

Main belt asteroids taxonomical information from dark energy survey data

V. Carruba ^{1,2*}, J. I. B. Camargo ^{2,3}, S. Aljbae ⁴, F. S. Ferreira ^{2,3}, E. Lin ⁵,
 V. Figueiredo-Peixoto ^{2,3,6}, M. V. Banda-Huarca ^{2,3}, A. Pieres ², R. C. Bouffeur ^{2,3}, L. N. da Costa ²,
 T. M. C. Abbott ⁷, M. Agüena ², Sahar S. Allam ⁸, O. Alves ⁵, P. H. Bernardinelli ⁹, E. Bertin ^{10,11},
 D. Brooks ¹², A. Carnero Rosell ^{2,13,14}, J. Carretero ¹⁵, M. E. S. Pereira ¹⁶, T. M. Davis ¹⁷,
 J. De Vicente ¹⁸, S. Desai ¹⁹, P. Doel ¹², I. Ferrero ²⁰, D. Friedel ²¹, J. Frieman ^{8,22},
 J. García-Bellido ²³, M. Gatti ²⁴, G. Giannini ¹⁵, D. Gruen ²⁵, R. A. Gruendl ^{21,26}, K. Herner ⁸,
 S. R. Hinton ¹⁷, D. L. Hollowood ²⁷, D. J. James ²⁸, S. Kent ^{8,22}, K. Kuehn ^{29,30}, O. Lahav ¹²,
 J. L. Marshall ³¹, J. Mena-Fernández ¹⁸, R. Miquel ^{15,32}, A. Palmese ³³, A. A. Plazas Malagón ^{34,35},
 M. Rodríguez-Monroy ³⁶, E. Sanchez ¹⁸, B. Santiago ^{2,37}, M. Schubnell ⁵, M. Smith ³⁸,
 E. Suchyta ³⁹, M. E. C. Swanson [†], G. Tarle ⁵, A. R. Walker ⁷, N. Weaverdyck ^{5,40} and
 P. Wiseman ³⁸ (DES Collaboration)

Affiliations are listed at the end of the paper

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ABSTRACT

While proper orbital elements are currently available for more than 1 million asteroids, taxonomical information is still lagging behind. Surveys like SDSS-MOC4 provided preliminary information for more than 100 000 objects, but many asteroids still lack even a basic taxonomy. In this study, we use Dark Energy Survey (DES) data to provide new information on asteroid physical properties. By cross-correlating the new DES data base with other data bases, we investigate how asteroid taxonomy is reflected in DES data. While the resolution of DES data is not sufficient to distinguish between different asteroid taxonomies within the complexes, except for V-type objects, it can provide information on whether an asteroid belongs to the C- or S-complex. Here, machine learning methods optimized through the use of genetic algorithms were used to predict the labels of more than 68 000 asteroids with no prior taxonomic information. Using a high-quality, limited set of asteroids with data on *gri* slopes and *i* – *z* colours, we detected 409 new possible V-type asteroids. Their orbital distribution is highly consistent with that of other known V-type objects.

Key words: catalogues – celestial mechanics – minor planets, asteroids: general.

1 INTRODUCTION

Currently, we know more than 1 million asteroids for which synthetic proper elements can be reliably obtained using the method of Knežević & Milani (2003). Unlike osculating elements, proper elements are constants of motion on time-scales of Myr, which allows for the identification of asteroid families. Our knowledge of the physical properties of asteroids is, however, much more limited. A full spectral classification of asteroids is available in various surveys for slightly more than 2000 objects (Bus & Binzel 2002; Lazzaro et al. 2004; DeMeo et al. 2009). Preliminary taxonomical information can be obtained from surveys like the Sloan Digital Sky Survey-Moving Object Catalogue data (SDSS-MOC4; Ivezić et al.

2001), using the method described in DeMeo & Carry (2013), for more than 100 000 asteroids. More recent works on the matter are those of Popescu et al. (2018) on the taxonomic classification of asteroids based on MOVIS near-infrared colours, and the new 3D machine learning classification scheme based on SDSS-MOC4 data of Roh et al. (2022). Yet, many objects, especially those at higher magnitudes and smaller diameters, lack any physical information. This limits studies on asteroid families, which are assumed to be mostly uniform in physical properties.

