A study of 3-dimensional shapes of asteroid families with an application to Eos

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Abstract

In order to fully understand the shapes of asteroids families in the 3-dimensional space of the proper elements \( (a_p, e_p, \sin I_p) \) it is necessary to compare observed asteroids with N-body simulations. To this point, we describe a rigorous yet simple method which allows for a selection of the observed asteroids, assures the same size-frequency distribution of synthetic asteroids, accounts for a background population, and computes a \( \chi^2 \) metric. We study the Eos family as an example, and we are able to fully explain its non-isotropic features, including the distribution of pole latitudes \( \beta \). We confirm its age \( t = (1.3 \pm 0.3) \) Gyr; while this value still scales with the bulk density, it is verified by a Monte-Carlo collisional model. The method can be applied to other populous families (Flora, Eunomia, Hygiea, Koronis, Themis, Vesta, etc.).

1. Introduction

A rigorous comparison of observations versus simulations of asteroid families is a rather difficult task, especially when the observations look like Figure 11. Observed proper elements \( a_p, e_p, \sin I_p \), supplied by physical data (colour indices \( a^\alpha \), \( t - \varepsilon \) in this case), show a complicated structure of the Eos family, halo, together with many neighbouring families, overlapping halos, and background asteroids, of course. The hierarchical clustering method alone (HCM, Zappalà et al. 1995) is then practically useless.

Family identification itself affects dynamical studies and vice versa. We would need the family to determine initial conditions. On the other hand, we would need a dynamical study to understand whereever family members could be. There are several well-known weaknesses of HCM, which were demonstrated e.g. in a ‘crime-scene’ Fig. 8 of Nesvorný et al. (2015).

The HCM needs a free parameter, either the cutoff velocity \( v_{cut} \), or the quasi-random level QRL. It is also unable to associate halos. Last but not least, the background is never precisely uniform which can be clearly seen at the edges of currently stable zones, close or inside gravitational resonances, or even in stable zones where the population was deteriorated by dynamical processes in the distant past (cf. Cybele region; Carruba et al. 2015).

On the other hand, synthetic families evolve in the course of simulation and loose their members, consequently we should use a variable \( v_{cut} \), but its optimal value is again generally unknown. No direct comparison is thus possible.

That is a motivation for our work. We describe a method suitable to study 3-dimensional shapes of asteroid families, taking into account all proper orbital elements, including possibly non-uniform background, and matching the size-frequency distribution at the same time. Our method still relies on a preliminary selection of observed asteroids according to their colours (or albedos) to suppress – but not fully exclude – interlopers. A comparison of the observed asteroids with an output of N-body simulation is performed by means of counting the bodies in proper-element ’boxes’, and a suitable \( \chi^2 \) metric. Because we are forced to select synthetic asteroids randomly (a Monte-Carlo approach), we can expect some stochasticity of the results.

We present an application to the Eos family (family identification number, FIN = 606), one of the most studied families to date, mentioned already by Hirayama (1918). Together with our previous works (Vokrouhlický et al. 2006; Brož and Morbidelli, 2013), this paper forms a long-term series focused on its long-term evolution. We use up-to-date catalogues of proper elements (Knežević and Milani, 2003), and brand new spin data (Hanuš et al., 2018).

Let us recall that the Eos family is of K taxonomic type, while the background is mostly C type. Mothé-Diniz et al. (2008) suggested either a partially differentiated parent body, with meteorite analogues CV, CO or R; or an undifferentiated one, with CK analogues. There was a discovery of a recent breakup of (6733) 1992 EF (Novaković and Tsvirouliš, 2014), belonging to the family core, what makes Eos even more interesting for space weathering studies, because we may see both old (1.3 Gyr) and young (4 Myr) surfaces.

