Going back to basics: accelerating exoplanet transit modelling using Taylor-series expansion of the orbital motion

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ABSTRACT
A significant fraction of an exoplanet transit model evaluation time is spent calculating projected distances between the planet and its host star. This is a relatively fast operation for a circular orbit, but slower for an eccentric one. However, because the planet’s position and its time derivatives are constant for any specific point in orbital phase, the projected distance can be calculated rapidly and accurately in the vicinity of the transit by expanding the planet’s $x$ and $y$ positions in the sky plane into a Taylor series at mid-transit. Calculating the projected distance for an elliptical orbit using the four first time derivatives of the position vector (velocity, acceleration, jerk, and snap) is $\sim$ 100 times faster than calculating it using the Newton’s method, and also significantly faster than calculating $z$ for a circular orbit because the approach does not use numerically expensive trigonometric functions. The speed gain in the projected distance calculation leads to 2-25 times faster transit model evaluation speed, depending on the transit model complexity and orbital eccentricity. Calculation of the four position derivatives using numerical differentiation takes $\sim 1 \mu s$ with a modern laptop and needs to be done only once for a given orbit, and the maximum error the approximation introduces to a transit light curve is below 1 ppm for the major part of the physically plausible orbital parameter space.

Key words: Methods: numerical – Techniques: photometric – Planets and satellites

1 INTRODUCTION
An exoplanet transit model aims to reproduce the photometric signal caused by a planet crossing over the limb-darkened disk of its host star (Mandel & Agol 2002; Seager & Mallen-Ornelas 2003; Winn 2010). Evaluation of the transit model can generally be divided into two parts: a) calculation of the projected planet-star centre distance, $z$, and b) calculation of the flux decrement caused by a planet occluding a part of the stellar disk visible to the observer.

The main focus in transit model development has been on the second part, but the calculation of projected distances can actually take a significant fraction of the total model evaluation time. The standard approach for calculating $z$ for a single point in time requires several ($\approx 6$) trigonometric function calls and solving the Kepler’s equation numerically. While worrying about the computational cost of using trigonometric functions might seem frivolous, $z$ needs to be calculated at least once for each photometric data point when evaluating an exoplanet transit model, and multiple times if the model needs to be supersampled (such as for Kepler and TESS long cadence light curves, Kipping 2010). Further, it is already common to have photometric data sets of tens or hundreds of thousands of data points (such as a four-year Kepler light curve), and a transit light curve analysis consisting of a posterior optimisation and Markov Chain Monte Carlo (MCMC) sampling steps can require the model to be evaluated a large number ($\sim 10^6$) of times over all the data points.

Thus, while calculating $z$ using the standard approaches is a trivial matter for small data sets, speeding up the calculation has a potential to yield significant real-life performance gains when modelling modern data sets. While accuracy is more important than speed for a scientific code, a speed increase without any significant sacrifices in accuracy gives freedom for exploratory analyses and experimentation, which can lead to new interesting discoveries, or, at least, increase the reliability of our analyses.

In this short paper we show how a very simple change in the computation of $z$ can lead to a significant speed-up of a transit model without sacrificing model accuracy. The approach is based on high-school level mathematics (Taylor series expansion, a tool that has been used in astronomy and astrophysics for centuries, especially in the research of eclipsing binaries) and has been tested thoroughly. The approach still requires the ability to calculate the eccentric anomaly to a high precision in order to calculate the position derivatives using numerical differentiation, but this needs to be done only for a small number of points in time (seven in our implementation) for a single Keplerian orbit, rather than calculating it for each datapoint separately.

We provide an example Python implementation of the method in Appendix A, and the approach has been adopted as the main $z$ computation method in the PyTransit\textsuperscript{1} (Parviainen 2015) transit modelling package.

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\textsuperscript{1} https://github.com/hpparvi/PyTransit

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2 THEORY

Calculation of the projected planet-star separation \( z \) as a function of time is a necessary step for exoplanet transit model evaluation. The standard approach for calculating \( z \) for a generic eccentric orbit requires the calculation of the eccentric anomaly from the mean anomaly, which requires us to solve Kepler’s equation, for which no closed-form solutions exist. Thus, the Kepler’s equation needs to be solved using numerical methods, such as iteration or the Newton’s method. After the eccentric anomaly has been solved, the computation of \( z \) still requires six trigonometric function calls, which are relatively expensive operations.