Asteroids can be classified into three main taxonomical groups based on their reflectance spectra. Asteroids belonging to the C-complex are typically dark in colour and have low albedos (reflectivity), while S-complex asteroids are typically brighter and more reflective. In the DeMeo & Carry (2013) taxonomy, the X class is divided into three classes, *E*, *M*, and *P*, which are distinguished solely by their albedo ($P < 0.075$, $0.075 < M < 0.30$, $E > 0.30$). V-type asteroids are characterized by a deep absorption band around

* E-mail: valerio.carruba@unesp.br

† Independent Researcher

1 micrometre, which is thought to be caused by the presence of the mineral olivine. They are thought to originate from the mantle of differentiated parent bodies.

The Dark Energy Survey [DES; Flaugher 2005, see also Dark Energy Survey Collaboration (2016)] is used here to provide additional information on a set of more than 60 000 asteroids. DES is a collaborative effort that covered 5000 deg² of the sky in the grizY bands from 2013 to 2019, primarily in the southern celestial sphere, aiming at investigating the dark energy. It is clear, however, its importance as also a Solar System survey. In fact, among its contributions, DES discovered and characterized a large and distant scattered disc object (Gerdes et al. 2017), improved predictions of stellar occultations by numerous TNOs and Centaurs (Banda-Huarca et al. 2019), made hundreds of discoveries in the TNO region (Bernardinelli et al. 2022), provided a detailed photometric analysis of a large sample of Jupiter Trojans (Pan et al. 2022), produced the largest TNO colour and light curve catalogue, facilitated the development of techniques to obtain optimal measurements of fluxes, colours, binarity, and variability for these slow-moving objects (Bernardinelli et al. 2023), and even discovered a messenger from the outskirts of the Solar System (Bernardinelli et al. 2021).

We used two sets of data from DES photometric measurements, a high-quality set with *gri* slopes and *i* – *z* colours, where the classification method of DeMeo & Carry (2013) can be applied, and a much larger (*g* – *r*, *g* – *i*) data base. We then cross-referenced the DES data with taxonomical, SDSS-MOC4, and albedo information, to understand how asteroid taxonomies are mapped in the new data set. Based on the distribution of known asteroid taxonomies, predictions on unlabelled bodies can then be made, using machine learning methods, optimized by the use of genetic algorithms (Chen, Wang & Lee 2004). Special attention is then given to more robust taxonomical classification for asteroid taxonomies that showed a good performance in DES data, such as the important V-type asteroids.

In Section 2, we describe how the data were obtained from the DES data base and organized. In Section 3, we describe the use of the DeMeo and Carry taxonomy. Studies using the (*g* – *r*, *g* – *i*) data are done in Section 4. Section 5 presents DES candidates for V-type asteroids, and conclusions are given in Section 6.

2 OBTAINING DES DATA

All colours presented in this study were obtained through observations and measurements conducted by DES.

To search for known Solar System objects, we queried the entire DES data base (Abbott et al. 2021) using keywords from image headers like pointing coordinates, date, and time of observations (Diehl et al. 2023), exposure time, and filter (Flaugher et al. 2015). We used an SQL-based tool called *easyaccess* (Carrasco Kind, Drlica-Wagner & Koziol 2018) for all queries of the DES data base.

Having those pieces of information in hand, we identify the single-epoch CCDs that could have captured the image of a small Solar System body using the Sky Body Tracker (SkyBoT; Berthier et al. 2006). SkyBoT, among other functionalities, yields a list of all known Solar System objects within a given field of view (FOV) when pointing coordinates, UTC date and time of observation, observing site coordinates, and FOV angular size are provided. We selected objects whose dynamical classification, as provided by the SkyBoT, were Hungaria, MB>Inner, MB>Middle, MB>Outer, MB>Cybele, MB>Hilda, or Jupiter Trojan, where MB stands for main belt.

We obtained positions, magnitudes, and other pieces of information of the selected objects from the Year-6 (Y6) list of objects the

Table 1. Known main belt, Hungaria and Trojan objects in DES images as indicated by the SkyBoT. Note that the number of objects effectively present in the images is smaller.

Class	Number of objects
Hungaria	12 199
Main Belt – Inner	102 979
Main Belt – Middle	134 277
Main Belt – Outer	125 823
Cybele	1 876
Hilda	1 751
Trojan	1 684

so-called Y6A1_FINALCUT_OBJECTS. The respective zero-points of each CCD were added to the magnitudes using the forward global calibration method (Burke et al. 2018), prepared by the collaboration (refer to sections 1 and 4.8 in Morganson et al. 2018, for more details about the finalcut catalogue).