2. Methods

Before we proceed with the description of the method, let us explain three problems we have to solve and describe the underlying dynamical model.
Figure 1: Top panel: the proper semimajor axis $a_p$ vs proper inclination $\sin I_p$ for all asteroids in the broad surroundings of Eos family. The range of proper eccentricities is $e_p \in (0.0; 0.3)$. If they have colour data in the SDSS MOC4 catalogue (Parker et al., 2008), the colours correspond to indices $a^* = i - z$ which are closely related to taxonomy, namely blue is close to C-complex taxonomy, red to S-complex, and magenta to K-type. The whole sample contains 18 471 asteroids. There are other prominent families visible: Hygeia (C-type, bottom-right), Veritas (C, next to Eos), Tirela (S, upper right), Telramund (S, below Eos); a close inspection would show 32 families in total! Bottom panel: the same plot for a typical outcome of N-body simulations, assuming a disruption of a parent body, ejection of fragments with some velocity field, and their long-term dynamical evolution due to gravitational perturbations, resonances, chaotic diffusion, the Yarkovsky effect, the YORP effect, etc. The two panels are not directly comparable.

Figure 2: K-type asteroids selected from Figure 1 with known colour indices $a^* \in (0.0; 0.1), i - z \in (-0.03; 0.08)$. The visual geometric albedo had to be $p_V > 0.07$ (or unknown). This subset is much more homogeneous and contains 1 991 asteroids. No other prominent families except Eos can be seen; the only exception may be some contamination by Tirela (upper right) due to inherent photometric noise. This subset seems already suitable for a comparison with N-body simulations.

2.1. Problem 1: Selection of asteroids

In principle, we can select any subset of asteroids (e.g. by using SDSS colour data, or WISE albedo data) to decrease a contamination by interlopers, or an overlap with other families in the neighbourhood [Parker et al., 2008; Masiero et al., 2011]; an approach also used in a multidomain HCM (Carruba et al., 2013). We can also simulate any subset at will, but we should definitely check surroundings where the bodies can be scattered to, because this may be a key constraint.

For Eos family, it is easy because of its distinct K taxonomic type which is defined for our purposes in terms of the SDSS colour indices $a^* \in (0.0, 0.1), i - z \in (-0.03, 0.08)$, and the geometric albedo $p_V > 0.07$ (if known in WISE or IRAS catalogues). If only colours are known, we select the asteroids according to them, and assume their $p_V = 0.158$ which corresponds to the median value of Eos members. As a result, only $1/10^{th}$ of asteroids remain, but this is still sufficient (Figure 2). Practically all other families have disappeared, the background is much more uniform. The only exception may be some contamination from the Tirela family (seen as a concentration in the upper right corner of Fig. 2), arising from a photometric noise on S-type asteroids, and a gap at large $\sin I_p > 0.25$.

Regarding the homogeneity of albedos, the WISE data exhibit a wide distribution, and we should check whether it can be related to a heterogeneous parent body. The uncertainties $\sigma_p$ arise mainly from photon noise, and NEATM model systematics. In a statistical sense, even the single albedo value $\bar{p}_V = 0.158$ would result in a relatively wide distribution because $\sigma_p$ values are relatively large, which is demonstrated in Figure 3 where we used the $\sigma$’s of individual measurements together with the (constant) $\bar{p}_V$ to randomly generate the new distribution of $p_V$’s. The Eos family thus seems homogeneous rather than heterogeneous.
Figure 3: The observed differential distribution of visual geometric albedos $p_V$ for the Eos family from the WISE catalogue [Masiero et al. 2011] (black solid), and for the same set of bodies with $p_V$ values assigned randomly, assuming a Gaussian distribution with a constant mean $\bar{p}_V = 0.158$, and 1-$\sigma$ uncertainty declared in the catalogue (dashed gray). The widths of the two distributions are similar, so using the constant $\bar{p}_V$ (if unknown) is not a poor approximation.

2.2. **Problem 2: Size-frequency distribution**

The size-frequency distributions (SFDs) should match for both the observed and synthetic populations, but the latter changes in the course of time (Figure 4). In order to compare apples with apples, we have to scale the SFD. In other words, we randomly select the same number of synthetic bodies (together with their orbits, of course) as the number of observed bodies, in each of prescribed size bins ($D, D + dD$). Let us emphasize we do not rely on the assumption of a constant SFD. To this point, it is definitively useful to start with a larger number of synthetic bodies, so that we still have more than observed at the end of simulation.

