: for example, with $x_1 = 0.00001$, we get $e^{x_1} \approx 1$ again, where in reality $e^{x_1} = 1.000010000$ (also shown here up to the ninth decimal). On the other hand, if we as we get farther away from $x = 0$, the approximation becomes worse and worse, as seen in Table 1.2 below.

x	e^x (exact)	Δ (error)
0.000	1.000000000	0.000000000
0.001	1.001000500	0.001000500
0.010	1.010050167	0.010050167
0.100	1.105170918	0.105170918
0.500	1.648721271	0.648721271
0.510	1.665291195	0.665291195
0.520	1.682027650	0.682027650
1.000	2.718281828	1.718281828
1.100	3.004166024	2.004166024
1.500	4.481689070	3.481689070
2.000	7.389056099	6.389056099

Table 1.2: Zero order Taylor series approximation of e^x .

Now, let's take this a step further: since the derivative of a function at a point tells us how the functions changes close to the point, we can use this information to improve our approximation by adding the value of the first derivative of e^x at $x = 0$, which is also 1. In fact, this is true for any order derivative of e^x , since $\frac{d^n}{dx^n} e^x = e^x$ for any $n \in \mathbb{N}$. Since the derivative changes with the value of x , we will multiply it by x in the approximation. Thus we get

$$e^x \approx 1 + 1 \cdot x = 1 + x, \quad (1.6)$$

which we call the **first order** approximation of the function e^x .

Table 1.3 below shows the same values from Table 1.2 for x , but using the first order approximation for e^x .

By examining Table 1.3 it's clear that $1 + x$ is a much better approximation for e^x than just 1. Of course, we can take a further step: the second derivative of a function at a point tells us how the change in the function itself changes around that point. We can take the second derivative of e^x (which is by itself e^x), again substitute $x = 0$, get $e^0 = 1$, and use this with x^2 to approximate e^x even better. In fact, let's already take this idea to its logical conclusion: we'll simply use the infinite power series in x , i.e. $1, x, x^2, x^3, \dots, x^n, \dots$ and

x	$1 + x$	e^x (exact)	Δ (error)
0.000	1.000000000	1.000000000	0.000000000
0.001	1.001000000	1.001000500	0.000000500
0.010	1.010000000	1.010050167	0.000050167
0.100	1.100000000	1.105170918	0.005170918
0.500	1.500000000	1.648721271	0.148721271
0.510	1.510000000	1.665291195	0.155291195
0.520	1.520000000	1.682027650	0.162027650
1.000	2.000000000	2.718281828	0.718281828
1.100	2.100000000	3.004166024	0.904166024
1.500	2.500000000	4.481689070	1.981689070
2.000	3.000000000	7.389056099	4.389056099

Table 1.3: First order Taylor series approximation of e^x .

attach to each x^k the value given by $\left. \frac{d^n}{dx^n} e^x \right|_{x=0} = 1$, so we get

$$e^x \approx 1 + x + x^2 + x^3 + \dots + x^n + \dots = \sum_{n=0}^{\infty} x^n. \quad (1.7)$$

This, however doesn't quite work. The coefficients of each term have to be adjusted. By assuming that at the limit where $n \rightarrow \infty$ the approximation should be with zero error, we get that for each n ,

$$\frac{d^n}{dx^n} e^x = \frac{d^n}{dx^n} (1 + x + x^2 + \dots) \equiv s(x). \quad (1.8)$$

Calculating the general n -th derivative of Equation 1.7 is rather easy. Let's examine the case for $n = 3$: the first three derivatives of Equation 1.7 are

$$\begin{aligned} \frac{d}{dx} s(x) &= 1 + 2x + 3x^2 + 4x^3 + \dots, \\ \frac{d^2}{dx^2} s(x) &= 2 + 6x + 12x^2 + \dots, \\ \frac{d^3}{dx^3} s(x) &= 6 + 24x + \dots, \end{aligned}$$

and if we substitute $x = 0$ we get... TBW!

The most general polynomial expansion of a smooth function $f(x)$ is an infinite polynomial in x :

$$P(x) = a_0 + a_1x + a_2x^2 + a_3x^3 + \dots \quad (1.9)$$

If $P(x) = f(x)$ then all their derivatives are the equal, i.e.

$$\begin{aligned}
 f'(x) &= P'(x) = a_1 + 2a_2x + 3a_3x^2 + 4a_4x^3 + \dots \\
 f''(x) &= P''(x) = 2a_2 + (2 \cdot 3)a_3x + (4 \cdot 3)a_4x^2 + \dots, \\
 f'''(x) &= P'''(x) = (2 \cdot 3)a_3 + (4 \cdot 3 \cdot 2)a_4x + \dots, \\
 f''''(x) &= P''''(x) = (4 \cdot 3 \cdot 2)a_4 + \dots, \\
 &\vdots \\
 f^{(n)}(x) &= P^{(n)}(x) = n!a_n + \dots, \\
 &\vdots
 \end{aligned}
 \tag{1.10}$$

Substituting $x = 0$ into the equality we get

$$\begin{aligned}
 f'(0) &= P'(0) = a_1 + \cancel{2a_2x} + \cancel{3a_3x^2} + \cancel{4a_4x^3} + \dots \\
 f''(0) &= P''(0) = 2a_2 + \cancel{(2 \cdot 3)a_3x} + \cancel{(4 \cdot 3)a_4x^2} + \dots, \\
 f'''(0) &= P'''(0) = (2 \cdot 3)a_3 + \cancel{(4 \cdot 3 \cdot 2)a_4x} + \dots, \\
 f''''(0) &= P''''(0) = (4 \cdot 3 \cdot 2)a_4 + \dots, \\
 &\vdots \\
 f^{(n)}(0) &= P^{(n)}(0) = n!a_n + \dots, \\
 &\vdots
 \end{aligned}
 \tag{1.11}$$

or more succinctly:

$$a_n = \frac{f^{(n)}(0)}{n!}.$$
(1.12)

Therefore, the full taylor polynomial for $f(x)$ is

$$\begin{aligned}
 P(x) &= f(0) + f'(0)x + \frac{f''(0)}{2}x^2 + \frac{f'''(0)}{3!}x^3 + \frac{f''''(0)}{4!}x^4 + \dots \\
 &= \sum_{n=0}^{\infty} \frac{f^{(n)}(0)}{n!} x^n.
 \end{aligned}
 \tag{1.13}$$

If we take only the first N terms in $P(x)$ we get

$$P_N(x) = \sum_{n=0}^N \frac{f^{(n)}(0)}{n!} x^n.$$
(1.14)

Note 1.2 $(N + 1)$ -th derivative an N -th partial Taylor sum

Since the N -th partial sum of any taylor series $P(x)$ is

$$P_N(x) = \sum_{n=0}^N \frac{f^{(n)}(0)}{n!} x^n,$$

all terms except $\frac{f^{(N)}(0)}{N!} x^N$ cancel out by the N -derivative. And at the $(N + 1)$ -derivative this term cancels out as well, since its a constant. Therefore from its $(N + 1)$ -th term and on, the partial sum $P_N(x)$ reduces to zero:

$$P^{(k>N)}(x) = 0.$$



Example 1.2 Taylor series of $\exp(x)$

Any order derivative of $\exp(x)$ is equal to $\exp(x)$ itself, i.e. in more mathematical notation:

$$\frac{d^n}{dx^n} \exp(x) = \exp(x), \quad \forall n \in \mathbb{N}.$$

Thus, we know the value at $x = 0$ of any order derivative of $\exp(x)$:

$$\left. \frac{d^n}{dx^n} \exp(x) \right|_{x=0} = \exp(0) = 1.$$

The Taylor series of $\exp(x)$ is therefore very easy to calculate around $x = 0$:

$$\begin{aligned} \exp(x) &= \sum_{k=0}^{\infty} \frac{\exp^{(k)}(0)}{k!} x^k \\ &= \frac{1}{0!} x^0 + \frac{1}{1!} x^1 + \frac{1}{2!} x^2 + \frac{1}{3!} x^3 + \dots \\ &= 1 + x + \frac{1}{2} x^2 + \frac{1}{3!} x^3 + \frac{1}{4!} x^4 + \dots \end{aligned}$$



Example 1.3 Taylor series of $\sin x$ and $\cos(x)$

The derivatives of both $\sin(x)$ and $\cos(x)$ are periodic:

$$\begin{array}{ccc} \cos(x) & \xrightarrow{\frac{d}{dx}} & -\sin(x) \\ \uparrow \frac{d}{dx} & & \downarrow \frac{d}{dx} \\ \sin(x) & \xleftarrow{\frac{d}{dx}} & -\cos(x) \end{array}$$

And here it is summed as a table:

n	$\frac{d^n}{dx^n} \sin(x)$	$\frac{d^n}{dx^n} \cos(x)$
0	$\sin(x)$	$\cos(x)$
1	$\cos(x)$	$-\sin(x)$
2	$-\sin(x)$	$-\cos(x)$
3	$-\cos(x)$	$\sin(x)$
4	$\sin(x)$	$\cos(x)$
5	$\cos(x)$	$-\sin(x)$
	\vdots	

And since

$$\sin(0) = -\sin(0) = 0,$$

$$\cos(0) = 1,$$

$$-\cos(0) = -1,$$

we can easily derive the Taylor series for both functions

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around $x = 0$:

$$\begin{aligned}
 \sin(x) &= \sum_{k=0}^{\infty} \frac{\sin^{(k)}(0)}{k!} x^k \\
 &= \frac{\sin(0)}{0!} x^0 + \frac{\cos(0)}{1!} x^1 + \frac{-\sin(0)}{2!} x^2 + \frac{-\cos(0)}{3!} x^3 + \dots \\
 &= 0 + x + 0 - x^3 + 0 + x^5 + 0 - x^7 + 0 + \dots \\
 &= x - x^3 + x^5 - x^7 + \dots \\
 \cos(x) &= \sum_{k=0}^{\infty} \frac{\cos^{(k)}(0)}{k!} x^k \\
 &= \frac{\cos(0)}{0!} x^0 + \frac{-\sin(0)}{1!} x^1 + \frac{-\cos(0)}{2!} x^2 + \frac{\sin(0)}{3!} x^3 + \dots \\
 &= 1 + 0 - x^2 + 0 + x^4 + 0 - x^6 + \dots \\
 &= 1 - x^2 + x^4 - x^6 + \dots
 \end{aligned}$$