Methods for improved photometry, applied to outer Solar System objects present in the images of DES, have been recently developed (Bernardinelli et al. 2023). However, the Y6 catalogue has features that make it an extremely attractive and valuable source to a variety of photometric studies, the one presented here in particular: (i) the same methods/procedures are used to derive flux measurements over the whole survey area, allowing us to coherently correlate colours of objects from different sky regions and (ii) readily available (upon password) data, thus saving a lot of CPU time, in addition to (iii) high quality single epoch photometry.

Table 1 shows the total number of asteroids that SkyBoT identified as belonging to a DES frame. We selected objects classified as Hungaria ($1.0 < a < 2.0$ au and $a(1 - e) > 1.666$ au), inner main belt ($2.0 \leq a < 2.5$ au), middle main belt ($2.5 \leq a < 2.82$ au), outer main belt ($2.82 \leq a < 3.27$ au), (main belt) Cybele ($3.27 \leq a < 3.7$ au), Hilda ($3.7 \leq a < 4.6$ au), and Trojan ($4.6 \leq a < 5.5$ au) according to the SkyBoT classification.¹ However, not all of them were detected, often due to a faint magnitude that is difficult to observe in a single-epoch image or to a very large (many degrees in some cases) positional uncertainty. A brief description of the task and tools used to find known Solar System objects in DES images can be found in the work by Banda-Huarca et al. (2019). All positions and magnitudes were obtained from the Y6 final cut DES catalogues and were queried using the *easyaccess* tool.

It is important to note that the DES observational cadence may not always be suitable for determining the colours of small objects, since observations of a same object in different filters may be separated by long periods of time and we do not have accurate enough rotational information to correct for rotational effects. As a result, we formed colours for a given object using only observations that were obtained within 10 min of each other (Abbott et al. 2021).

If we were to conduct an observational run specifically dedicated to measuring the colours of small objects, 10 min would be a reasonable estimate for obtaining magnitudes in different filters, taking into account the exposure time, readout, and filter change. However, a very fast rotator would likely require simultaneous observations in the different filters.

The ALCDEF (Asteroid Light Curve Data Exchange Format; see Stephens, Warner & Harris 2010; Warner, Stephens & Harris 2011; Stephens & Warner 2018) website displays a plot² indicating that

¹<https://vo.imcce.fr/webservices/skybot/?documentation>

²https://alcdef.org/php/alcdef_aboutLightcurves.html

Table 2. DES ($g - r$) colours for inner main belt asteroids. We report the standard deviations of measurements in ($g - r$), the number of colours recorded per object, and the total count of ($g - r$) measurements used to calculate the mean value. Additionally, we present the total count of objects with a ($g - r$) colour, regardless whether they were utilized in this study.

Stand. Dev in ($g - r$)	# of ($g - r$) colours per object	Total number of ($g - r$) colours used
0.041	1 to 3	20 825
0.058	4 to 6	403
0.059	7 to 9	30
0.064	10 to 12	32
0.051	13 to 15	7
0.028	16 to 18	10
0.044	19 to 21	5
0.042	22 or more	7

most asteroids have rotational periods longer than 2 h, with many falling between 4 and 10 h. If we consider the shortest period of 2 h, 10 min of observations corresponds to a rotation of 30 deg, while for a 7-h period, we have a rotation of only 8.6 deg. Therefore, our decision to group observations of the same object that were acquired within 10 min appears to be a good compromise between minimizing the impact of rotational effects on colours while keeping a sample as large as possible.

We can estimate the maximum error caused by neglecting pure rotational effects (i.e. considering a surface with homogeneous albedo) in our data base. If we simplify the asteroid light curves as a double-peaked triangle wave and assume a large amplitude, of 0.5 mag for a 4-h rotation period, a 10-min separation between observations corresponds to a magnitude change of approximately $10/(4 \times 60/4) \times 0.5 \simeq 0.083$. For a 10-h rotation period, the magnitude change would be only 0.033. Fast-rotating asteroids ($\simeq 4$ h period) would generally have smaller amplitudes because elongated objects are more prone to breaking (see fig. 9 in Chang et al. 2019). In most cases, the rotational effect on colour should be of the order of 10^{-2} , and should not exceed 0.1.