This random selection of synthetic asteroids to match the SFD of observed asteroids is needed at every single output time step of the simulation. Even multiple selections at one time step might be useful. This way, we would naturally account for an additional (and often neglected) uncertainty which arises from the fact we always choose the initial conditions from some underlying distributions (e.g. from a prescribed velocity field), but we cannot be absolutely sure that our single selection is not a lucky fluke.

2.3. **Problem 3: Non-uniform background**

A background has to be accounted for otherwise it is essentially impossible to explain a lot of bodies far from the family. First, we need to find some observed background, not very far from the family; in our case, a suitable population seems to be at $\sin I_p \in (0.06, 0.12)$ and $(0.24, 0.30)$. It has its own size-frequency distribution, and we should use the same SFD for the synthetic background. As a first approximation, we model the background as a random uniform distribution in the space of proper elements.

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1In principle, we can estimate the original SFD of the family but it is not our goal here. The overall change of slope due to dynamical decay (for selected $t$) can be estimated already from Fig. 4.
However, Murphy’s law for backgrounds states: The background is never uniform. Especially below and above the 7/3 mean-motion resonance with Jupiter we can expect a difference (see the example in Figure 5).

Again, there is a non-negligible stochasticity. We shall at least try a different random seed. The number density of background objects can be also treated as a free parameter. There is also a priori unknown systematic contamination by neighbouring families, but this is not necessarily present right ‘under’ the a priori also ff least try a di

2.4. Dynamical model

Our dynamical model was described in detail in Brož et al. (2011). We briefly recall it contains a modified SWIFT integrator (Levison and Duncan, 1994; Laskar and Robutel, 2001), both the diurnal and seasonal Yarkovsky thermal effects (Vokrouhlický, 1998; Vokrouhlický and Farinella, 1999), which induce a semimajor axis drift $da/dt$; all mean-motion and secular resonances, captures and corresponding drifts $de/dt$, $dl/dt$, the YORP effect, changing the spin rate $\omega$ and the obliquity $\gamma$ (Čapek and Vokrouhlický, 2004), with the efficiency parameter $c_{\text{YORP}} = 0.33$ (Hanuš et al., 2011), simplified collisional reorientations by means of a prescribed time scale dependent on size $D$ (Farinella et al., 1998), random period changes due to mass shedding after reaching the critical spin rate $\omega_{\text{crit}}$ (Pravec and Harris, 2000), and suitable digital filters for computations of mean and proper elements (Quinn et al., 1991; Šidlichovský and Nesvorný, 1996).

Initial conditions are kept as simple as possible. We assume an isotropic disruption, with the ejection velocity components Gaussian, with the dispersion proportional to $1/D$, and $V_3 = 93$ m s$^{-1}$ for $D_3 = 5$ km, an estimate based on our previous work (Vokrouhlický et al., 2006). Consequently, the distribution of the velocity magnitude $|v_3|$ is Maxwellian (see Figure 6). We start with 6,545 synthetic bodies, with the SFD covering $D \in (1.5; 100)$ km. Spins are also isotropic and periods uniform, $P \in (2; 10)$ h.

The thermal parameters remain the same as in our previous works: the bulk density $\rho = 2.500$ kg m$^{-3}$, the surface density $\rho = 1.500$ kg m$^{-3}$, the conductivity $K = 0.001$ W m$^{-1}$ K$^{-1}$, the specific capacity $C = 680$ J kg$^{-1}$, the Bond albedo $A = 0.1$, the infrared emissivity $\epsilon = 0.9$. For simplicity, we assumed these parameters to be constants, although some of them may be size-dependent (as $K$ in Delbo et al., 2015), or temperature-dependent (Anderson et al., 1991).

