LETTER TO THE EDITOR

Asteroseismic versus Gaia distances: A first comparison

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ABSTRACT

Context. The Kepler space mission led to a large number of high-precision time series of solar-like oscillators. Using a Bayesian analysis that combines asteroseismic techniques and additional ground-based observations, the mass, radius, luminosity, and distance of these stars can be estimated with good precision. This has given a new impetus to the research field of galactic archeology.

Aims. The first data release of the Gaia space mission contains the Tycho-Gaia Astrometric Solution (TGAS) catalogue with parallax estimates for more than 2 million stars, including many of the Kepler targets. Our goal is to make a first proper comparison of asteroseismic and astrometric parallaxes of a selection of dwarfs, subgiants, and red giants observed by Kepler for which asteroseismic distances were published.

Methods. We compare asteroseismic and astrometric distances of solar-like pulsators using an appropriate statistical errors-in-variables model on a linear and on a logarithmic scale.

Results. For a sample of 22 dwarf and subgiant solar-like oscillators, the TGAS parallaxes considerably improved on the Hipparcos data, yet the excellent agreement between asteroseismic and astrometric distances still holds. For a sample of 938 Kepler pulsating red giants, the TGAS parallaxes are much more uncertain than the asteroseismic ones, making it worthwhile to validate the former with the latter. From errors-in-variables modelling we find a significant discrepancy between the TGAS parallaxes and the asteroseismic values.

Conclusions. For the sample of dwarfs and subgiants, the comparison between astrometric and asteroseismic parallaxes does not require a revision of the stellar models on the basis of TGAS. For the sample of red giants, we identify possible causes of the discrepancy, which we will likely be able to resolve with the more precise Gaia parallaxes in the upcoming releases.

Key words. asteroseismology – stars: fundamental parameters – stars: oscillations – astronomical databases: miscellaneous – parallaxes – Galaxy: structure

1. Introduction

The seismic study of stars has undergone a revolution during the past decade, thanks to the space missions CoRoT (launched in 2006; Auvergne et al. 2009) and Kepler (launched in 2009; Borucki et al. 2010). Not only did these space data confirm the method of asteroseismology (for an extensive monograph; see Aerts et al. 2010), they also allowed powerful applications to thousands of stars across stellar evolution for a wide variety of stellar birth masses. Major breakthroughs of relevance to the current study of stellar distances were the discovery of acoustic non-radial pulsation modes in red giants (De Ridder et al. 2009) and the excitation of dipole mixed modes probing both the deep interior and the structure of the outer envelope of such stars (e.g. Beck et al. 2011). Thanks to their mixed gravity and acoustic character, mixed modes allow the core properties of a star to be tuned and therefore can be used to pinpoint the evolutionary status (Bedding et al. 2011).

Asteroseismology of red giants offers the unique opportunity of providing stellar ages for studies of the Milky Way, termed galactic archeology (e.g. Miglio et al. 2013). Indeed, the measurement of the frequency at maximum oscillation power and of the large frequency separation, along with a spectroscopic estimate of the effective temperature, can be transformed into high-precision estimates of the stellar mass and radius by assuming that the input physics of solar models is also applicable to solar-like stars. Under this reasonable assumption, stellar masses and radii can be derived with relative precisions of merely a few per cent, while further comparison with stellar models provides a seismic age estimate with a precision below 20% when systematic uncertainties due to modest variations in the input physics are taken into account (Chaplin et al. 2014; Metcalfe et al. 2014). Proper computation of the apparent CoRoT or Kepler magnitude according to the passbands of these satellites then allows the luminosity of the stars to be transformed into an “asteroseismic” distance (Silva Aguirre et al. 2012; Rodrigues et al. 2014; Anders et al. 2016).

So far, the asteroseismic distances of stars in the solar neighbourhood have been compared a posteriori with Hipparcos values for the distances whenever available, with good agreement (e.g. Silva Aguirre et al. 2012). With the Gaia mission in full swing, we foresee a quantum leap forward in this research, both in the number of targets and the precision in measuring the distance. After five years of nominal monitoring, the Gaia distance estimates are expected to be so precise that they can serve as input to improve the physics of stellar interiors, leading to model-independent radii and better ages than currently available as input for exoplanet studies and galactic archeology. Here we take a
first step to compare asteroseismic distances with the astrometric values by considering the first Gaia data release (Gaia DR1; e.g. Gaia Collaboration et al. 2016a,b,c).