Looking at the above results for $\exp(x)$, $\sin(x)$ and $\cos(x)$, one can derive a nice connection: notice that

$$i^0 = 1, i^1 = i, i^2 = -1, i^3 = -i, i^4 = 1, i^5 = i, \dots \quad (1.15)$$

and if we plug ix into the Taylor expansion of $\exp(x)$ we get

$$\begin{aligned}
 \exp(ix) &= 1 + ix + \frac{1}{2!}(ix)^2 + \frac{1}{3!}ix^3 + \frac{1}{4!}ix^4 + \frac{1}{5!}ix^5 + \dots \\
 &= i^0 + i^1x + \frac{1}{2!}i^2x^2 + \frac{1}{3!}i^3x^3 + \frac{1}{4!}i^4x^4 + \frac{1}{5!}i^5x^5 + \dots \\
 &= 1 + ix - \frac{1}{2!}x^2 - \frac{1}{3!}ix^3 + \frac{1}{4!}x^4 + \frac{1}{5!}ix^5 + \dots \\
 &= 1 + \left(-\frac{1}{2!}x^2 + \frac{1}{4!}x^4 + \dots\right) + ix - i\left(\frac{1}{3!}x^3 - \frac{1}{5!}x^5 + \dots\right) \\
 &= \left(1 - \frac{1}{2!}x^2 + \frac{1}{4!}x^4 + \dots\right) + i\left(x - \frac{1}{3!}x^3 + \frac{1}{5!}x^5 + \dots\right) \\
 &= \cos(x) + i\sin(x).
 \end{aligned} \quad (1.16)$$

Since in practice we can't use the full Taylor series to approximate a function at a point, we must intelligently choose how many terms we want to use to get a "good enough" approximation - that is, we want that the difference between our polynomial approximation is the actual function to be no more than some value Δ :

$$|f(a) - P_n(a)| \leq \Delta. \quad (1.17)$$

We can construct a **remainder function** which gives the difference between the actual function value and our n -th degree approximation at every point on the interval $I = [0, a]$:

$$R_n(x) = f(x) - P_n(x). \quad (1.18)$$

Thanks to Lagrange and other great mathematicians, the following always holds for the remainder function: if the the $n + 1$ -th derivative of the original function is bounded on the open interval $(0, a)$, i.e.

$$|f^{(n+1)}(x)| \leq M, \quad (1.19)$$

then there's a finite value M such that

$$|R_n(x)| \leq \left| \frac{Mx^{n+1}}{(n+1)!} \right|. \quad (1.20)$$

This is called the **Lagrange error bound**. No proof is provided for it in this course (TBW: add reference to one?).

Example 1.4 Lagrange error bound

Let's calculate what is the minimum n needed to calculate $\cos(0.25)$ to within $\Delta = 10^{-3} = 0.001$ using a Taylor series: the first four derivatives of $\cos(x)$ are

$$\begin{array}{ccc} \cos(x) & \xrightarrow{\frac{d}{dx}} & -\sin(x) \\ \uparrow \frac{d}{dx} & & \downarrow \frac{d}{dx} \\ \sin(x) & \xleftarrow{\frac{d}{dx}} & -\cos(x) \end{array}$$

All of these functions are bounded to the interval $[-1, 1]$, and so their absolute value is bounded by 1. We can replace M in Equation 1.20 by 1 and we get

$$|R_n(0.25)| \leq \left| \frac{0.25^{n+1}}{(n+1)!} \right|.$$

It's rather difficult to isolate n from this relation, but we can simply substitute different values to it until we get a result smaller than 10^{-3} :

n	$ R_n $
0	$\frac{1}{4^1 \cdot 1!} = 0.25$
1	$\frac{1}{4^2 \cdot 2!} = 0.03125$
2	$\frac{1}{4^3 \cdot 3!} \approx 0.0026$
3	$\frac{1}{4^4 \cdot 4!} \approx 0.0001627$

Meaning that we need at least the $n = 3$ term to reach our required precision.



1.4.3 Projections, rejections and the dot product

Every vector $\mathbf{a} \in \mathbb{R}^n$ can be decomposed into two components: one in the direction of another vector \mathbf{b} , and one orthogonal to \mathbf{b} . These components are called, respectively, the **projection** of \mathbf{a} onto \mathbf{b} and the **rejection** of \mathbf{a} from \mathbf{b} . In \mathbb{R}^3 the same procedure can be

applied to projecting a vector onto- and rejecting it from a plane, since every plane in \mathbb{R}^3 has a single normal vector up to a sign (Figure 1.3).

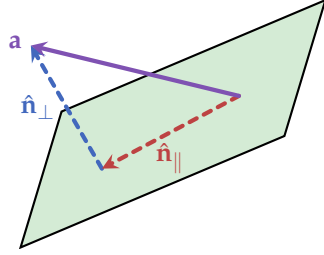


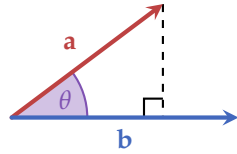
Figure 1.3: Decomposing the vector \mathbf{a} into two components in respect to a plane: one parallel to the plane (\mathbf{a}_{\parallel}) and one orthogonal to it (\mathbf{a}_{\perp}). These are also called the projection and rejection of \mathbf{a} on the plane, respectively.

The projection of a vector onto another vector gives rise to an important operation between two vectors: the **dot product**: given a vector \mathbf{a} , its projection on the vector \mathbf{b} is

$$\text{proj}_{\mathbf{b}} \mathbf{a} = |\mathbf{a}| \cos(\theta), \quad (1.21)$$

where θ is the angle between the vectors. See Figure 1.4 for a visual representation. We then define the dot product between the two vectors as

$$\langle \mathbf{a}, \mathbf{b} \rangle = |\mathbf{b}| \text{proj}_{\mathbf{b}} \mathbf{a} = |\mathbf{b}| |\mathbf{a}| \cos(\theta). \quad (1.22)$$



It is of course convenient to have a way to calculate the dot product component-wise. To find such form, we use the same two vectors \mathbf{a} and \mathbf{b} from before, and their difference $\mathbf{c} = \mathbf{a} - \mathbf{b}$ (Figure 1.5).

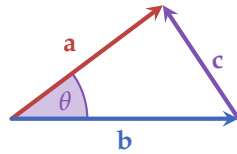


Figure 1.4: Projection of \mathbf{a} onto \mathbf{b} : no matter how many dimensions we use, we can always rotate our view such that we look at the plane spanned by both vectors, and \mathbf{b} lies horizontally. In this way it's easy to see why the projection of \mathbf{a} onto \mathbf{b} is $|\mathbf{a}| \cos(\theta)$: it's simply the definition of the cosine function ("side next to the angle divided by the hypotenuse").

Thanks to trigonometry we know that the following relation holds ("the law of cosines"):

$$\begin{aligned} |\mathbf{c}|^2 &= |\mathbf{a}|^2 + |\mathbf{b}|^2 - 2|\mathbf{a}||\mathbf{b}|\cos(\theta) \\ &= |\mathbf{a}|^2 + |\mathbf{b}|^2 - 2\langle \mathbf{a}, \mathbf{b} \rangle, \end{aligned} \quad (1.23)$$

which we can rearrange into

$$\langle \mathbf{a}, \mathbf{b} \rangle = \frac{1}{2} (|\mathbf{a}|^2 + |\mathbf{b}|^2 - |\mathbf{c}|^2). \quad (1.24)$$

In \mathbb{R}^2 the component version of the vectors are

$$\mathbf{a} = \begin{bmatrix} a_x \\ a_y \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} b_x \\ b_y \end{bmatrix}, \quad \mathbf{c} = \begin{bmatrix} c_x \\ c_y \end{bmatrix} = \begin{bmatrix} a_x - b_x \\ a_y - b_y \end{bmatrix}, \quad (1.25)$$

and Equation 1.24 becomes

$$\begin{aligned}
 \langle \mathbf{a}, \mathbf{b} \rangle &= \frac{1}{2} \left(a_x^2 + a_y^2 + b_x^2 + b_y^2 - (a_x - b_x)^2 - (a_y - b_y)^2 \right) \\
 &= \frac{1}{2} \left(\cancel{a_x^2} + \cancel{a_y^2} + \cancel{b_x^2} + \cancel{b_y^2} - \cancel{a_x^2} + 2a_x b_x - \cancel{b_x^2} - \cancel{a_y^2} + 2a_y b_y + \cancel{b_y^2} \right) \\
 &= a_x b_x + a_y b_y.
 \end{aligned} \tag{1.26}$$

In general, we can apply the same procedure in \mathbb{R}^n : the components of the vectors are

$$\mathbf{a} = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}, \quad \mathbf{c} = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix} = \begin{bmatrix} a_1 - b_1 \\ a_2 - b_2 \\ \vdots \\ a_n - b_n \end{bmatrix}, \tag{1.27}$$

and Equation 1.24 becomes

$$\begin{aligned}
 \langle \mathbf{a}, \mathbf{b} \rangle &= \frac{1}{2} \left(a_1^2 + a_2^2 + \dots + a_n^2 + b_1^2 + b_2^2 + \dots + b_n^2 - (a_1 - b_1)^2 \right. \\
 &\quad \left. - (a_2 - b_2)^2 - \dots - (a_n - b_n)^2 \right) \\
 &= \frac{1}{2} \left(\cancel{a_1^2} + \cancel{a_2^2} + \dots + \cancel{a_n^2} + \cancel{b_1^2} + \cancel{b_2^2} + \dots + \cancel{b_n^2} - \cancel{a_1^2} + 2a_1 b_1 - \right. \\
 &\quad \left. \cancel{b_1^2} - \cancel{a_2^2} + 2a_2 b_2 + \cancel{b_2^2} + \dots + \cancel{a_n^2} + 2a_n b_n + \cancel{b_n^2} \right) \\
 &= a_1 b_1 + a_2 b_2 + \dots + a_n b_n \\
 &= \sum_{i=1}^n a_i b_i.
 \end{aligned} \tag{1.28}$$

Let us now explore some key properties of the dot product (the reader is encouraged to prove items 2 and 3):

1. Since the product of real numbers is commutative (i.e. $ab = ba$), the dot product is also commutative: if we exchange v_i and w_i in Equation 1.28 the result stays the same.
2. The dot product is distributive over vector addition: $\langle \mathbf{a}, \mathbf{b} + \mathbf{c} \rangle = \langle \mathbf{a}, \mathbf{b} \rangle + \langle \mathbf{a}, \mathbf{c} \rangle$.
3. We can take out scalars from within the product: given the scalars ϵ_1 and ϵ_2 , $\langle \epsilon_1 \mathbf{a}, \epsilon_2 \mathbf{b} \rangle = \epsilon_1 \epsilon_2 \langle \mathbf{a}, \mathbf{b} \rangle$.
4. The angle between two non-zero orthogonal vectors \mathbf{a} and \mathbf{b} is $\theta = \frac{\pi}{2}$, and therefore $\langle \mathbf{a}, \mathbf{b} \rangle = |\mathbf{a}| |\mathbf{b}| \cos\left(\frac{\pi}{2}\right) = 0$ (since $\cos\left(\frac{\pi}{2}\right) = 0$). This works in the other way around as well: if the dot product of two non-zero vectors is zero, then they are orthogonal.

5. We can't cancel vectors in the dot product the same we do with real numbers: consider three vectors $\mathbf{a} \neq \mathbf{0}$, \mathbf{b} , \mathbf{c} such that $\langle \mathbf{a}, \mathbf{b} \rangle = \langle \mathbf{a}, \mathbf{c} \rangle$. We can redistribute the equation to give $\langle \mathbf{a}, \mathbf{b} - \mathbf{c} \rangle = 0$, which means that \mathbf{a} and $\mathbf{b} - \mathbf{c}$ are orthogonal, and doesn't imply that $\mathbf{b} - \mathbf{c} = \mathbf{0}$ (i.e. \mathbf{b} and \mathbf{c} might not be equal).