Could there be a way to calculate \( z \) without the need to solve Kepler’s equation or use trigonometric functions? Planet’s \( x \) and \( y \) positions in the sky-plane draw smooth and well-behaved curves as a function of time, as shown in Fig. 2. The \( x \) position is a monotonically increasing function of time near the transit, and the \( y \) position is a smooth unimodal function with a single minimum near the transit (this for non-zero impact parameter since \( y \) is constant for \( b = 0 \)). These factors mean that the positions can likely be accurately approximated with low-order polynomials near the transit.

Thus, we choose to use a Taylor series expansion to represent the planet’s position in the sky plane as a function of time,

\[
I(t) = \sum_{n=0}^{\infty} \frac{I^{(n)}(t_0)}{n!} (t - t_0)^n,
\]

where \( l \) is the position (either \( x \) or \( y \)), \( t_0 \) is the point around which the Taylor series is expanded, \( I^{(n)} \) is the \( n \)th derivative of \( l \) evaluated at point \( t_0 \), and \( n! \) is the factorial of \( n \). Mid-transit time where \( x = 0 \) is a natural choice for \( t_0 \) (although other possibilities exist, such as the time of minimum projected distance or the time of minimum \( y \) position), after which we only need to select \( n \) to ensure a sufficient accuracy so that the approximation does not affect the transit model in any significant fashion. After testing the accuracy of different \( n \) (see discussion about accuracy later in Sect. 4), we chose to use the four first time derivatives of position: velocity, acceleration, jerk, and snap.

The first step is to calculate the planet’s position in the sky-plane at mid-transit time and its four time derivatives. For a circular orbit, the position at mid-transit is \([0, -b]\), where \( b \) is the impact parameter. However, for an eccentric orbit the \( y \) position differs from \(-b\).

We use a seven-point central finite difference method (Fornberg 1988) to calculate the derivatives. This requires us to calculate the positions at seven uniformly spaced times centred around the mid-transit time. The calculation of these locations requires using a standard accurate method for evaluating Keplerian orbits, but this needs to be done only once for a given orbit.

The velocity, acceleration, jerk, and snap vector component \( d \) (either \( x \) or \( y \)) can be computed given the points \( \bar{l} = \bar{l}(t_0 - ih) \) where \( i = [-3, -2, -1, 0, 1, 2, 3] \) and \( h \) is the time step, as

\[
v_d = \frac{d_{-3} + 9d_{-2} - 45d_{-1} + 45d_1 - 9d_2 + d_3}{60h},
\]

\[
a_d = \frac{2d_{-3} - 27d_{-2} + 270d_{-1} - 490d_0 + 270d_1 - 27d_2 + 2d_3}{180h^2},
\]

\[
j_d = \frac{d_{-3} - 8d_{-2} + 13d_{-1} - 13d_1 + 8d_2 - d_3}{8h^3},
\]

\[
s_d = \frac{-d_{-3} + 12d_{-2} - 39d_{-1} + 56d_0 - 39d_1 + 12d_2 - 2d_3}{6h^4}.
\]

After the derivatives have been calculated, the projected distance can be computed for time \( t_c \) by first calculating the time difference to the nearest transit centre, \( t \),

\[
E = \left[ t_c - t_0 + 0.5p \right] \left\lfloor \frac{p}{p} \right\rfloor,
\]

\[
t = t_c - (t_0 + E)p,
\]

where \( E \) is the epoch, \( t_0 \) the mid-transit time, \( \lfloor \cdot \rfloor \) denotes the floor operation, and \( p \) the orbital period, and then evaluating the Taylor series at \( t \) as

\[
\bar{t} = \bar{t}_0 + \bar{v}t + \frac{1}{2}\bar{a}t^2 + \frac{1}{6}\bar{j}t^3 + \frac{1}{24}\bar{s}t^4.
\]

\[
z = |\bar{t}|
\]

where \( \bar{t}_0 \) is the position vector at mid-transit, and \( \bar{v}, \bar{a}, \bar{j}, \) and \( \bar{s} \) are the velocity, acceleration, jerk, and snap vectors, respectively.