Albedo variations, however, do exist and have statistical significance on colour variations [see discussion in Szabó et al. (2004)]. Its overall impact can be inferred from table 2. The colour of a given object (e.g. $g - r$) is determined by the mean value of multiple measurements of that colour and its uncertainty is represented by the respective standard deviation. One might reasonably anticipate that colour uncertainties in colour depend on the number of measurements under the influence of a variable albedo (the smaller the average over a rotation, the more probable it is to obtain a large uncertainty). Table 2 shows that this is not the case. Based on this analysis, it can be concluded that albedo variations on the surface of asteroids are not likely to introduce significant systematic effects on our results.

Finally, in order to compare our data with known sources in the literature on asteroid taxonomy (see Sections 3 and 4), we transformed DES colours into SDSS (Sloan Digital Sky Survey) ones using the transformation equations from Abbott et al. (2021).

However, instead of using individual equations for each filter, we used their differences. In this context, to provide the necessary photometric information for Sections 3 and 4, we only needed the following colours from DES: ($g - r$)_{DES}, ($g - i$)_{DES}, ($r - i$)_{DES}, and ($i - z$)_{DES}. Solar apparent magnitudes and central wavelengths in the SDSS, where necessary, were taken from Willmer (2018). More

specifically, the DES to SDSS colour transformations used were

$$\begin{aligned}
 (g - r)_{\text{SDSS}} &= (g - r)_{\text{DES}} + 0.060 \times (g - i)_{\text{DES}} \\
 &\quad - 0.150 \times (r - i) - 0.019 \\
 (g - i)_{\text{SDSS}} &= 1.060 \times (g - i)_{\text{DES}} - 0.167 \times (r - i)_{\text{DES}} \\
 &\quad + 0.022 \\
 (i - z)_{\text{SDSS}} &= (i - z)_{\text{DES}} + 0.113 \times (r - i) - 0.003, \quad (1)
 \end{aligned}$$

where the first members of equations (1) are SDSS colours and colours in the second members are from DES. The variances (σ^2) in SDSS colours are given by the error propagation of these equations, added quadratically the root mean square of the respective band transformations from Abbott et al. (2021) used to determined each of the colour equations above.

3 DES DATA BASE: DEMEO AND CARRY TAXONOMY

As a preliminary step of our analysis, we selected asteroids from the DES data base for which both gri slopes and $i - z$ colours are both available. The colour indices contain information that can be used to derive a very low-resolution reflectance spectrum denoted as Rf . We calculate Rf using the equation $Rf = 10^{-0.4(M_f - M_{\odot f})}$, where M_f is the magnitude of the asteroid and $M_{\odot f}$ is the magnitude of the Sun in the same filter. To normalize Rf to 1 at a given wavelength we use the relationship $Rf, v = 10^{-0.4[(M_f - M_v) - (M_{\odot f} - M_{\odot v})]}$. The spectral gradient, or slope, is then given by [with λ in nanometres, see also Sabater et al. (2009)]:

$$S(\lambda_1, \lambda_2) = 10^4 \frac{Rf_{2,v} - Rf_{1,v}}{|\lambda_2 - \lambda_1|}. \quad (2)$$

In our case, v is the SDSS g filter.

This slope, as given above, is expressed as a percentage per 100 nanometres. To calculate the Rf, v value for each asteroid, we used the Sloan g , r , and i magnitudes and the effective wavelengths (see Willmer 2018) of each of these filters and then fit a straight line to the pairs (λ, Rf, v) . The angular coefficient from this adjustment gives the gri slope in per cent/100 nm.

These parameters are necessary to classify the asteroid taxonomies according to the method of DeMeo & Carry (2013), which is based on data from the SDSS-MOC4 (Ivezić et al. 2001). Our objective here is to identify asteroids present in both catalogues and to infer what taxonomical properties can be obtained from DES data. Given their temporal constraints, 17 154 asteroids with gri slopes and $i - z$ colours are present in the DES data base. To eliminate outliers, we only consider asteroids within the 0.15 to 0.85 quantile interval for both parameter distributions.³ This corresponds to intervals in gri slope between -34.738 and 43.466 , and in $i - z$ between -0.627 and 0.612 . Outliers are displayed as blue full transparent circles in Fig. 1. For statistical reasons, we will exclude outliers from our analysis, hereafter, yielding a sample of 17 135 asteroids, 10 685 of which are numbered, and the rest multi-opposition asteroids. Most of the outliers are data points beyond the range for which the DeMeo & Carry (2013) method applies, shown as a dashed blue box in Fig. 1, and most of them are not shown in Fig. 1. Therefore, removing these

³Quantiles are cut points that divide the range of a probability distribution into continuous intervals with equal probabilities. The median value of a distribution would correspond to a quantile of 0.50. More information on the procedure to compute quantiles can be found in Blitzstein & Hwang (2019).