2

Mechanics

2.1 Preface

Text text text

2.2 Pendulum

2.2.1 Simple pendulum

Text text text.

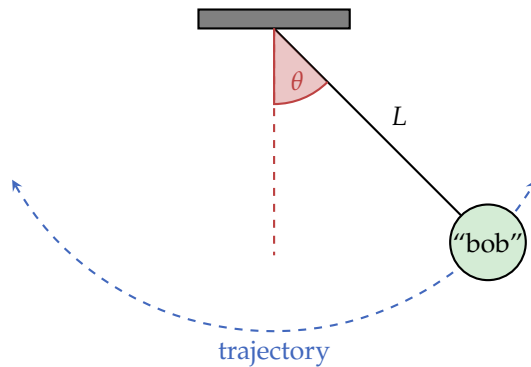


Figure 2.1: A simple pendulum. TBW: add more info.

Using force analysis we can derive an equation of motion for the bob (see Figure 2.2): since the rod can't change its length (it's always L), the only variable quantity is the angle θ , and the bob's trajectory is a circle. Any force acting in a radial direction to the trajectory must be counter-balanced (otherwise there will be some acceleration - and therefore motion - in that direction). We are therefore left with only a tangential force, with magnitude

$$F = -mg \sin(\theta), \quad (2.1)$$

the minus sign here is chosen to represent that gravity acts in the negative y direction (i.e. "down").

Applying Newton's second law we get that

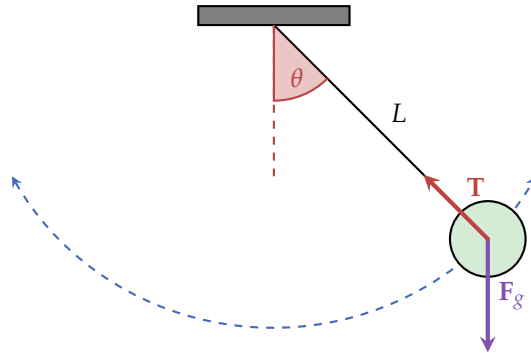
$$F = ma = -mg \sin(\theta), \quad (2.2)$$

i.e.

$$a = -g \sin(\theta). \quad (2.3)$$

Now we see that the minus sign also makes sense physically, as it shows that the acceleration is always in the opposite direction to the angle (which is negative to the left and positive to the right).

Figure 2.2: Forces acting on a simple pendulum. TBW: Force compnents.



The tangential position s of the bob can be calculated from the angle θ by

$$s = L\theta \quad (2.4)$$

(recall that θ is given in radians), and therefore the tangential velocity is

$$v = \dot{s} = L\dot{\theta}, \quad (2.5)$$

and the acceleration is therefore

$$a = \dot{v} = \ddot{s} = L\ddot{\theta}. \quad (2.6)$$

Since we know that $a = -g \sin(\theta)$, we get

$$L\ddot{\theta} = -g \sin(\theta), \quad (2.7)$$

and by moving the rhs term to the left and divide by l we get

$$\ddot{\theta} + \frac{g}{L} \sin(\theta) = 0. \quad (2.8)$$

This is a differential equation without analytical solution. We will therefore take two approaches: (1) use an approximation to yield an analytical solution, and (2) solve the equation numerically.

2.2.2 Small-angle approximation

The Taylor series expansion of $\sin(x)$ around $x = 0$ is

$$\sin(x) = x - \frac{1}{3!}x^3 + \frac{1}{5!}x^5 - \frac{1}{7!}x^7 + \dots \quad (2.9)$$

We can therefore approximate $\sin(x)$ as x for small values of x :

$$\sin(x) \approx x. \quad (2.10)$$

This is known as the “small-angle approximation” of the sine function. By using this approximation we get that the (analytically) unsolvable Equation 2.8 reduces to

$$\ddot{\theta} + \frac{g}{L}\theta = 0, \quad (2.11)$$

for which we have an exact solution:

$$\theta(t) = A \cos(\omega t + \phi), \quad (2.12)$$

where $\omega = \sqrt{\frac{g}{L}}$. The parameters A and ϕ depend on the initial conditions (i.e. angle and tangential velocity).

TBW: resonance freq, how it looks like (+phase space), python plot?

2.2.3 Numerical integration

Another approach to solve Equation 2.8 is doing so numerically, i.e. essentially running a computer simulation. While computers can carry out many calculations per second (in the order of billions, in fact) - they are limited to performing discrete calculations. That is to say, a numerical calculation is also an approximation. However, unlike the small-angle approximation, in principle we can improve the approximation indefinitely, although in practice this is of course impossible.

Let us use a rather naive approach to numerically approximating Equation 2.8: instead of viewing θ as a continuous function of time t , we instead define t to only have equally spaced discrete values t_0, t_1, t_2, \dots such that

$$t_n = t_0 + n\Delta t, \quad (2.13)$$

or phrased differently: we look at the values t_i of t at intervals Δt starting from t_0 (see Figure 2.3).

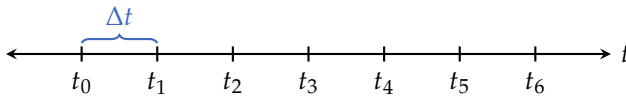


Figure 2.3: Discrete time steps.

In this discrete view, a function of t can also only be evaluated at discrete points. For example, the angle θ of the bob as a function of time is a discrete function taking the values $\theta_0, \theta_1, \theta_2, \dots$, where each θ_i corresponds to the time t_i .

How does the discrete angular velocity look like in this scheme? It's worth looking at the definition of angular velocity in the continuous case:

$$\omega(t) = \dot{\theta}(t) = \frac{d\theta(t)}{dt} = \lim_{\Delta t \rightarrow 0} \frac{\theta(t + \Delta t) - \theta(t)}{\Delta t}. \quad (2.14)$$

To discretize, we can replace $\theta(t)$ with $\theta_{t_i} = \theta_i$, and therefore $\theta(t + \Delta t)$ with $\theta_{t_i + \Delta t} = \theta_{i+1}$. The limit $\lim_{\Delta t \rightarrow 0}$ is pretty meaningless, since the smallest Δt possible in the discrete case is the Δt we chose to discretize out time steps. Therefore, we get that the discrete version of $\omega(t)$ is

$$\omega_i = \frac{\theta_{i+1} - \theta_i}{\Delta t}, \quad (2.15)$$

which is nothing more than saying that the angular velocity at time t_i is simply the difference between the angle at time t_{i+1} and the angle at time t_i , divided by the time difference Δt .

Equation 2.15 gives us a powerful tool: if we only know the current angle θ_i of the bob and its current velocity ω_i , then its next position θ_{i+1} is a simple rearrangement of the equation:

$$\theta_{i+1} = \theta_i + \omega_i \Delta t. \quad (2.16)$$

The same logic can be applied to the angular acceleration $\alpha(t)$: in the continuous version its given by

$$\alpha(t) = \dot{\omega}(t) = \frac{d\omega(t)}{dt} = \lim_{\Delta t \rightarrow 0} \frac{\omega(t + \Delta t) - \omega(t)}{\Delta t}, \quad (2.17)$$

and thus the discrete version is

$$\alpha_i = \frac{\omega_{i+1} - \omega_i}{\Delta t}, \quad (2.18)$$

and recovering the angular velocity ω_{i+1} from the angular velocity ω_i and the angular acceleration α_i is done by

$$\omega_{i+1} = \omega_i + \alpha_i \Delta t. \quad (2.19)$$

Since we use the inverse of the differentiation operation (in the discrete sense) to recover the quantity we're after, this scheme is known as a **numerical integration**. Specifically, this kind of numerical integration is called the **Forward-Euler** method, and it is generally considered unfavourable due to the fast rate with which it gains errors, and is rarely used in practice.

However, for the sake of simplicity of writing our first simulations, we will not discuss these issues now, nor will we generalize the method and present better ones - both of which we will do later in the course. Instead, for now we will continue with using this method to devise a numerical integration scheme for a simple pendulum.

In the case of the pendulum, recall that the acceleration at time t is $\ddot{\theta}(t) = -\frac{g}{L} \sin(\theta)$ (Equation 2.8), and therefore we can discretize it as

$$\alpha_{i+1} = -\frac{g}{L} \sin(\theta_i). \quad (2.20)$$

We then use the forward Euler method to get the angular velocity and angle, as seen in Equation 2.19 and Equation 2.16, respectively.

2.2.4 Damped oscillation

A slightly more realistic model of a pendulum also considers the way the pendulum loses energy over time (e.g. via friction). This can be modelled by adding a force which resists the angular velocity (see also Figure 2.4):

$$F_d = -bL\dot{\theta}, \quad (2.21)$$

where b is simply a parameter which adjusts how strong the damping is.

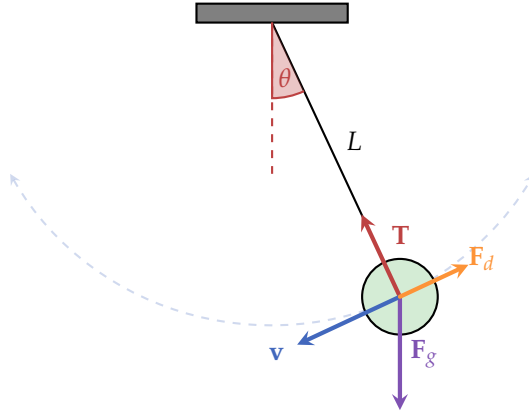


Figure 2.4: Forces on a pendulum including a damping force.

Using Newton's second law, recalling that $a = L\ddot{\theta}$ (Equation 2.6), we get

$$mL\ddot{\theta} = F_g + F_d = -mg \sin(\theta) - bL\dot{\theta}. \quad (2.22)$$

It's common to use $\beta = \frac{b}{2m}$ and of course $\omega_0 = \sqrt{\frac{g}{L}}$, which together with some rearrangement gives us

$$\ddot{\theta} = -2\beta\dot{\theta} - \omega_0^2 \sin(\theta). \quad (2.23)$$

2.2.5 Double pendulum

An interesting system arises when we take a simple pendulum and attach another simple pendulum to it (as seen in Figure 2.5): unlike a simple pendulum, this system exhibits very complex dynamics which is sensitive to the initial conditions - i.e. it is **chaotic**, meaning that small changes in the initial conditions evolve into big differences in the dynamics of the system.

To analyze the behaviour of the system we again seek to find a differential equation which will describe the system completely. Unfortunately, this differential equation is rather complicated and is composed of two coupled differential equations. We will only present the equation here but not how to derive it - this will be added in an appendix.

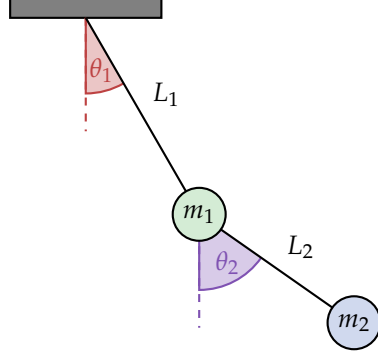


Figure 2.5: A double pendulum. TBW: add more info.