The free parameters of our model are the maximum of velocity distribution $v_{\text{max}}$ (Fig. 6), the true anomaly $f_{\text{imp}}$, and the argument of pericentre $e_{\text{imp}}$ at the time of impact, which are interrelated by means of the Gauss equations. We may be forced to tune also other osculating orbital elements of the parent body, but for the moment we take those of (221) Eos as the nominal case.

Among the fixed parameters is the bulk density $\rho$. Usually, the age scales linearly with $\rho$ due to the non-gravitational accelerations. Theoretically, if there are both gravitational and non-gravitational accelerations acting at the same time (e.g. Yarkovsky drift in $a$ and chaotic diffusion in $e$) we may be able to break this degeneracy. However, based on our previous experience, we do not expect this for Eos. Neighbouring Veritas may be more suitable for this approach, by the way. Alternatively, one can use collisional models which exhibit a different scaling with $\rho$ (cf. Sec. 4.1).

We integrate the equations of motion with the time step $\Delta t = 91$ d, and the time span 4 Gyr. The output time step after computations of mean elements, proper elements, and final running-window filter is $\Delta t_{\text{filt}} = 10$ Myr.
2.5. Black-box method

We can eventually proceed with a so-called ‘black-box’ method (see Figure 7) (i) choose 180 boxes with \( \Delta a = 0.0243 \) au, \( \Delta e = 0.025 \), \( \Delta \sin I = 0.240 \) in our case aligned with the J7/3 and J9/4 resonances; (ii) count the numbers of observed asteroids located in these boxes; (iii) compute the observed incremental SFD globally, in the full domain; (iv) compute the background incremental SFD globally; (v) at every single output time step we compute the synthetic incremental SFD globally again (saving also lists of bodies in the respective size bins); (vi) for every single size bin \((D, D + dD)\) we draw a synthetic background population of \( N_{\text{bg}} \) bodies from a random uniform distribution (in the whole range of \( a_p, e_p, \sin I_p \)); if the volume where the background was selected differs from our volume of interest, we have to use a suitable factor, i.e. \( f N_{\text{bg}} \); (vii) we scale the synthetic SFD to the observed one by randomly choosing \( N_{\text{obs}} - f N_{\text{bg}} \) bodies from the lists above; (viii) we count the numbers of all synthetic asteroids located in the boxes; (ix) finally, we compute the metric

\[
\chi^2 = \sum_{i=1}^{N_{\text{obs}}} \frac{(N_{\text{syn}} - N_{\text{obs}})_i^2}{\sigma_{\text{syn}}^2 + \sigma_{\text{obs}}^2},
\]

where the uncertainties are assumed Poisson-like, \( \sigma = \sqrt{N} \). Using both \( \sigma_{\text{obs}} \) and \( \sigma_{\text{syn}} \) in the denominator prevents ‘extreme’ \( \chi^2 \) contributions in boxes where \( N_{\text{obs}} \to 0 \). We shall keep in mind though the corresponding probability distribution of \( \chi^2 \) may be somewhat skewed. There is some freedom related to the box sizes (binning), but within the limits of meaningfulness (neither a single box nor zillions of boxes), the method should give statistically comparable results as we always analyse the same information.

Unlike traditional simplified methods fitting an envelope to \((a_g, H)\) or \((a_g, 1/D)\), we shall obtain not only an upper limit for the age, but also a lower limit.

3. Results

Hereinafter, we discuss not only the best-fit model, but also several bad fits which are actually more important, because the ‘badness-of-fit’ assures a solid conclusion about the Eos family.

3.1. The nominal model

The nominal model is presented in Figure 7. We focus on the proper semimajor axis \( a_p \) vs proper eccentricity \( e_p \) distribution, having only one box in inclination \( \sin I_p \). The initial conditions (top left) are so different from the observations (bottom middle) it is almost hopeless to expect a good fit anytime in the future. However, at around \( t = 1.3 \) Gyr the situation suddenly changes (top middle); it is almost unbelievable that the synthetic family is so similar to the observations! The final state (top right) is again totally different. The \( \chi^2 \) reaches values as low as \( N_{\text{box}} \), so we may consider the best fit to be indeed reasonable. The age interval is \( t = (1.3 \pm 0.3) \) Gyr. Let us emphasize that the fit so good only because we carefully accounted for all three problems outlined in Section 2.