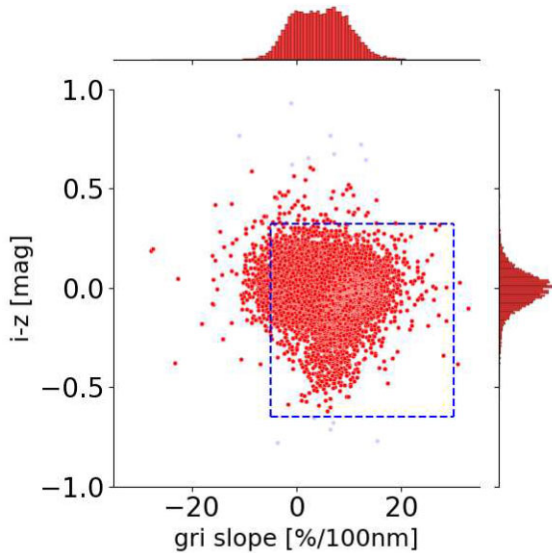


Figure 1. A joint plot of asteroids with available DES data, showing their (gri slope, $i - z$) values (transparent full circles). The full circles represent asteroids with values of the two parameters within the 0.15 to 0.85 quantile interval in the gri slope and $i - z$ colour distributions. The dashed box displays the interval for which the DeMeo & Carry (2013) classification scheme applies. The top and right side of the figure show histograms of the gri slope and $i - z$ colours distributions without outliers, respectively.

objects does not cause any significant loss of information.⁴ The range of absolute magnitudes for the data set without outliers goes from 9.0 to 20.5. On the top and right side of Fig. 1, we display histograms of gri slope and $i - z$ colours for the population of asteroids without outliers. While there is a single peak for the $i - z$ distribution, neither distribution is normal. If we compute the skewness, which is a measure of the symmetry or asymmetry of a distribution, and the kurtosis, which measures whether the data are heavy-tailed or light-tailed relative to a normal distribution, both distribution moments are significantly different from 0, which is the expected value for a Gaussian distribution. Both are skewed, one towards the right and one towards the left flank, with values of skewness of 0.14 and -0.53 , respectively. Both are leptokurtic distributions, with heavier tails than a normal distribution. The values of kurtosis are 3.31 and 6.45, respectively. These results show that neither of the distributions can be modelled as Gaussian distributions, and that more complex models should be used for this data.

Having obtained the data set of asteroids in the (gri slope, $i - z$) plane, it may be important to check how properties like the asteroids' absolute magnitudes may correlate with these data. While the size of an asteroid also depends on its albedo, objects with small absolute magnitudes tend to have larger sizes, and vice versa. In the main belt, more than 40 percent of the asteroid population belong to asteroid families (Milani et al. 2014), and the distribution of large and small objects tends to be not significantly different in proper element domains (Granvik et al. 2017). This is also observed in the

⁴Other quantile intervals were also considered. For a distribution within the 0.10 to 0.90 quantile interval only 2 outliers were found. This number increases to 42 if we consider distributions within 0.20 and 0.80. However, for such distributions, some data points within the range for which the classification method proposed by DeMeo & Carry (2013) applies were also excluded. For these reasons, we decided to work with the 0.15 to 0.85 quantile interval.

(gri slope, $i - z$) plane, as shown in the left panel of Fig. 2, where the distributions of large and small objects are fairly similar. While smaller objects cover a larger area than the larger ones, this is likely due to their greater number. Regions with a high number density of asteroids are inhabited by both small and large objects. Also, DES data observed fainter objects than previous surveys, like the SDSS-MOC4 or SDSS, for brevity (Ivezić et al. 2001). The right panel of Fig. 2 shows histograms of absolute magnitude distributions for the two surveys. Absolute magnitude H were obtained from the Asteroid Families Portal AFP [<http://asteroids.matf.bg.ac.rs/fam/index.php>], Radović et al. (2017), accessed on 2023 June]. Nominal errors on H are of the order of one decimal digit (Pravec et al. 2012).⁵ We expect that the new data on fainter objects observed by DES could provide new insights on the physical properties of small asteroids.