The differential equations governing the dynamics of the system are:

$$0 = (m_1 + m_2) L_1 \ddot{\theta}_1 + m_2 L_2 \ddot{\theta}_2 \cos(\theta_1 - \theta_2) + m_2 L_2 \dot{\theta}_2^2 \sin(\theta_1 - \theta_2) + (m_1 + m_2) g \sin(\theta_1), \quad (2.24)$$

and

$$0 = L_2 \ddot{\theta}_2 + L_1 \ddot{\theta}_1 \cos(\theta_1 - \theta_2) - L_1 \dot{\theta}_1^2 \sin(\theta_1 - \theta_2) + g \sin(\theta_2). \quad (2.25)$$

To help us get explicit expressions for $\ddot{\theta}_1$ and $\ddot{\theta}_2$ we can define the following quantities:

$$\alpha = \frac{m_2}{m_1}, \quad \beta = \frac{L_2}{L_1}, \quad \gamma = \frac{g}{L_1}, \quad \Delta\theta = \theta_1 - \theta_2. \quad (2.26)$$

we then get the following "beautiful" expressions for $\ddot{\theta}_1$ and $\ddot{\theta}_2$:

$$\begin{aligned} \ddot{\theta}_1 &= - \frac{(1 + \alpha) \gamma \sin(\theta_1) + \alpha \beta \dot{\theta}_2^2 \sin(\Delta\theta) + \alpha \cos(\Delta\theta) \left[\dot{\theta}_1^2 \sin(\Delta\theta) - \gamma \sin(\theta_2) \right]}{1 + \alpha \sin^2(\Delta\theta)}, \\ \ddot{\theta}_2 &= \frac{(1 + \alpha) \left(\dot{\theta}_1^2 \sin(\Delta\theta) - \gamma \sin(\theta_2) + \cos(\Delta\theta) \left[(1 + \alpha) \gamma \sin(\theta_1) + \alpha \beta \dot{\theta}_2^2 \sin(\Delta\theta) \right] \right)}{\beta \left[1 + \alpha \sin^2(\Delta\theta) \right]} \end{aligned} \quad (2.27)$$

Discretizing Equation 2.27 can then be done as following: first,

1. α , β and γ are defined as in Equation 2.26.
2. Set $\Delta\theta_k = \theta_{1,k} - \theta_{2,k}$.
3. For compactness, we set $N_k = 1 + \alpha \sin^2(\Delta\theta_k)$, $S(x) = \sin(x)$, $C(x) = \cos(x)$ and $A = 1 + \alpha$.

Then, we get

$$\begin{aligned}\ddot{\theta}_{1,i} &= -\frac{A\gamma S(\theta_{1,i-1}) + \alpha\beta\dot{\theta}_{2,i-1}^2 S(\Delta\theta_{i-1}) + \alpha C(\Delta\theta_{i-1}) \left[\dot{\theta}_{1,i-1}^2 S(\Delta\theta_{i-1}) - \gamma S(\theta_{2,i-1}) \right]}{N_{i-1}}, \\ \ddot{\theta}_{2,i} &= \frac{A \left(\dot{\theta}_{1,i-1}^2 S(\Delta\theta_{i-1}) - \gamma S(\theta_{2,i-1}) + C(\Delta\theta_{i-1}) \left[A\gamma S(\theta_{1,i-1}) + \alpha\beta\dot{\theta}_{2,i-1}^2 S(\Delta\theta_{i-1}) \right] \right)}{\beta N_{i-1}}.\end{aligned}\quad (2.28)$$

(lovely, isn't it?)

2.3 Numerical Integration Methods

TBW: explanation for all derivation, etc. At the moment it's just the equations (for use by me during the lectures).

2.3.1 Forwards (Explicit) Euler

Substituting $a = t_n$ and $x - a = \Delta t$ into the Taylor expansion of $x(t)$, we get

$$x_{n+1} = x(t_n + \Delta t) = \dot{x}(t_n) \Delta t + \frac{1}{2} \ddot{x}(t_n) \Delta t^2 + \dots \quad (2.29)$$

We truncate the expansion after the linear term, and get:

$$x_{n+1} = x_n + \dot{x}_n \Delta t + \mathcal{O}(\Delta t^2). \quad (2.30)$$

Local (truncation) error: $\mathcal{O}(\Delta t^2)$. Global error: if reduce Δt by α , we would need α times the amount of steps, hence the global error is $\mathcal{O}(\Delta t)$.

2.3.2 Backwards (Implicit) Euler

2.3.3 Modified Euler (2nd-order Runge-Kutta)

2.3.4 4th-order Runge-Kutta

2.3.5 Other Runge-Kutta

2.4 Orbital Mechanics

2.4.1 Classic Gravitational Force

Already in the 17th century, *Isaac Newton* formulated the gravitational force existing between any two objects with masses greater than zero. The strength of the force is given by the equation

$$F = G \frac{m_1 m_2}{r^2}, \quad (2.31)$$

where m_1 and m_2 are the respective masses of the two objects, r is the distance between them, and G is the *universal gravitational constant*,

$$G = (6.6743 \pm 0.0015) \times 10^{-11} \left[\text{N m}^2 \text{ kg}^{-2} \right] \quad (2.32)$$

The direction of the force is the line connecting the centers of mass of the two objects. Due to Newton's third law, the forces acting on the two objects are equal and opposite: the force applied by m_1 on m_2 , $F_{1 \rightarrow 2}$, is pointing **from** m_2 **onto** m_1 , and the force applied by m_2 on m_1 , $F_{2 \rightarrow 1}$ is pointing **from** m_1 **onto** m_2 - and is exactly opposite to $F_{1 \rightarrow 2}$, i.e. in vector notation

$$\mathbf{F}_{1 \rightarrow 2} = -\mathbf{F}_{2 \rightarrow 1}. \quad (2.33)$$

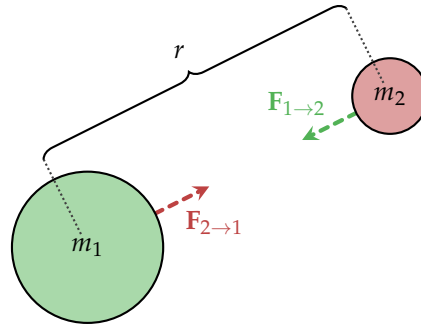
If the two objects have positions \mathbf{r}_1 and \mathbf{r}_2 , the vector pointing from object 1 to object 2 is

$$\mathbf{r}_{1 \rightarrow 2} = \mathbf{r}_2 - \mathbf{r}_1, \quad (2.34)$$

with the vector pointing from object 2 to object 1 having the exact opposite components, i.e. $\mathbf{r}_{2 \rightarrow 1} = -\mathbf{r}_{1 \rightarrow 2}$. The norms of $\mathbf{r}_{1 \rightarrow 2}$ and $\mathbf{r}_{2 \rightarrow 1}$ are simply r (the distance between the objects), and their directions are the unit vectors in the direction of $\mathbf{r}_{1 \rightarrow 2}$ and $\mathbf{r}_{2 \rightarrow 1}$, respectively:

$$\begin{aligned} \hat{\mathbf{r}}_{1 \rightarrow 2} &= \frac{\mathbf{r}_{1 \rightarrow 2}}{|\mathbf{r}_{1 \rightarrow 2}|} = \frac{\mathbf{r}_{1 \rightarrow 2}}{r}, \\ \hat{\mathbf{r}}_{2 \rightarrow 1} &= \frac{\mathbf{r}_{2 \rightarrow 1}}{|\mathbf{r}_{2 \rightarrow 1}|} = \frac{\mathbf{r}_{2 \rightarrow 1}}{r} = -\hat{\mathbf{r}}_{1 \rightarrow 2}. \end{aligned} \quad (2.35)$$

Figure 2.6: Gravitational force between two objects with masses m_1 and m_2 . Each object applies an attractive force on the other object, with norm $F = G \frac{m_1 m_2}{r^2}$ (where r is the distance between the objects) and in the direction pointing from each object to the other object.



In total, the vector notation of the gravitational force applied by the objects on each other are

$$\begin{aligned} \mathbf{F}_{1 \rightarrow 2} &= G m_1 m_2 \frac{\hat{\mathbf{r}}_{1 \rightarrow 2}}{r^2}, \\ \mathbf{F}_{2 \rightarrow 1} &= G m_1 m_2 \frac{\hat{\mathbf{r}}_{2 \rightarrow 1}}{r^2} = -\mathbf{F}_{1 \rightarrow 2}. \end{aligned} \quad (2.36)$$

Note 2.1 Another gravity force vector notation

In some textbooks, Equation 2.36 are written without the unit vectors $\hat{\mathbf{r}}_{1 \rightarrow 2}$ and $\hat{\mathbf{r}}_{2 \rightarrow 1}$, instead using the distance vectors and dividing by r^3 , i.e.

$$\begin{aligned} \mathbf{F}_{1 \rightarrow 2} &= G m_1 m_2 \frac{\mathbf{r}_{1 \rightarrow 2}}{r^3}, \\ \mathbf{F}_{2 \rightarrow 1} &= G m_2 m_1 \frac{\mathbf{r}_{2 \rightarrow 1}}{r^3}. \end{aligned}$$

The result is of course the same as in Equation 2.36, since for

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any non zero vector \mathbf{v} ,

$$\frac{\mathbf{v}}{|\mathbf{v}|^3} = \frac{1}{|\mathbf{v}|^2} \frac{\mathbf{v}}{|\mathbf{v}|} = \frac{1}{|\mathbf{v}|^2} \hat{\mathbf{v}}.$$



Let us look at an example of calculating the gravitational forces between two objects.

Example 2.1 Calculating a gravitational force

Let us calculate the gravitational forces between two objects A and B , using the following parameters:

$$\mathbf{r}_A = \begin{bmatrix} 1 \\ -2 \\ 0 \end{bmatrix}, m_A = 1,$$

$$\mathbf{r}_B = \begin{bmatrix} 2 \\ 5 \\ -2 \end{bmatrix}, m_B = 2.$$

For the sake of simplicity, we use $G = 1$ and don't consider units with this example.