3.2. Bad fit 1: Ejection velocity tail

Because our sample is 3 times larger than the observed sample, we can easily resample our synthetic bodies without actually computing the N-body simulation anew, e.g. selecting only those with low ejection velocity \( v_{ej} < 200 \) m s\(^{-1}\). Consequently, all bodies are initially located above the J7/3 resonance, and below the J11/5.

Using the same post-processing as above we arrived at Figure 8. It is clear that the ‘best fit’ is actually a bad fit compared to the nominal model. The notable differences are below the J7/3 resonance, and above the J11/5 where the numbers of bodies are never sufficient to match the observations (cf. Fig. 7, bottom middle).

It is worth to note there is a small family just below the J7/3 resonance, namely (36256) 1999 XT (FIN 629). Tsirvoulis et al. (2018) discovered a link to Eos by analysing the overall V-shape in the semimajor axis \( a_p \) vs the absolute magnitude \( H \) diagram. It seems aligned with the original velocity field of the Eos family — it has the same \( \sin I_p \) as the family core, but slightly larger \( e_p \approx 0.1 \), because of the ‘ellipse’ in \((a_p, e_p)\) visible in Fig. 7 (top left). We thus conclude, (36256) family is actually a remnant of the original velocity field.

If this is true, it may further contribute to the contamination of the ‘pristine zone’ between the J7/3 and J5/2 resonances, apart from low-probability crossings of the former resonance. This region was analysed by Tsirvoulis et al. (2018), where authors carefully subtracted the contribution of all families (including Eos), extracted the SFD of remaining background asteroids and computed the slope of the primordial (post-accretion) SFD.

3.3. Bad fit 2: Parent body inclination

If we look on contrary on the proper semimajor axis \( a_p \) vs proper inclination \( \sin I_p \) distribution (Figure 9) there is a problem with the nominal model. Inclinations are all the time too low (and the \( \chi^2 \) too high compared to \( N_{\text{box}} \)). This would affect a 3-dimensional fit too, of course.

Nevertheless, it seems sufficient to adjust the inclination by approximately 0.005 rad to get a significantly better fit. \( \chi^2 \) decreased from 238 down to 181. This seems still too high wrt. 130, but this approach is possibly too simplified, because we only shifted the output data. In reality, the resonances (in particular the \( z_1 \)) do not shift at all, they are determined by the positions of giant planets, and we should perform the N-body integration anew to obtain a correct (\( a_p, \sin I_p \)) distribution.

3.4. Bad fit 3: True anomaly \( f_{\text{imp}} < 120^\circ \)

To demonstrate the sensitivity of our ‘black-box’ method with respect to the impact parameters, we present an alternative N-body simulation which started with the true anomaly.
The proper semimajor axis $a_p$ vs proper eccentricity $e_p$ for the nominal simulation scaled to the observed SFD (as described in the main text) (top row). Bodies are plotted as green dots. Colours correspond to the number of bodies in 180 boxes, outlined by $N_{\text{box}} = 2.934\text{ au}$. Positions of major mean-motion and 3-body resonances are also indicated ($J7/3, J9/4, J11/5$, and $3J-2S-1$). The correspondence between the best-fit and the observations is surprisingly good, with $\chi^2$ metric compared to the actual number of boxes $N_{\text{box}}$ (bottom right). The 1-σ, 2-σ and 3-σ levels (dotted lines) and the inferred 3-σ uncertainty of the age (yellow strip) are indicated too.