2. Nearby Kepler dwarfs and subgiants

In their original study to verify asteroseismically determined parameters with HIPPARCOS distances in a self-consistent way, Silva Aguirre et al. (2012) investigated 22 dwarfs and subgiants having HIPPARCOS parallaxes with a relative error better than 20%. These stars are close neighbours of the Sun, with distances ranging from 20 to 260 pc, and are not known to have companions. In their seismic modelling method, Silva Aguirre et al. (2012) included corrections for reddening in an iterative approach based on distance-dependent integrated maps of extinction. The seismic distance estimates are also mildly dependent (2012) included corrections for reddening in an iterative approach based on distance-dependent integrated maps of extinction. The seismic distance estimates are also mildly dependent on the adopted metallicity and this was taken into account in their estimates of the uncertainty for the seismic distance.

Figure 1 shows the uncertainties of the parallaxes obtained with TGAS versus those corresponding with HIPPARCOS values. The TGAS parallaxes considerably improve the HIPPARCOS values, making a new comparison between astrometric and asteroseismic parallaxes appropriate. In Fig. 2 we show the comparison between the two. The overall correspondence between the two parallax determinations is excellent, showing the remarkable achievement of asteroseismology. The best fit, shown in red, is slightly offset with respect to the bisector, but as we show in the following, the offset is not nearly significant enough to justify a revision of the seismic models.

The statistical problem at hand is one where we wish to verify a linear relationship between the true parallaxes \( \sigma^*_i \) obtained from TGAS and the true parallaxes \( S^*_i \) derived using seismology:

\[
S^*_i = \alpha + \beta \sigma^*_i. \tag{1}
\]

Neither \( \sigma^*_i \) nor \( S^*_i \) is observed. Instead, the corresponding observed quantities \( \sigma_i \) and \( S_i \) can be modelled with

\[
\sigma_i = \sigma^*_i + \eta_i, \tag{2}
\]

\[
S_i = S^*_i + e_i, \tag{3}
\]

where \( \eta_i \) and \( e_i \) are the measurement errors. Such a model is called an errors-in-variables model (also known as the measurement error model) and has been studied extensively in the statistics literature, see e.g. Fuller (1987). Here, we have the additional feature that the variances of both \( \eta_i \) and \( e_i \) have been quantified. Because these measurement errors may vary considerably in magnitude between stars, it is beneficial to explicitly take this between-star heterogeneity into account.

Equations (1)–(3) yield the following relationship in terms of observable quantities:

\[
S_i = \alpha + \beta (\sigma_i - \eta_i) + e_i, \quad \text{or} \quad S_i = \alpha + \beta \sigma_i + e_i. \tag{4}
\]

here we decomposed the error of \( S_i \) in a purely measurement error component, \( e_i \), with star-specific known variance \( \sigma^2_i \), and an error component capturing variability in the relationship between \( \sigma_i \) and \( S_i \), denoted by \( \eta_i \), for which we assume a constant but unknown variance \( \sigma^2_\eta \). Likewise, the star-specific variance of \( \eta_i \) is denoted by \( \sigma^2_\eta \). If the relationship is indeed linear, then estimates of \( \sigma^2 \) close to zero are expected. Furthermore, if the linear relationship coincides with the bisector, then \( \alpha \) is expected to be close to zero and \( \beta \) close to one. From Eq. (4) it follows that

\[
E(S_i) = \alpha + \beta \sigma_i, \tag{5}
\]

\[
\text{Var}(S_i) = \beta^2 \sigma^2_{\sigma_i} + \sigma^2_\eta + \sigma^2, \tag{6}
\]

implying that the parameters can be estimated by maximising the likelihood, for example based on a normal distribution

\[
S_i \sim N\left(\alpha + \beta \sigma_i, \beta^2 \sigma^2_{\sigma_i} + \sigma^2_\eta + \sigma^2\right). \tag{7}
\]

or the moments derived therefrom. We note that by setting \( \sigma^2_{\sigma_i} = \sigma^2_\eta = 0 \), an ordinary linear regression follows with homoskedastic measurement error (contrary to what is observed). Ordinary linear regression is expected to yield similar regression estimates as the errors-in-variables model when the uncertainty in the dependent variable is considerably larger than the uncertainty in the predictor; in the reverse case, a different regression is obtained, an effect termed regression attenuation in statistics.