The approximation requires that the planet’s position is evaluated at seven points in time for an orbit that does not evolve in time (that is, the orbital parameters do not evolve in time). However, if the orbit is perturbed by external forces, such as other massive bodies in a multiplanet system, the terms need to be calculated separately for each transit. This leads to a photodynamical model where the terms are calculated using a set of positions calculated with an n-body integrator, as done in PyTTV by Korth et al. (2020, in preparation).

The derivatives known, the computation of \( z \) requires only multiplications, summation, and a single square-root operation. Given the simplicity of the approximation, we provide an example Python implementation in Appendix A.

3 PERFORMANCE

The real-world improvement in the transit model evaluation speed depends on how heavy the transit shape model is relative to the \( z \) calculation method (that is, how large fraction of the transit model execution time is spent on computing the orbit). For the transit model assuming quadratic stellar limb darkening by Mandel & Agol (2002), the speed gain is between 6 (eccentric orbit calculated using the Newton’s method) and 2 (circular orbit), that is, the model is 6 times faster to evaluate for an eccentric orbit when \( z \) is calculated using a Taylor series expansion rather than Newton’s method. For the most simple transit shape model that assumes uniform stellar disk, the speed gain is between 24 (eccentric orbit) and 2 (circular orbit). In

\footnote{We group the expressions slightly differently for the actual implementation to reduce sensitivity to floating point round-off errors, as shown in Appendix A.}
both cases, the minimum speed gain is around 2 (that is, the model is at least twice as fast to calculate).

4 ACCURACY

While the Taylor series approximation of $z$ is significantly computationally faster than the other approaches for calculating $z$, its practical usability depends on the error caused to the exoplanet transit model. The accuracy of the approximation depends on the three-dimensional curvature of the orbit at the mid-transit time, what again depends on the semi-major axis, eccentricity, and argument of periastron.

Figure 2 shows the actual and approximated $x$ and $y$ coordinates and the projected distance $z$ for three increasingly eccentric short-period orbits with orbital period, $p$, of 1 d, scaled semi-major axis, $a$, of $4 \, R_\star$, and impact parameter, $b$, of 0.5. Figure 3 shows the maximum absolute errors in a transit light curve caused by the approximation for a circular orbit as a function of the planet-star separation (that is, the semi-major axis) at mid-transit (upper panel) and orbital period (lower panel). The orbits correspond to three planets with radius ratio of 0.15, 0.1, and 0.05 orbiting a star with a stellar density, $\rho_\star$, of 1.2 g cm$^{-3}$ with an impact parameter of 0.5. The figure focuses on the ultra-short-period and short-period regime because the error is below 1 ppm for semi-major axes larger than $5 \, R_\star$. Considering the currently known transiting exoplanets, the maximum absolute error introduced by the approximation would be $\sim 10$ ppm, staying below 1 ppm for all but the most extreme ultra-short-period planets.

Figure 4 shows the maximum absolute errors in transit light curves caused by the approximation for four sets of orbital parameters as a function of increasing eccentricity. The scenarios are: a) $p = 2.5 \, d$ and $a = 7.5 \, R_\star$, b) $p = 5 \, d$ and $a = 15 \, R_\star$, c) $p = 15 \, d$ and $a = 25 \, R_\star$, and d) $p = 30 \, d$ and $a = 40.0 \, R_\star$, and the eccentricities cover the range of eccentricities for known planets with periods less or equal to the scenario period. The maximum error for most of the physically plausible orbits is below 1 ppm, and still below 10 ppm for the very eccentric orbits.

5 CONCLUSIONS AND DISCUSSION

A planet’s normalised planet-star centre distance near a transit (or a secondary eclipse) can be calculated using the planet’s sky-plane position at mid-transit time and its four first time derivatives for the whole physically plausible orbital parameter space without sacrificing transit model accuracy. The approach is $\sim 100$ times faster to calculate than an approach using Newton’s method to solve the Kepler’s equation, and yields a 2-24 gain in transit model evaluation speed. Further, since the approach is based on expanding the sky-plane position, the position can be used directly with transit models that break the radial symmetry, such as the gravity-darkened transit model for rapidly rotating stars by Barnes (2009). A gravity-darkened model utilising the approach to compute the $(x, y)$ position has been added to a coming PyTRANSIT version (v2.4), but here the speed gain over the standard approach is relatively small due to computational cost of the transit model itself.