Figure 7: The proper semimajor axis $a_p$ vs proper eccentricity $e_p$, for the best-fit at $t = 1340$ Myr (middle), the end of simulation (right); as well as the observations (bottom middle), and the respective $\chi^2$ metric compared to the actual number of boxes $N_{\text{box}}$ (bottom right). The correspondence between the best-fit and the observations is surprisingly good, with $\chi^2 = 141$, $N_{\text{box}} = 134$ (not all boxes are populated), and $\chi^2 = N_{\text{box}}$. The 1-σ, 2-σ and 3-σ levels (dotted lines) and the inferred 3-σ uncertainty of the age (yellow strip) are indicated too.

Figure 8: Bad fit 1: the proper semimajor axis $a_p$ vs proper eccentricity $e_p$ for a subset of bodies with the ejection velocities $v_{ej} < 200$ m s$^{-1}$, i.e. without the tail of the distribution. Initially, all bodies were located above the $J7/3$ resonance. Observations were shown in Fig. 7 (bottom middle). The ‘best-fit’ at $t = 1340$ Myr, with $\chi^2 = 197$, $N_{\text{box}} = 134$, is much worse than the nominal case. The number of bodies below the $J7/3$ resonance is too low. Consequently, the velocity tail is needed to get a better fit.

Figure 9: Bad fit 2: the proper semimajor axis $a_p$ vs proper inclination $I_p$ for the synthetic population (top panel), and the temporal evolution of $\chi^2$ (bottom panel). The boxes are consequently different, $Au = 0.0234$ au, $\Delta I = 0.02$, $\Delta e_p = 0.0; 0.3$, so is the resulting ‘best-fit’ value $\chi^2 = 238$, $N_{\text{box}} = 130$. The parent body would have to be shifted in inclination by approximately 0.005 rad to get a better fit.
tial condition might be also too simple. In particular, the velo-
city field might have been non-isotropic even though in cata-
strophic disruptions (like Eos) we rather expect a high degree of
isotropy (Sevčík et al., 2017). Generally, it is better to keep
both as simple as possible to have the lowest possible number
of free parameters.

Let us finally compare our nominal best-fit model to another
two distributions (size and spin) and the respective models (col-
lisional and rotational).

4.1. Collisional evolution

In a Monte-Carlo collisional model, size-frequency distribu-
tions are evolved due to fragmentation and reaccumulation.
We assume two populations: the main belt, and the Eos fam-
ily. Their physical properties are summarized by the scaling
law $Q_N(r)$, for which we assume parameters of basalt at
5 km s$^{-1}$ from Benz and Asphaug (1999). To compute the ac-
tual evolution, we use the Boulder code by Morbidelli et al.
(2009). Parametric relations in the Boulder code, which are
needed to compute the fragment distributions, are derived from
SPH simulations of Durda et al. (2007).

We assume the initial SFD of the main belt relatively similar
to the currently observed SFD, because we focus on the already
stable solar system, with the fixed intrinsic impact probability
$P_{\text{imp}} = 3.1 \times 10^{-18}$ km$^{-2}$ yr$^{-1}$ and the mean velocity $v_{\text{imp}} = 5.28$ km s$^{-1}$.

The initial SFD of the Eos family has the same slope as the observed SFD in the range $D \in (15; 50)$ km, and it is
prolonged down to $D_{\min} = 0.005$ km. We also account for the
size-dependent dynamical decay due to the Yarkovsky effect,
with $N(t + \Delta t) = N(t) \exp(-\Delta t/r)$, where the time scale $\tau(D)$ is
taken from Bottke et al. (2005).

The resulting collisional evolution is shown in Figure 11.
The observed knee at $D \approx 15$ km is very important, because
it usually arises from a collisional grinding. If we start with the
constant slope from above, we can match the observed SFD at
about 1.3 Gyr which is in accord with the dynamics.

It is worth to note the scaling of the age with the bulk den-
sity $\rho$ is different from dynamics, which in principle allows to
resolve the problem. However, the collisional model is sensitive
to the initial conditions and using a steeper SFD would result in
longer age. In other words, everything is based on the simple
assumption of the constant slope. It would be useful to base the
initial conditions on a specific SPH model for the Eos family,
with the parent body size reaching up to 380 km (according to
an extrapolation of Durda et al. 2007 results).