We fitted model (7) using the SAS procedure NLMIXED (SAS Institute Inc. 2014) and list the results in Tables 1 and A.3 (including the measurement errors), and A.1 and A.2 (ignoring the measurement errors). As expected from the plot, in both cases the value 0 (resp. 1) is in the 95% confidence interval of the intercept \( \alpha \) (resp. the slope \( \beta \)), clearly indicating that the bisector cannot be rejected as a plausible model.
Table 1. Parameters of the errors-in-variables model (5)–(7), fitted to 22 nearby pulsators, taking the measurement errors into account.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Estimate</th>
<th>Std. error</th>
<th>95% confidence interval</th>
</tr>
</thead>
<tbody>
<tr>
<td>α</td>
<td>0.437</td>
<td>0.333</td>
<td>[-0.254, 1.128]</td>
</tr>
<tr>
<td>β</td>
<td>0.999</td>
<td>0.031</td>
<td>[0.935, 1.062]</td>
</tr>
<tr>
<td>σ^2</td>
<td>0.197</td>
<td>0.163</td>
<td>[-0.140, 0.537]</td>
</tr>
</tbody>
</table>

Notes. The values of α and σ are expressed in milliarcsec, while β is dimensionless.

3. Kepler red giants

Relying on a Bayesian framework and a similar procedure for reddening corrections to those adopted by Silva Aguirre et al. (2012), Rodrigues et al. (2014) derived distances to 1989 red giants observed by Kepler and followed up spectroscopically with APOGEE (Pinsonneault et al. 2014). This resulted in seismic distances between 0.5 and 5 kpc, with relative uncertainties of less than 2%. For 938 of these giants we were able to find a reliable crossmatch in the Gaia DR1 TGAS catalogue. We discarded stars with negative TGAS parallaxes as we compare them later on with asteroseismic parallaxes that were imposed to be positive. The red giants are considerably more distant than the pulsating dwarfs, leading to much more uncertain TGAS parallaxes. Figure 3 shows that the uncertainties of the TGAS parallaxes are substantially larger than those obtained from the seismic distance. Hence the current TGAS parallaxes cannot be used to calibrate the seismic values, but a reverse validation is meaningful at this stage of the Gaia mission. We carried out such a validation using the same methodology as in the previous section. We prefer to analyse parallaxes rather than distances in order to avoid having to take the reciprocal of the more variable measurements. We note that, unlike for the dwarfs, we have no information on their possible binarity; the current TGAS parallaxes and their error estimates assume that they are single stars.

Figure 4 shows the astrometric and asteroseismic parallaxes of the sample of 938 red giants. Clearly the relation is far less stringent than for the dwarfs and subgiants in the previous section. This, together with the fact that the uncertainties of one measure are much larger than the other, prompted us to add two other models to our analysis. First, like for ordinary least-squares, the errors-in-variables model (4)–(7) is not symmetric, i.e. fitting the seismic parallax \( S_i \) as a function of the TGAS parallax \( \varpi_i \), or the other way around can lead to different results. We therefore added the errors-in-variables model \( \varpi_i = x + \rho S_i \).

Second, the models outlined above work under the assumption that the noise sources are additive. To assess the impact of this assumption, we add a model that assumes multiplicative noise, which can be conveniently set up on a logarithmic scale.

The relation to be tested is 10* \( D_i \) = \( \pi_i \), where \( D_i = 1/S_i \) is the asteroseismic distance expressed in pc and \( \pi_i \) the astrometric parallax expressed in mas. Using decadic logarithms, this leads to the following linear model of the logarithmic observables \( \log D_i \) and \( \log \pi_i \):

\[
\log D_i = 3 + \alpha l_j - \beta \ln(10) \log \sigma_j + \varepsilon_{ij},
\]

where we assume the noise component \( \varepsilon_{ij} \) to be normally distributed,

\[
\varepsilon_{ij} \sim N(0, \sigma_l^2 + \text{Var}(\log D_j) + (1 + \beta)^2 \text{Var}(\log \sigma_j)),
\]

with \( \sigma_l^2 \) a constant but unknown variance that captures the variability in the relationship between \( \log D_j \) and \( \log \sigma_j \). The variance of the decadic logarithms of the observed quantities can be approximated as

\[
\text{Var}(\log D_j) = \frac{\text{Var}(D_j)}{\ln(10)^2 D_j^2},
\]

\[
\text{Var}(\log \sigma_j) = \frac{\text{Var}(\sigma_j)}{\ln(10)^2 \sigma_j^2}.
\]

The parameters were again estimated using the SAS procedure NL MIXED.