As clear from Figs. 3 and 4, the errors introduced by the approximation into the transit model are negligible. The absolute maximum error is below 1 ppm for all but the shortest orbital periods and highest eccentricities, and generally below 10 ppm for any currently known planets.

We could also expand $z$ directly into a Taylor series instead of the sky-plane $x$ and $y$ positions. However, the projected distance has a relatively sharp minimum (compared to the behaviour of $x$ and $y$ positions), and the time of the minimum does not necessarily match our mid-transit time for which $x = 0$. Thus, expanding $z$ would
The planet-star contact points (beginning of ingress, $T_1$, end of ingress, $T_2$, beginning of egress, $T_3$, and end of egress $T_4$, Winn 2010) are easy to compute numerically. Calculation of a single point takes $\approx 500$ ns, and the calculation of different durations ($T_{14}$, $T_{23}$, $T_{12}$, and $T_{23}$) takes between 1–2 $\mu$s. We do not include the code to calculate the contact points here, but make it available from PyTransit repository in GitHub. PyTransit also uses the $T_1$ and $T_4$ points to create a transit bounding box in time that is used to ensure we do not waste time evaluating the model over the out-of-transit points.

The centre time for the series expansion, $t_0$, affects the accuracy. It could be beneficial to choose $t_0$ to match the time where $y$ is minimum (so that $y$ velocity is zero), or the time of minimum $z$. It could also be possible to choose a different $t_0$ for the $x$ and $y$ expansion. However, both approaches would require more computation to solve those locations than just choosing the mid-transit time, and are probably not worth the work considering that the current approach already reaches an accuracy that has basically no effect on the transit light curve model.

The speed gains discussed in Sect. 3 depend significantly on the overall implementation of the whole transit model. The examples in this study consider light curves where most of the points are in transit (that is, most of the out-of-transit data has been removed). Having a light curve with a small fraction of in-transit points (such as when modelling a full Kepler or TESS light curve directly) will significantly increase the speed gain unless the transit model is smart enough to skip the out-of-transit points.

The final effect on the evaluation speed also depends on the other parts of the posterior computation, such as the noise model. The gain will be smaller when the posterior computation time is dominated by the noise model evaluation (such as when using brute-force Gaussian Processes), and greatest in an analysis with a computationally cheap noise model and a large number of data points.

The approach has been adopted as the main $z$ computation method in the PyTransit exoplanet transit modelling package by Parviainen (2015). However, considering the simplicity of the approach, we believe it can be useful to everyone developing exoplanet transit models and modelling frameworks independent of the programming language used. Thus, the approach can be easily added to other commonly used transit modelling packages, such as EXOFAST by Eastman et al. (2013), batman by Kreidberg (2015), ELLC by Maxted (2016), or TLCM by Csizmadia (2020).

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DATA AVAILABILITY
There are no new data associated with this article.