The vector pointing from A to B is

$$\mathbf{r}_{A \rightarrow B} = \mathbf{B} - \mathbf{A} = \begin{bmatrix} 2 \\ 5 \\ -2 \end{bmatrix} - \begin{bmatrix} 1 \\ -2 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 7 \\ -2 \end{bmatrix},$$

and the vector pointing from B to A is

$$\mathbf{r}_{B \rightarrow A} = -\mathbf{r}_{A \rightarrow B} = \begin{bmatrix} -1 \\ -7 \\ 2 \end{bmatrix}.$$

The distance r between the objects is the norm of either of the above vectors, so we'll use $\mathbf{r}_{A \rightarrow B}$:

$$r = |\mathbf{r}_{A \rightarrow B}| = \sqrt{1^2 + 7^2 + 2^2} = \sqrt{19} \approx 7.3485.$$

The direction vectors are therefore

$$\hat{\mathbf{r}}_{A \rightarrow B} = \frac{1}{7.3485} \begin{bmatrix} 1 \\ 7 \\ -2 \end{bmatrix} = \begin{bmatrix} 0.1361 \\ 0.9526 \\ -0.2722 \end{bmatrix},$$

$$\hat{\mathbf{r}}_{B \rightarrow A} = -\hat{\mathbf{r}}_{A \rightarrow B} = \begin{bmatrix} -0.1361 \\ -0.9526 \\ 0.2722 \end{bmatrix}.$$

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The gravity force which A applies onto B is then

$$\mathbf{F}_{A \rightarrow B} = \cancel{G} \frac{\overbrace{m_1 m_2}^{=2 \times 1}}{r^2} \hat{\mathbf{r}}_{A \rightarrow B} = \frac{2}{54} \begin{bmatrix} 0.1361 \\ 0.9526 \\ -0.2722 \end{bmatrix} = \begin{bmatrix} 0.0050 \\ 0.0353 \\ -0.101 \end{bmatrix}.$$

and similarly,

$$\mathbf{F}_{B \rightarrow A} = -\mathbf{F}_{A \rightarrow B} = \begin{bmatrix} -0.0050 \\ -0.0353 \\ 0.101 \end{bmatrix}.$$

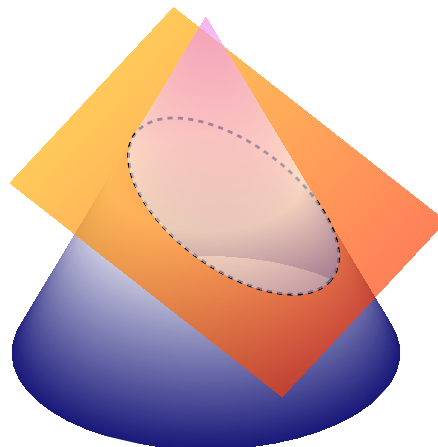
In the case where we only consider two objects, and choose our frame of reference such that one of the objects is stationary - an analytical solution to the spatial trajectory taken by the second object is known and well studied. It is called a **Keplerian orbit**, and it always takes the form of a conic section. Let us take a short detour to discuss conic sections.

2.4.2 Conic Sections

A conic section (sometimes simply just called “a conic”) is a 2-dimensional shape resulting from the intersection of a plane and a cone (see Figure 2.7). Depending on the angle α by which the plane intersects the cone relative to the cone’s side, the resulting shape can be one of 3 general types (here θ is the cone’s angle):

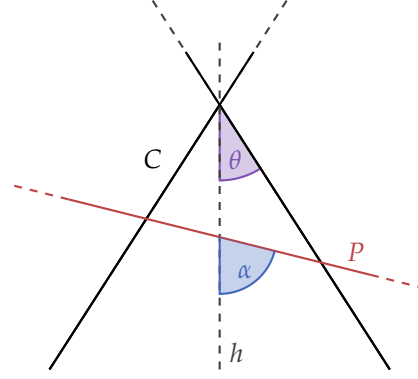
1. If $\alpha > \theta$ the intersection is an **ellipse**. If in addition $\alpha = 90^\circ$ the ellipse becomes a **circle**.
2. If $\alpha = \theta$ the intersection is a **parabola**.
3. If $\alpha < \theta$ the intersection is a **hyperbola**.

Figure 2.7: An intersection of a cone and a plane. Both the cone and plane are infinite - the cone extends infinitely “down”, but also has a second “inverted” part on the top, also extending to infinity. In the case here shown, the intersection is an ellipse. Image reproduced with modifications from !SOURCE!



Depending on the exact parameters of both C and P , the resulting

Figure 2.8: Side view of an infinite cone C and an infinite plane P intersecting it. The angle between P and the cone's height line h is α , and the angle between the cone's surface and h is θ . In this figure $0 < \theta < \alpha < 90^\circ$, and thus the shape formed by the intersection of C and P is an



ellipse. conic section can be **degenerate** - either a point, a line or two intersection lines. This happens if P goes through the vertex point of C : if $\alpha = 90^\circ$ the result is a single point ¹, if $\alpha = \theta$ the result is a single line, and if $\alpha > \theta$ the result is two intersecting lines.

¹ one can understand this as being a circle with radius $r = 0$.

2.4.3 Geometric Properties of Conic Sections

Of the non-degenerate conic sections, the ellipse is the only closed curve. Both the parabola and hyperbola are open: in essence, this means that they diverge to infinity. A common geometric definition for all conic sections is the following: given a line L (called the **directrix**) and a point F (called the **locus**), a conic section is the set C of all points $\{p\}$ for which the distance Fp is equal to a constant multiple of the distance Lp :

$$C = \left\{ p \mid |Fp| = e |Lp| \right\}. \quad (2.37)$$

The constant e is called the **eccentricity** of the conic section.

MORE TEXT HERE

2.4.4 Cartesian coefficients

All conic sections can be expressed as the solutions to the following general equation in \mathbb{R}^2 :

$$Ax^2 + Bxy + Cy^2 + Dx + Ey + F = 0, \quad (2.38)$$

where A, B, C, D, E, F are all real coefficients such that A, B and C are all nonzero. The above equation can be written in matrix form as

$$\begin{bmatrix} x & y & 1 \end{bmatrix} \begin{bmatrix} A & B/2 & D/2 \\ B/2 & C & E/2 \\ D/2 & E/2 & F \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = 0. \quad (2.39)$$

In this form, the different types of conic sections arise from the sign of the term

$$\Delta = B^2 - 4AC, \quad (2.40)$$

called the **discriminant** of the conic equation, as following:

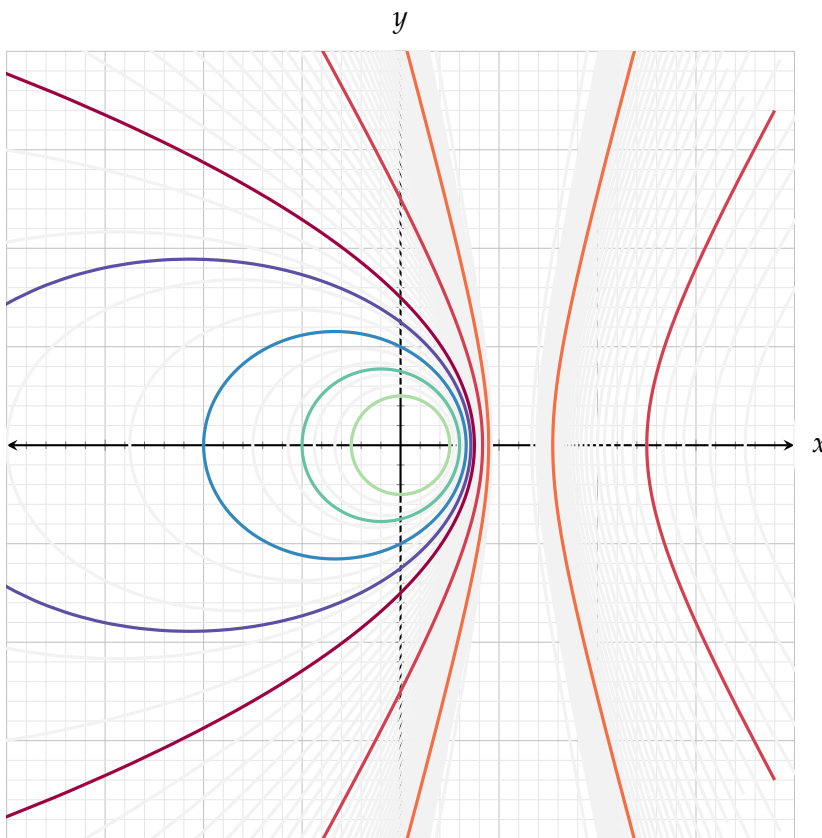


Figure 2.9: Several conic sections with different eccentricities $\{e_i\}$ and the same locus F and directrix L . (NOTE: figure still WIP)

1. If $B^2 - 4AC < 0$ the equation represents an ellipse. If in addition $A = C$ and $B = 0$ the discriminant collapses to $-A^2$ - which represents a circle.
2. If $B^2 - 4AC = 0$, the equation represents a parabola.
3. If $B^2 - 4AC > 0$, the equation represents a hyperbola.

Note that the discriminant can be represented as a 2×2 determinant derived from the matrix form of the Cartesian coefficients:

$$\Delta = -4 \begin{vmatrix} A & B/2 \\ B/2 & C \end{vmatrix}. \quad (2.41)$$

2.4.5 Conic Section From 5 Points

If we know 5 points lying on a conic,

$$\begin{cases} p_1 = (x_1, y_1) \\ p_2 = (x_2, y_2) \\ \vdots \\ p_5 = (x_5, y_5) \end{cases}, \quad (2.42)$$

we can determine all the conic Cartesian coefficients by solving the equation

$$\begin{bmatrix} x_1^2 & x_1 y_1 & y_1^2 & x_1 & y_1 & 1 \\ x_2^2 & x_2 y_2 & y_2^2 & x_2 & y_2 & 1 \\ x_3^2 & x_3 y_3 & y_3^2 & x_3 & y_3 & 1 \\ x_4^2 & x_4 y_4 & y_4^2 & x_4 & y_4 & 1 \\ x_5^2 & x_5 y_5 & y_5^2 & x_5 & y_5 & 1 \end{bmatrix} \begin{bmatrix} A \\ B \\ C \\ D \\ E \\ F \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \quad (2.43)$$

which can be done by finding the null space of the matrix in the above equation.

2.4.6 Some More About Ellipses

Unlike the other conic sections, an ellipse has two focus points (simply called its **foci**): F_1 and F_2 . One of these always corresponds to the conic section definition of the focus. In an ellipse, the sum of the distances from any point to the two foci is always constant. In a sense, an ellipse is an elongated circle: instead of having a single radius, it has two orthogonal **axes**: the semi-major axis a and the semi-minor axis b (such that $a \geq b$).

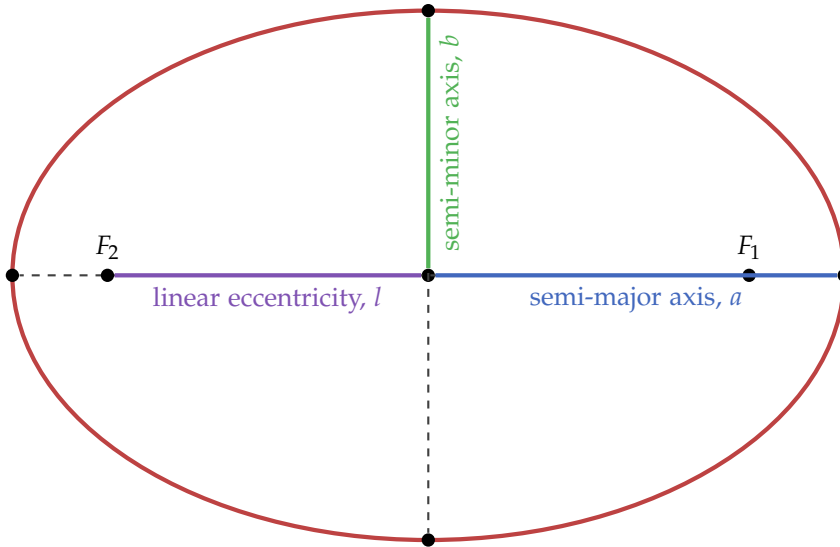


Figure 2.10: Some common geometric properties of an ellipse. NOTE: figure still WIP

The two foci are at a distance of $c = \sqrt{a^2 - b^2}$ away from the center of the ellipse, a measure that also called **linear eccentricity**. The eccentricity of the ellipse is

$$e = \frac{c}{a} = \frac{\sqrt{a^2 - b^2}}{a} = \sqrt{1 - \frac{b^2}{a^2}}. \quad (2.44)$$

As mentioned already, in the case of a circle $e = 0$, and we get that $c = 0$ and that $\sqrt{1 - \frac{b^2}{a^2}} = 0$, i.e. $a = b$. The first equality means that the foci are located at the center of the ellipse, and the second equality means that the semi-major and semi-minor axes of the ellipse are the same. This is exactly what we expect for a circle.