Figure 4 synthesises the analysis results for the fits including all 938 stars. The corresponding parameter estimates and their uncertainties are listed in Tables 2 and A.5–A.8. Not surprisingly, of all linear fits, the ordinary least-squares fit \( S_i = \alpha + \beta \pi_i \)
and crossmatched with TGAS, the relative uncertainties on the astrometric parallaxes are much larger than on the seismic distances, turning the latter into a valuable instrument to validate the former. All the models we applied – an errors-in-variables model, ordinary least-squares, and a logarithmic model – lead to a significant difference between astrometric and asteroseismic parallaxes. There can be several underlying causes. The uncertainties of the TGAS parallaxes may be underestimated or could be subject to systematic errors. On the other hand, interstellar extinction corrections and/or too poorly known bulk metallicity may have introduced a systematic uncertainty for the asteroseismic parallax. Given that all stars in the Gaia DR1 TGAS catalogue were assumed to be single, binarity may also be part of the cause.

Expectations are that the accuracy of the Gaia astrometric distance measurements will surpass the seismic measurements by the end of the mission. Gaia Data Release 2 (end 2017) will contain the astrometry of a billion stars, while Gaia Data Release 3 (end 2018) will deliver the orbital astrometric solutions for binaries with periods longer than 2 months. This will allow us to improve our current research and transfer it into a quantitative calibration of asteroseismic distances for a variety of stellar populations.

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References

Aerts, C., Christensen-Dalsgaard, J., & Kurtz, D. W. 2010, Asteroseismology (Heidelberg: Springer Berlin)
Fuller, W. 1987, Measurement Error Models (New York: John Wiley & Sons)
SAS Institute Inc. 2014, SAS OnlineDoc 9.4
### Appendix A: Additional parameter estimates

In this appendix, we provide the outcome of parameter estimates using the models discussed in the main text. Tables A.1–A.4 concern the sample of 22 nearby dwarfs and subgiants. Tables A.5–A.8 relate to the sample of 938 red giants. Tables A.2 and A.3 provide the estimates of the intercept and slope for the reverse model \( \sigma_i = \kappa + \rho S_i \), which was introduced specifically for the giants, but which we also fitted for the dwarfs for the sake of completeness. The fits lead to exactly the same conclusion as the normal model (7). Also here the value zero is well within the 95% confidence interval of the intercept, and the value 1 is within the corresponding interval of the slope, indicating that the bisector is a plausible model for the relation between astrometric and asteroseismic parallaxes. Not taking the uncertainties into account leads to an estimate of \( \sigma^2 \) (the error component capturing variability in the linear relationship), which is significantly different from zero, but this significance disappears when the uncertainties are taken into account showing that the linear model is adequate.

**Table A.1.** Parameters of the model (5)–(7) fitted to 22 nearby pulsators, without taking the measurement errors into account.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Estimate</th>
<th>Std. error</th>
<th>95% Confidence interval</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \alpha )</td>
<td>0.549</td>
<td>0.296</td>
<td>[−0.064, 1.162]</td>
</tr>
<tr>
<td>( \beta )</td>
<td>0.990</td>
<td>0.022</td>
<td>[0.945, 1.034]</td>
</tr>
<tr>
<td>( \sigma^2 )</td>
<td>0.556</td>
<td>0.168</td>
<td>[0.208, 0.904]</td>
</tr>
</tbody>
</table>

**Notes.** The values of \( \alpha \) and \( \sigma \) are expressed in milliarcsec, while \( \beta \) is dimensionless.

**Table A.2.** Parameters of the model (5)–(7) using the relation \( \sigma_i = \kappa + \rho S_i \), applied to 22 nearby pulsators, without taking the measurement errors into account.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Estimate</th>
<th>Std. error</th>
<th>95% Confidence interval</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \kappa )</td>
<td>−0.430</td>
<td>0.306</td>
<td>[−1.065, 0.205]</td>
</tr>
<tr>
<td>( \rho )</td>
<td>1.000</td>
<td>0.022</td>
<td>[0.955, 1.045]</td>
</tr>
<tr>
<td>( \sigma^2 )</td>
<td>0.562</td>
<td>0.170</td>
<td>[0.211, 0.914]</td>
</tr>
</tbody>
</table>

**Notes.** The values of \( \kappa \) and \( \sigma \) are expressed in milliarcsec, while \( \rho \) is dimensionless.

**Table A.3.** Parameters of the model (5)–(7) using the relation \( \sigma_i = \kappa + \rho S_i \).

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Estimate</th>
<th>Std. error</th>
<th>95% Confidence interval</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \kappa )</td>
<td>−0.240</td>
<td>0.327</td>
<td>[−0.917, 0.437]</td>
</tr>
<tr>
<td>( \rho )</td>
<td>0.982</td>
<td>0.029</td>
<td>[0.921, 1.042]</td>
</tr>
<tr>
<td>( \sigma^2 )</td>
<td>0.185</td>
<td>0.158</td>
<td>[−0.142, 0.511]</td>
</tr>
</tbody>
</table>