4.2. Spin distribution

At the same time, it is worth to check the observed dis-
tribution of pole latitudes $\beta$, reported in Hanuš et al. (2018).
Our dynamical model evolves the spin $(\omega, \gamma)$, which affects
the Yarkovsky drift rate $d\omega/dt$, but we do not account for spin-
orbital resonances (so we would not explain a clustering in the
Koronis family, Slivan 2002). Nevertheless, if we use the cur-
cent model for Eos, with the same post-processing, but focus on
$(a_p, \sin \beta)$ boxes instead, we obtain the results summarized in
Figure 12.

We start from an isotropic distribution of spins, which means
isotropic also in $\sin \beta$. After about 1.3 Gyr, it is possible to

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Figure 10: Bad fit 3: a detail of the proper semimajor axis $a_p$ vs proper eccentricity $e_p$ (top panel), and the temporal evolution of $\chi^2$ (bottom panel) for the simulation with the true anomaly at the time of impact $f_{\text{imp}} = 0^\circ$, and the argument of perihelion $\omega_{\text{imp}} = 30^\circ$. The ‘best-fit’ $\chi^2 = 711$ is so high compared to $N_{\text{box}} = 124$ that the simulation was not computed up to 4000 Myr. The value has to be $f > 120^\circ$ to get a better fit.

$\chi^2 = \sum (N_{\text{obs}} - N_{\text{mod}})^2 / \sigma^2 = 0.05$.
Figure 11: The cumulative size-frequency distributions computed by our Monte-Carlo collisional model of the two populations: the main belt (red), the Eos family (orange), together with the respective initial conditions (gray), and observations (black). At the time around \( t = 1300 \text{ Myr} \) the correspondence is good, except the tail below \( D \lesssim 2 \text{ km} \) where an observational incompleteness makes the SFD’s shallow. In particular, we successfully fit the knee of the family at \( D = 15 \text{ km} \).

fit both the asymmetry of the distribution with respect to \( a_c = 3.014 \text{ au} \), and the substantially lower number of bodies at mid-latitudes \( |\sin \beta| < 0.5 \). There are two systematics still present in our analysis, as we account neither for the observational selection bias, nor for the bias of the inversion method, but they should not overturn our conclusions.

Unfortunately, the uncertainty is larger than in the nominal model, because the number of bodies with known latitudes is limited, namely 46 within the family core. As a solution, we may use the distribution of \( |\beta| \) of [Cibulková et al. (2016)] which is available for many more asteroids, but we would need to determine the ’point-spread function’, describing a relation between input \( |\beta| \) and output \( |\beta| \) for this (approximate) method, which smears the distribution substantially. Their sample also contains a lot of bodies smaller than we had in the previous simulations, so we would have to compute everything again. This is postponed as a future work.

Acknowledgements

The work of MB has been supported by the Grant Agency of the Czech Republic (grant no. P209-18-04514J). In this paper, we used observations made by BlueEye 600 robotic observatory, supported by the Technology Agency of the Czech Republic (grant no. TA0301171). We thank the referees V. Carruba and F. Roig for their valuable input.

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Figure 12: The proper semimajor axis \( a_p \) vs the sine of pole latitude \( \sin \beta \) for the initial synthetic population (top panel), the evolved synthetic population (middle), and the observed population of 46 bodies (bottom). The individual bodies are shown as green dots, while their numbers in 8 boxes are indicated by the gray scale. The simulation started from initially isotropic random distribution, i.e. isotropic in \( \sin \beta \). The synthetic SFD was again scaled to the observed one. We account neither for the observational selection bias, nor for the bias of the inversion method. Nevertheless, it is possible to fit both the asymmetry of the distribution with respect to \( a_c = 3.014 \text{ au} \), and the substantially lower number of bodies at mid-latitudes \( |\sin \beta| < 0.5 \).