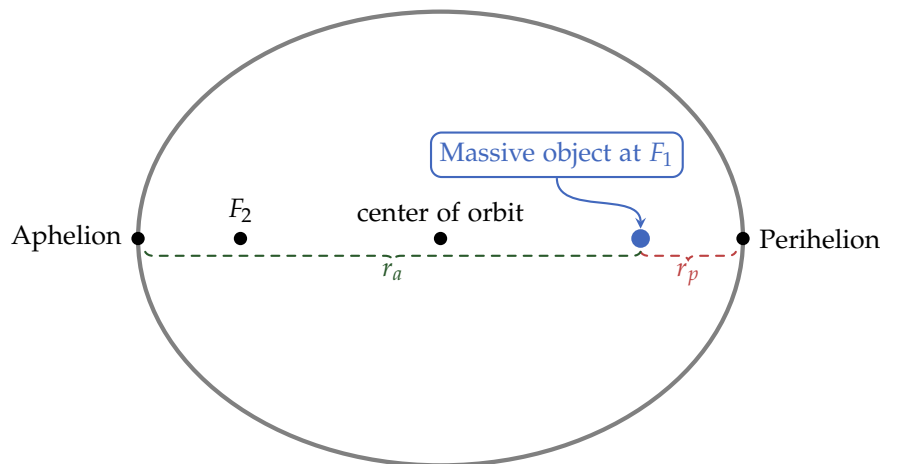
2.4.7 Orbital Shapes

Back from our short detour, we can now discuss the Keplerian orbits in more details. Suppose there are two objects: the first has mass m_1 , and second mass m_2 . For simplicity we will assume that $m_1 \gg m_2$, such that we can assume it is stationary, while the second object experiences the Keplerian orbit - consider for example a satellite orbiting the earth. We will indeed from now on refer to the first (more massive) object simply as the “massive object”, and the second object as the “satellite”.

Although space is 3-dimensional, a Keplerian orbit is always 2-dimensional - since as mentioned before, it is always a conic section. The plane on which the orbit takes place, the **orbital plane**, is determined by the direction of velocity of the satellite and the direction connecting the centers of mass of the massive object and the satellite. The massive object is found at the locus of the conic section. The relative values of m_1 and the angular momentum of the satellite determine the eccentricity of the orbit.

For now, let us concentrate on an elliptic orbit, i.e. where $0 \leq e < 1$. As mentioned, in such an orbit the massive object is at the locus of the ellipse, which is one of its foci - we will call it F_1 here. The point of closest to F_1 on the elliptical orbit is called the **periapsis** and denoted r_p . The point directly opposite r_p , i.e. the point closest to the second focus F_2 of the ellipse and furthest away from F_1 is called the **apoapsis**, denoted r_a (Figure 2.11). The direction from F_1 to the periapsis point is (unsurprisingly) called the **direction of periapsis**.

Figure 2.11: Common terms in elliptical orbits.



The relation between the eccentricity of the orbit and the two distances r_p, r_a is

$$e = \frac{r_a - r_p}{r_a + r_p} = 1 - \frac{2}{\frac{r_a}{r_p} + 1}. \quad (2.45)$$

Conversely, we can write the above relation as

$$\frac{r_a}{r_p} = \frac{1+e}{1-e}. \quad (2.46)$$

Example 2.2 Earth's orbit around the sun

In the sun-earth system, where the sun is the massive object and the earth is the satellite, the periapsis distance is $r_p = 1.470\,984\,50 \times 10^{11}$ [m], and the apoapsis distance is $r_a = 1.520\,975\,97 \times 10^{11}$ [m]. Therefore, the orbital eccentricity of Earth's orbit around the sun is about

$$\begin{aligned} e &= 1 - \frac{2}{\frac{r_a}{r_p} + 1} = 1 - \frac{2}{\frac{1.52097597}{1.47098450} + 1} \\ &= 1 - \frac{2}{1.033985 + 1} = 1 - \frac{2}{2.033985} \\ &= 1 - 0.983291 = 0.016709. \end{aligned}$$

This is a pretty round orbit (not a scientific term).



At any point in a spatial trajectory, the velocity of the object in motion is always tangent to the trajectory it traces. In the case of a Keplerian orbit, the angle between the direction of periapsis and the line connecting F_1 and the position of the satellite $r(t)$ at some given time t is called the **true anomaly**, denoted as θ .

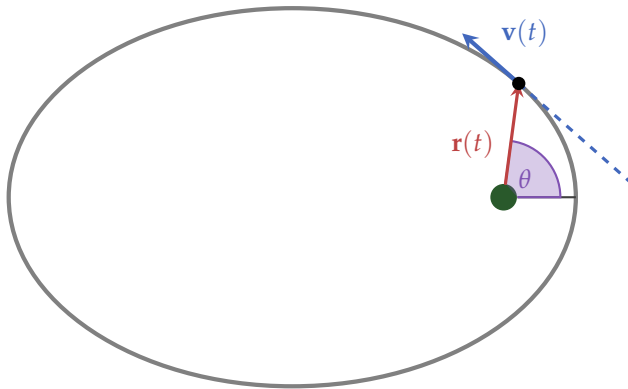


Figure 2.12: Position \mathbf{r} , velocity \mathbf{v} and true anomaly θ in an elliptic orbit. Notice how the velocity is tangent to the trajectory.

In the 17th century, **Johannes Kepler** formalized his **laws of planetary motion**. There are three of them, and we already met the first one: in a 2-body system, the trajectory of the satellite object under gravity is a conic section. His second law states that in any time period Δt along its path, the satellite will sweep a constant area A in relation to the massive object (see Figure 2.13). A direct result of this law is that the satellite moves faster when close to the massive object, and slower when it is far away.

Kepler's third law states that for a given massive object, there is a constant relation between a^3 , the semi-major axis of the orbit to third power - and T^2 , the orbital period (the time it takes the

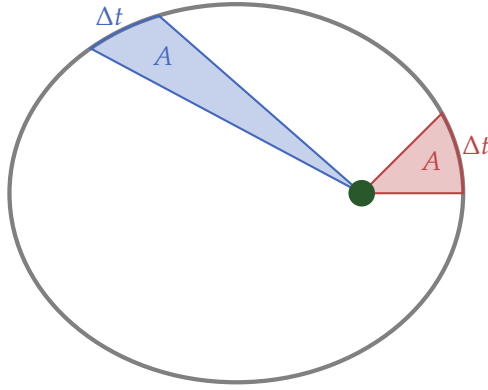


Figure 2.13: Kepler's second law: at any constant time period Δt , the satellite sweeps a constant area A defined between its two positions at times t and $t + \Delta t$ and the massive object.

satellite to complete a single orbit), given by

$$\frac{a^3}{T^2} = \frac{GM}{4\pi^2}. \quad (2.47)$$

In general, Keplerian trajectories have some scalar and vector constants - i.e. quantities that are conserved throughout the orbit. The first is the **standard gravitational parameter** μ , which is the universal gravitational constant G times the total mass of the two objects:

$$\mu = G(m_1 + m_2), \quad (2.48)$$

which in practice is approximately just

$$\mu \approx Gm_1, \quad (2.49)$$

since $m_1 \gg m_2$.

The second constant is the **specific relative angular momentum** of the satellite, denoted \mathbf{h} . It is the cross product of the position vector of the satellite at some time τ , and its velocity at the same time:

$$\mathbf{h} = \mathbf{r} \times \mathbf{v}. \quad (2.50)$$

Note 2.2 Angular momentum is not a vector

Since angular momentum in general is defined via the cross product, it is not a vector but a **pseudovector**. Therefore, it does not denoted using the vector symbol and instead is denoted by a bold-face letter. However, for the sake of simplicity we will consider it as a vector, since in \mathbb{R}^3 vectors and pseudovectors behave very similarly. To better understand the concept of pseudo-vectors it is recommended to learn **geometrical algebra** (also sometimes called **Clifford algebra**). !SOURCE!



The third constant is the **specific orbital energy** ε . It is defined by the **vis-viva equation** using the kinetic energy ε_k and potential energy ε_u :

$$\varepsilon = \varepsilon_k + \varepsilon_u = \frac{v^2}{2} - \frac{\mu}{r}, \quad (2.51)$$

where v is the speed (norm of the velocity) of the satellite at any time t' , and r is the distance to the massive object at the same time t' . Another form for this same energy is

$$\varepsilon = -\frac{\mu}{2a}, \quad (2.52)$$

where here a is the semi-major axis of the trajectory.

We can use the above constants to calculate some conditions for the shape of the trajectory. For example - when it is expected to be perfectly circular. This happens when $e = 0$. By solving Equation 2.51 for v we get

$$\frac{v^2}{2} = \varepsilon + \frac{\mu}{r} = -\frac{\mu}{2a} + \frac{\mu}{r}.$$

For a circular orbit $a = r$, and the above equation becomes

$$\frac{v^2}{2} = -\frac{\mu}{2r} + \frac{\mu}{r} = -\frac{\mu}{2r} + \frac{2\mu}{2r} = \frac{\mu}{2r},$$

i.e.

$$v = \sqrt{\frac{\mu}{r}} = \sqrt{\frac{Gm_1}{r}}. \quad (2.53)$$

In perfect circular motion we know that the total force acting on the object in motion is directed into the center of motion, is always orthogonal to its velocity and has a magnitude of exactly

$$F = m \frac{v^2}{r}. \quad (2.54)$$

We can bring Equation 2.53 to the form of Equation 2.54 by raising v to the second power, multiplying by m_2 and dividing by the radius. Its right hand side would then be

$$F = \frac{m_2}{r} \frac{Gm_1}{r} = G \frac{m_1 m_2}{r^2},$$

exactly the form of the gravitational force. This confirms Equation 2.53.