**Notes.** The fit was applied to 22 nearby pulsators, taking the measurement errors into account. The values of \( \kappa \) and \( \sigma \) are expressed in milliarcsec, while \( \rho \) is dimensionless.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Estimate</th>
<th>Std. error</th>
<th>95% Confidence interval</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \alpha )</td>
<td>−0.054</td>
<td>0.032</td>
<td>[−0.119, 0.012]</td>
</tr>
<tr>
<td>( \beta )</td>
<td>−0.034</td>
<td>0.029</td>
<td>[−0.095, 0.027]</td>
</tr>
<tr>
<td>( \sigma^2 )</td>
<td>0.00038</td>
<td>0.00028</td>
<td>[−0.00020, 0.00096]</td>
</tr>
</tbody>
</table>

**Table A.4.** Parameters of the model (8) on the logarithmic scale, fitted to 22 nearby pulsators, taking the measurement errors into account.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Estimate</th>
<th>Std. error</th>
<th>95% Confidence interval</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \alpha )</td>
<td>0.008</td>
<td>0.005</td>
<td>[−0.001, 0.018]</td>
</tr>
<tr>
<td>( \beta )</td>
<td>−0.540</td>
<td>0.028</td>
<td>[−0.596, −0.485]</td>
</tr>
<tr>
<td>( \sigma^2 )</td>
<td>−0.174</td>
<td>0.0005</td>
<td>[−0.175, −0.173]</td>
</tr>
</tbody>
</table>

**Notes.** The values of \( \kappa \) and \( \sigma \) are expressed in milliarcsec, while \( \rho \) is dimensionless.

**Table A.5.** Parameters of the model (8) on the logarithmic scale, fitted to 938 pulsating red giants, taking the measurement errors into account.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Estimate</th>
<th>Std. error</th>
<th>95% Confidence interval</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \kappa )</td>
<td>0.886</td>
<td>0.022</td>
<td>[0.042, 0.130]</td>
</tr>
<tr>
<td>( \rho )</td>
<td>0.853</td>
<td>0.020</td>
<td>[0.813, 0.892]</td>
</tr>
<tr>
<td>( \sigma^2 )</td>
<td>−0.024</td>
<td>0.002</td>
<td>[−0.029, −0.019]</td>
</tr>
</tbody>
</table>

**Notes.** The fit was applied to 938 red giants, without taking the measurement errors into account. The values of \( \kappa \) and \( \sigma \) are expressed in milliarcsec, while \( \rho \) is dimensionless.

**Table A.6.** Parameters of the model (5)–(7) using the relation \( \sigma_i = \kappa + \rho S_i \), applied to a sample of 938 red giants, taking the measurement errors into account.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Estimate</th>
<th>Std. error</th>
<th>95% Confidence interval</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \alpha )</td>
<td>0.341</td>
<td>0.021</td>
<td>[0.300, 0.383]</td>
</tr>
<tr>
<td>( \beta )</td>
<td>0.666</td>
<td>0.020</td>
<td>[0.626, 0.706]</td>
</tr>
<tr>
<td>( \sigma^2 )</td>
<td>0.072</td>
<td>0.003</td>
<td>[0.066, 0.079]</td>
</tr>
</tbody>
</table>

**Notes.** The fit was applied to 938 red giants, without taking the measurement errors into account. The values of \( \kappa \) and \( \sigma \) are expressed in milliarcsec, while \( \rho \) is dimensionless.

**Table A.7.** Parameters of the model (5)–(7).

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Estimate</th>
<th>Std. error</th>
<th>95% Confidence interval</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \kappa )</td>
<td>0.166</td>
<td>0.025</td>
<td>[0.116, 0.216]</td>
</tr>
<tr>
<td>( \rho )</td>
<td>0.797</td>
<td>0.024</td>
<td>[0.749, 0.845]</td>
</tr>
<tr>
<td>( \sigma^2 )</td>
<td>0.086</td>
<td>0.004</td>
<td>[0.079, 0.094]</td>
</tr>
</tbody>
</table>

**Notes.** The values of \( \kappa \) and \( \sigma \) are expressed in milliarcsec, while \( \rho \) is dimensionless.