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APPENDIX A: EXAMPLE IMPLEMENTATION
Here we show an example numba-accelerated Python implementation of the approach used by the PyTransit transit modelling package. First, a method to calculate the sky-plane x and y derivatives

```python
from numba import njit
from numpy import (arctan2, cos, sin, sqrt, floor, mod, pi)

@njit
def ta_newton_s(t, t0, p, e, w):
    offset = arctan2(sqrt(1.0-e**2)*sin(0.5*pi-w),
                     e + cos(0.5*pi - w))
    offset -= e*sin(offset)
    ma = mod(2*pi*(t-(t0-offset*p/2*pi))/p, 2*pi)
    ea = ma
    err = 0.05
    k = 0
    while abs(err) > 1e-8 and k < 1000:
        err = ea - e*sin(ea) - ma
        ea = ea - err/(1.0-e*cos(ea))
        k += 1
    sta = sqrt(1.0-e**2)*sin(ea)/(1.0-e*cos(ea))
    cta = (cos(ea)-e)/(1.0-e*cos(ea))
    return arctan2(sta, cta)

@njit
def xyeo(t, t0, p, e, w, ae, ci):
    f = ta_newton_s(t, t0, p, e, w)
    r = ae / (1.+ e*cos(f))
    x = -r * cos(w + f)
    y = -r * sin(w + f) * ci
    return x, y

@njit(fastmath=True)
def z_taylor(tc, t0, p, y0, vx, vy, ax, ay, jx, jy, sx, sy):
    epoch = floor((tc - t0 + 0.5*p) / p)
    t = tc - (t0 + epoch * p)
    t2 = t**2

# Time step for central finite difference
# ---------------------------------------
# I've tried to choose a value that is small
# enough to work with USP orbits and large
# enough not to cause floating point problems
# with the fourth derivative (anything much
# smaller starts hitting the double precision
# limit.)
dt = 2e-2

# Calculation of X and Y positions
# ---------------------------------
# These could be calculated with a loop
# with X and Y as arrays, but I'm unrolling
# the loop manually because this seems to
# give a small speed advantage with numba.
# ae = a*(1.-e**2)
# ci = cos(i)
x0, y0 = xyeo(t0-3*dt, t0, p, e, w, ae, ci)
x1, y1 = xyeo(t0-2*dt, t0, p, e, w, ae, ci)
x2, y2 = xyeo(t0-1*dt, t0, p, e, w, ae, ci)
x3, y3 = xyeo(t0  , t0, p, e, w, ae, ci)
x4, y4 = xyeo(t0+1*dt, t0, p, e, w, ae, ci)
x5, y5 = xyeo(t0+2*dt, t0, p, e, w, ae, ci)
x6, y6 = xyeo(t0+3*dt, t0, p, e, w, ae, ci)

# First time derivative of position: velocity
# -------------------------------------------
# a, b, c = 1/60, 9/60, 45/60
vx = (a*(x6-x0)+b*(x1-x5)+c*(x4-x2))/dt
vy = (a*(y6-y0)+b*(y1-y5)+c*(y4-y2))/dt

# Second time derivative: acceleration
# ------------------------------------
# a, b, c, d = 1/90, 3/20, 3/2, 49/18
ax = (a*(x0+x6)-b*(x1+x5)+c*(x2+x4)-d*x3)/dt**2
ay = (a*(y0+y6)-b*(y1+y5)+c*(y2+y4)-d*y3)/dt**2

# Third time derivative: jerk
# ---------------------------
# a, b, c = 1/8, 1, 13/8
jx = (a*(x0-x6)+b*(x5-x1)+c*(x2-x4))/dt**3
jy = (a*(y0-y6)+b*(y5-y1)+c*(y2-y4))/dt**3

# Fourth time derivative: snap
# ----------------------------
# a, b, c, d = 1/6, 2, 13/2, 28/3
sx = (-a*(x0+x6)+b*(x1+x5)-c*(x2+x4)+d*x3)/dt**4
sy = (-a*(y0+y6)+b*(y1+y5)-c*(y2+y4)+d*y3)/dt**4

return y3, vx, vy, ax, ay, jx, jy, sx, sy
```

Here xyeo calculates the x and y positions at given times using the Newton’s method to calculate the true anomaly (ta_newton_s). Now, the projected distance can be calculated using the derivatives as a Taylor series

```python
@njit(fastmath=True)
def z_taylor(tc, t0, y0, vx, vy, ax, ay, jx, jy, sx, sy):
    """Projected planet-star distance using a Taylor series expansion."""

    epoch = floor((tc - t0 + 0.5*p) / p)
    t = tc - (t0 + epoch * p)
    t2 = t**2
```

MNAR 500, 1–6 (2020)
\( t_3 = t^2 \times t_2 \)
\( t_4 = t^3 \times t_3 \)
\( p_x = v_x \times t + a_x \times t_2/2 + j_x \times t_3/6 + s_x \times t_4/24 \)
\( p_y = y_0 + v_y \times t + a_y \times t_2/2 + j_y \times t_3/6 + s_y \times t_4/24 \)
\textbf{return} \ \sqrt{(p_x \times p_x) + (p_y \times p_y)}

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