2.4.8 Calculating the precise trajectory from known magnitudes

To get a precise projected trajectory for a satellite with known velocity \mathbf{v} and position \mathbf{r} around a massive object of known mass M we can use the following steps:

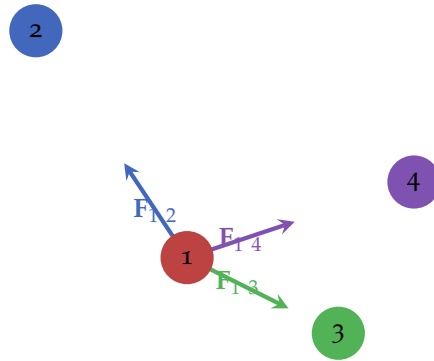
1. Calculate the standard gravitational parameter μ (REF) and the specific angular momentum vector \mathbf{h} (REF). The eccentricity vector can then be calculated using (REF).
2. If $|\mathbf{e}| < 1$, the orbit is elliptical and we can proceed. If not - MORE TEXT?
3. ...

2.4.9 Multiple Objects

We can extend the 2-body problem into an n -body problem: we define a system with n objects of masses m_1, m_2, \dots, m_n , positions $\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_n$ and velocities $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$. The total gravitational force experienced by the i -th object is the sum of all of the gravitational forces from the other objects:

$$\mathbf{F}_i = \sum_{j \neq i} \mathbf{F}_{j \rightarrow i} = G \sum_{j \neq i} \frac{m_i m_j}{r_{ij}^2} \hat{\mathbf{r}}_{ij}. \quad (2.55)$$

Figure 2.14: A 4-body system, showing the total gravitational force acting on one of the objects as the sum of the gravitational forces exerted on it by all other objects.



Unfortunately, there is no general analytical solution to this problem. We can either use approximations and special cases, or use numerical integration to simulate such a system. The next section deals exactly with these kinds of simulations.

2.5 Integration Methods

The term **integration** in the context of numerical simulations means the discrete propagation of a system. In the case of the n -body problem, an integration step includes for each object the following: first finding all the forces acting on the object, adding up all these forces, calculating the object's acceleration (according to Newton's second law) and thus its new velocity (since $a = \dot{v}$) - and finally using this velocity to calculate its new position (since $v = \dot{x}$).

The reason such a scheme is referred to as an "integration" is that we can reverse the derivative-connection between acceleration, velocity and position and write it as an integral:

$$x(t) = x(0) + \int_0^t v(t) dt, \quad (2.56)$$

and in turn

$$v(t) = v(0) + \int_0^t a(t) dt. \quad (2.57)$$

In the case of a constant acceleration, the above two equations add

up together to give

$$x(t) = x_0 + v_0 t + \frac{1}{2} a t^2, \quad (2.58)$$

where $x_0 = x(0)$ and $v_0 = v(0)$.

2.5.1 Euler Method

2.5.2 Verlet Integration

2.5.3 Runge-Kutta Method

2.6 Thermodynamics

3

Thermodynamics

3.1 *Preface*

Text text text

3.2 *Ideal gas*

3.2.1 *Theory*

There are several commonly used models for the behaviour of gasses. A very simple yet powerful one is the **ideal gas** model: it describes gas particles as being perfect spheres which move around in an enclosed container and undergo elastic collisions with other particles and the walls of the container. In common conditions such as atmospheric pressure and temperatures around 300 [K], gasses such as helium (He), argon (Ar), nitrogen (N₂), oxygen (O₂) and carbon dioxide (CO₂) behave like ideal gasses (assuming no chemical reactions take place). However, the ideal gas model fails under high pressures, low temperatures, chemical interactions and some physical processes such as adsorption or multipolar interactions.

The principle equation describing an ideal gas is the **ideal gas law**:

$$PV = nRT, \quad (3.1)$$

where (SI units in parentheses):

- P is the pressure of the gas (Pa),
- V is the volume of the container (m³),
- n is the amount of gas (mol),
- R is the **gas constant**, $R = 8.314 \left[\text{J K}^{-1} \text{mol}^{-1} \right]$, and

- T is the temperature of the gas (K).

Maxwell-Boltzman distribution

Mean free path

Temperature from energy

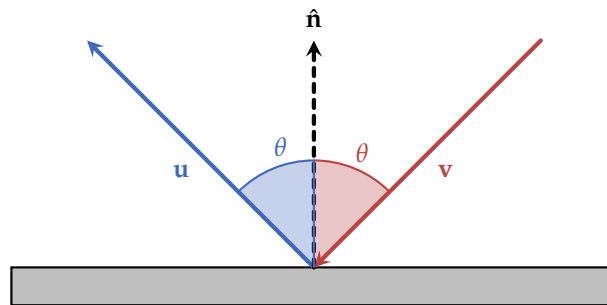
3.2.2 Simulating an ideal gas using perfectly elastic spheres

Text text text

Sphere-wall collision

An elastic collision between a particle and a wall causes the particle's velocity to flip in the direction of the wall's normal (Figure 3.1).

Figure 3.1: Collision of a particle and a wall. The particle bounces in such a way that the component of its velocity \mathbf{v} in the direction of the wall's normal $\hat{\mathbf{n}}$ is flipped. The resulting velocity \mathbf{u} has the same angle to $\hat{\mathbf{n}}$ as \mathbf{v} does.



The component of \mathbf{v} in the direction of $\hat{\mathbf{n}}$ is

$$\mathbf{v}_{\parallel} = \langle \mathbf{v}, \hat{\mathbf{n}} \rangle \hat{\mathbf{n}}. \quad (3.2)$$

Therefore, the component of \mathbf{v} orthogonal to $\hat{\mathbf{n}}$ is

$$\mathbf{v}_{\perp} = \mathbf{v} - \mathbf{v}_{\parallel} = \mathbf{v} - \langle \mathbf{v}, \hat{\mathbf{n}} \rangle \hat{\mathbf{n}}. \quad (3.3)$$

In the case of \mathbf{u} , the orthogonal component is the same as that of \mathbf{v} , but the parallel component is inverted:

$$\begin{aligned} \mathbf{u}_{\perp} &= \mathbf{v}_{\perp}, \\ \mathbf{u}_{\parallel} &= -\mathbf{v}_{\parallel}. \end{aligned} \quad (3.4)$$

and altogether we get

$$\begin{aligned} \mathbf{u} &= \mathbf{u}_{\perp} + \mathbf{u}_{\parallel} \\ &= \mathbf{v} - \langle \mathbf{v}, \hat{\mathbf{n}} \rangle \hat{\mathbf{n}} - \langle \mathbf{v}, \hat{\mathbf{n}} \rangle \hat{\mathbf{n}} \\ &= \mathbf{v} - 2\langle \mathbf{v}, \hat{\mathbf{n}} \rangle \hat{\mathbf{n}}. \end{aligned} \quad (3.5)$$

Example 3.1 A simple sanity check

To make a simple validation of this equation, let's examine the case for a collision with a wall which is oriented in the xy -plane (i.e. its normal points in the z -direction): given the

continuing on the next page ➡

 continuing from the previous page

velocity $\mathbf{v} = \begin{bmatrix} v_x \\ v_y \\ v_z \end{bmatrix}$, Equation 3.5 becomes

$$\begin{aligned}
 \mathbf{u} &= \mathbf{v} - 2\langle \mathbf{v}, \hat{\mathbf{n}} \rangle \hat{\mathbf{n}} \\
 &= \begin{bmatrix} v_x \\ v_y \\ v_z \end{bmatrix} - 2 \left\langle \begin{bmatrix} v_x \\ v_y \\ v_z \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\rangle \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \\
 &= \begin{bmatrix} v_x \\ v_y \\ v_z \end{bmatrix} - 2(\cancel{v_x \cdot 0} + \cancel{v_y \cdot 0} + v_z \cdot 1) \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \\
 &= \begin{bmatrix} v_x \\ v_y \\ v_z \end{bmatrix} - 2v_z \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \\
 &= \begin{bmatrix} v_x \\ v_y \\ v_z \end{bmatrix} - 2 \begin{bmatrix} 0 \\ 0 \\ v_z \end{bmatrix} \\
 &= \begin{bmatrix} v_x \\ v_y \\ v_z \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ -2v_z \end{bmatrix} \\
 &= \begin{bmatrix} v_x \\ v_y \\ -v_z \end{bmatrix},
 \end{aligned}$$

as expected.



Note 3.1 Studying advice

The reader is encouraged to repeat the above calculation for the cases of walls oriented in the xz - and yz -planes.



When a sphere collides with a wall, the distance d from the center of the sphere to the wall is exactly r , the radius of the sphere. The distance between a point P and a plane π can be defined as follows: start from P and move along the direction $\pm \hat{\mathbf{n}}$ (where $\hat{\mathbf{n}}$ is the normal to the plane) until the plane is intersected at a point B . The distance between P and π is then the length PB (Figure 3.2).

If we know a point A on the plane, then we can the vector \mathbf{c} as starting at A and ending at P (i.e. corresponding to the line segment \overline{AP}). The projection of \mathbf{c} onto $\hat{\mathbf{n}}$ then gives the length PM (Figure 3.3).

Figure 3.2: The distance between a point P and a plane π . The distance from P to the plane π is the length PB .

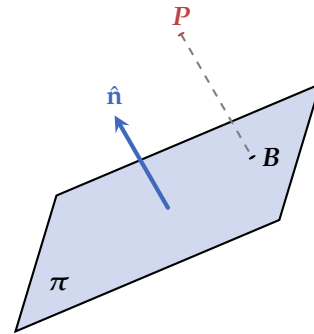
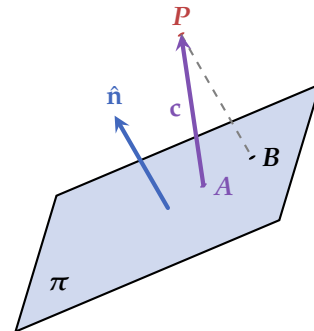


Figure 3.3: The vector \mathbf{c} connects a known point on the plane A and the point P . Its projection on $\hat{\mathbf{n}}$ gives the distance PB .



Since $\hat{\mathbf{n}}$ is a unit vector, the projection of \mathbf{c} onto it is simply the dot product between the two vectors, i.e. the distance d of point P and the plane π is

$$d = \langle \mathbf{c}, \hat{\mathbf{n}} \rangle. \quad (3.6)$$

Example 3.2 Distance of a point and a plane - sanity check

Let $\mathbf{P} = (2, -4, 3)$. Let's calculate its distance to the three principle planes xy , xz and yz . We will use the origin as the point A , since all of these planes contain it. In each case, \mathbf{c} will have as components the respective coordinates of P .

1. The normal to the xy -plane is $\hat{\mathbf{z}} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$. The dot product

$$\langle \mathbf{c}, \hat{\mathbf{n}} \rangle \text{ is therefore } d = \cancel{2 \cdot 0} + \cancel{(-4) \cdot 0} + 3 \cdot 1 = 3.$$

2. The normal to the xz -plane is $\hat{\mathbf{y}} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$. The dot product

$$\langle \mathbf{c}, \hat{\mathbf{n}} \rangle \text{ is therefore } d = \cancel{2 \cdot 0} + (-4) \cdot 1 + \cancel{3 \cdot 0} = -4.$$

3. The normal to the yz -plane is $\hat{\mathbf{x}} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$. The dot product

$$\langle \mathbf{c}, \hat{\mathbf{n}} \rangle \text{ is therefore } d = 2 \cdot 1 + \cancel{(-4) \cdot 0} + \cancel{3 \cdot 0} = 2.$$

As expected, the distance we get from P to each of the principle planes in \mathbb{R}^3 using Equation 3.6 is exactly the respective coordinate of P .



Sphere-sphere collision

Consider two solid spheres which have a single point of contact A . Let m_1, r_1, \mathbf{x}_1 and \mathbf{v}_1 be the mass, radius, position and velocity of the first sphere, and m_2, r_2, \mathbf{x}_2 and \mathbf{v}_2 the respective quantities for the second sphere (Figure 3.4). The line connecting the centers of the two spheres is in the direction $\hat{\mathbf{n}}$ (without loss of generality let us assume that the normal vector $\hat{\mathbf{n}}$ points from \mathbf{x}_1 towards \mathbf{x}_2). The unit vector $\hat{\mathbf{t}}$ is orthogonal to $\hat{\mathbf{n}}$ (and without loss of generality we will assume that it is oriented counter-clockwise from $\hat{\mathbf{n}}$).

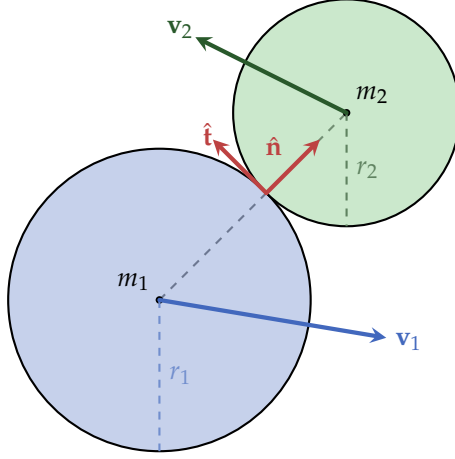


Figure 3.4: Text text text

Note that we did not define a coordinate system, nor the number of dimensions d for the problem. The only restriction is that $d \geq 2$.

Conservation of momentum means that the velocities of the spheres following the collision, $\mathbf{u}_1, \mathbf{u}_2$, are related to their velocities before the collision by

$$\begin{aligned} m_1 \mathbf{v}_1 + m_2 \mathbf{v}_2 &= m_1 \mathbf{u}_1 + m_2 \mathbf{u}_2 \\ \Rightarrow m_1 (\mathbf{v}_1 - \mathbf{u}_1) &= m_2 (\mathbf{u}_2 - \mathbf{v}_2). \end{aligned} \quad (3.7)$$

Conservation of energy means that the velocities are also related by

$$\begin{aligned} \frac{1}{2} m_1 \|\mathbf{v}_1\|^2 + \frac{1}{2} m_2 \|\mathbf{v}_2\|^2 &= \frac{1}{2} m_1 \|\mathbf{u}_1\|^2 + \frac{1}{2} m_2 \|\mathbf{u}_2\|^2 \\ \Rightarrow m_1 (\|\mathbf{v}_1\|^2 - \|\mathbf{u}_1\|^2) &= m_2 (\|\mathbf{u}_2\|^2 - \|\mathbf{v}_2\|^2). \end{aligned} \quad (3.8)$$

However, the forces involved in the collision can not have a component in the $\hat{\mathbf{t}}$ direction, and are limited to only point in the $\hat{\mathbf{n}}$ direction. Therefore, we can reduce the problem to this direction only by projecting all velocities involved in the problem on $\hat{\mathbf{n}}$, i.e. Equation 3.7 becomes

$$m_1 \langle \mathbf{v}_1, \hat{\mathbf{n}} \rangle + m_2 \langle \mathbf{v}_2, \hat{\mathbf{n}} \rangle = m_1 \langle \mathbf{u}_1, \hat{\mathbf{n}} \rangle + m_2 \langle \mathbf{u}_2, \hat{\mathbf{n}} \rangle. \quad (3.9)$$

TEXT TEXT TEXT

$$\begin{aligned}\mathbf{u}_1 &= \mathbf{v}_1 - \frac{2m_2}{m_1 + m_2} \langle \mathbf{v}_1 - \mathbf{v}_2, \hat{\mathbf{n}} \rangle \hat{\mathbf{n}}, \\ \mathbf{u}_2 &= \mathbf{v}_2 + \frac{2m_1}{m_1 + m_2} \langle \mathbf{v}_1 - \mathbf{v}_2, \hat{\mathbf{n}} \rangle \hat{\mathbf{n}}.\end{aligned}\tag{3.10}$$

To avoid redundant calculations, we can factor out the common quantity of both velocities:

$$K = \frac{2}{m_1 + m_2} \langle \mathbf{v}_1 - \mathbf{v}_2, \hat{\mathbf{n}} \rangle \hat{\mathbf{n}},\tag{3.11}$$

yielding

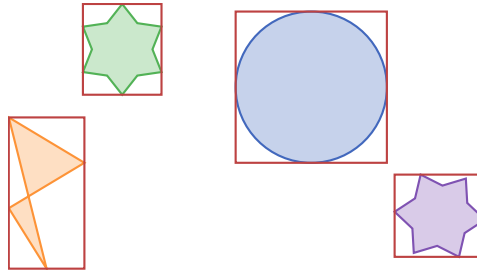
$$\begin{aligned}\mathbf{u}_1 &= \mathbf{v}_1 - Km_2, \\ \mathbf{u}_2 &= \mathbf{v}_2 + Km_1.\end{aligned}\tag{3.12}$$

Reducing collision test complexity

Much like in the case of gravitational simulations, the number of calculations needed to resolve collisions between N particles is $N(N-1) = N^2 - N$, which grows like $\mathcal{O}(N^2)$. And also like in that case, we would like to reduce this complexity. One possible method is using a quad- or oct-tree as in the Barnes-Hutt algorithm mentioned in (REF!), but in the case of perfectly spherical particles which only experience instantaneous elastic collisions, there are much simpler methods.

The first and simplest of these methods is the **Axis-Aligned Bounding Box**, or simply **AABB**. In the 2-dimensional case, a **bounding box** (BB) of an object is the minimal rectangular area surrounding it completely (i.e. that contains the entire object within it). An AABB of an object is (as the name suggests) a BB with sides parallel to the 2 axes (Figure 3.5).

Figure 3.5: Axis-Aligned Bounding Boxes (AABBs) for different objects in 2-dimensions.



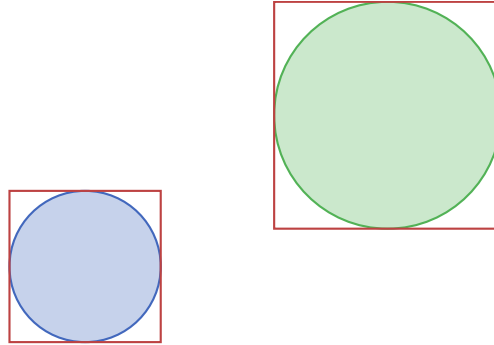
In the case of a circle of radius r , its AABB is a square of side $2r$ with the sides parallel to the x and y axes, respectively, and sharing its center with the circle. Therefore, given a circle of radius r centered at (x, y) , its AABB is the square with the two opposing corners

$$\begin{aligned}\mathbf{c}_{\text{LL}} &= (x - r, y - r), \\ \mathbf{c}_{\text{UR}} &= (x + r, y + r),\end{aligned}\tag{3.13}$$

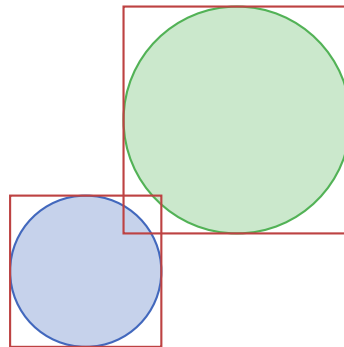
where LL stands for “Lower-Left” and UR stands for “Upper-Right”.

When two objects collide, their AABBs necessarily overlap (Figure 3.6). This means that if we know that two circles have no overlap in their AABBs, we can skip the check whether they collide or not, as it is entirely redundant. Luckily, checking overlaps in AABBs can be very fast compared to checking even the simplest collision, and this way we gain advantage in computation speed. Let's see how it is done.

No overlap, no collision



Overlap but no collision



Both overlap and collision

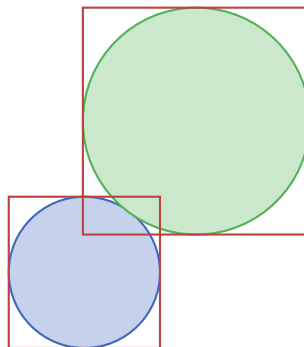
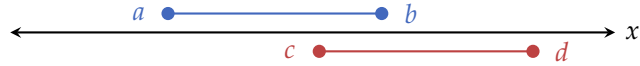


Figure 3.6: AABBs of two circles. There can be three different relations between the circles in regards to their AABBs - no overlap between the AABBs and no collision between the circles (top), and overlap of the AABBs but no collision (middle), and an overlap with a collision (bottom). A collision between the circles necessarily means that there is also an overlap between the AABBs.

We first examine the conditions for an overlap in a single dimension: given two intervals $I_1 = [a, b]$ and $I_2 = [c, d]$, let's assume without loss of generality that a is the smallest number of the four quantities a, b, c and d . The condition for overlap between the two intervals is simply $c < b$, i.e. that the left-most point of I_2 is to the *left* of the right-most point of I_1 (Figure 3.7).

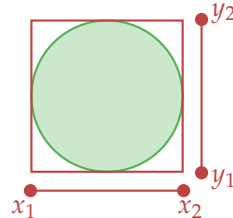
We can consider an AABB across each axis to be an interval: if the lower-left and upper-right corners of the AABB are at $\mathbf{c}_{LL} = (x_1, y_1)$

Figure 3.7: For two intervals $(a, b), (c, d)$ (assuming $a < b, c, d$) to overlap, the condition $c < b$ must be fulfilled.



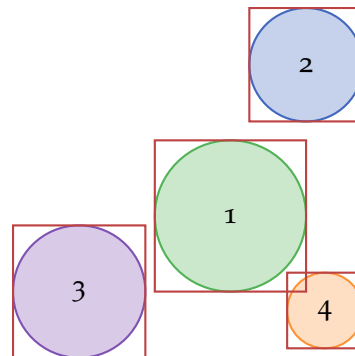
and $\mathbf{c}_{UR} = (x_2, y_2)$, respectively, then the vertical sides of the AABB are represented as an interval $I_{\text{horiz}} = [x_1, x_2]$, and the horizontal sides by $I_{\text{vert}} = [y_1, y_2]$ (Figure 3.8).

Figure 3.8: The sides of an AABB as two intervals $I_{\text{horiz}} = [x_1, x_2]$ and $I_{\text{vert}} = [y_1, y_2]$.



Two AABBs overlap **if and only if** both their horizontal and vertical intervals overlap (Figure 3.9)

Figure 3.9: Four circles with different overlaps between their AABBs: 1-2 and 2-4 overlap horizontally only. 1-3 and 3-4 overlap vertically only. 1-4 overlap in both directions.



3.3 Brownian dynamics