To begin, we aim at finding location of asteroids with known taxonomies in this new data set. For this purpose, we refer to the asteroid taxonomical data available in the surveys of Bus & Binzel (2002), Lazzaro et al. (2004), and DeMeo et al. (2009). For asteroids with more than one entry in the three surveys, we use a majority vote method to assign the most likely spectral type. There were 14 asteroids with taxonomical data in our selected DES data: 5 C-type, 3 S-type, and 6 X-type. Their location in the (gri , $i - z$) diagram is shown in Fig. 3. Apart from two X-type asteroid, there is no indication that the DES parameters are inconsistent with the classification scheme of DeMeo & Carry (2013). However, small number statistics prevent us from reaching more compelling conclusions. Most of the errors that we observed are in the horizontal axis, which are associated to errors in the gri slopes.

To increase the number of asteroids with taxonomical information, we turned our attention to the SDSS-MOC4 (Ivezić et al. 2001). Although the taxonomical information from the SDSS-MOC4 data has some limitations, as it is based on photometrical colours, it can still provide useful preliminary information on asteroids' physical properties. First, we eliminate asteroids with large errors in gri slope (error larger than 10 per cent/100nm) and $i - z$ colour (error larger than 0.1 mag). Then, we identify 950 asteroids for which the method of DeMeo & Carry (2013) can be used to obtain asteroid taxonomies. Results for the whole SDSS-MOC4 data are available at <https://sbn.psi.edu/pds/resource/sdsstax.html>. The identified asteroids include: 100 X-types, 54 D-types, 345 C-types, 151 L-types, 29 Q-types, 271 S-types, 5 A-types, and 32 V-types. We neglected subclasses like the CX, SQ, SV, LS, and QV, since we will show that DES data does not have the resolution needed to perceive these subtle differences. SV and QV objects were classified as V-type since they are all found in regions of the (gri slope, $i - z$) plane occupied by this class of objects. The left panel of Fig. 4 displays the position in the plane of $i - z$ colour versus gri slope for all these asteroids.

The classification obtained from DES data does not always agree with that from SDSS-MOC4, also because the usually large uncertainties on DES gri slopes. To further check the validity of the DES taxonomy, we also performed correlations with the Carvano et al. (2010) and Popescu et al. (2018) data sets. Table 3 presents the classification accuracy for asteroids in various spectral types for cross-correlations with the three data sets: C-types and V-types

⁵The same authors found that there is a systematic negative offset of absolute magnitudes in catalogues, which reaches a peak of -0.5 around $H = 14$. We are not correcting for these biases in this work, since our focus is on larger values of H . However, it is important to alert the reader to these biases for data around $H \simeq 14$.

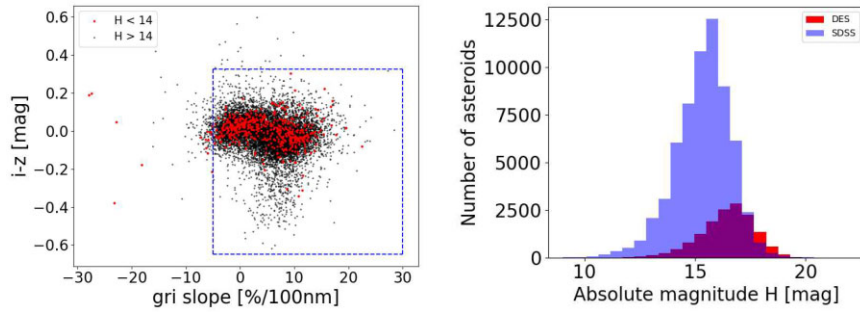


Figure 2. Left panel: a projection in the (gri slope, $i - z$) plane of DES large ($H < 14$) and small ($H > 14$) asteroids. The distribution of the two populations is not significantly different. In the right panel, we show the absolute magnitude distribution of asteroids in the SDSS-MOC4 data base and in the DES data set, for which gri slope and $i - z$ colours are available. The DES survey covers fainter objects than those reached by the SDSS one.

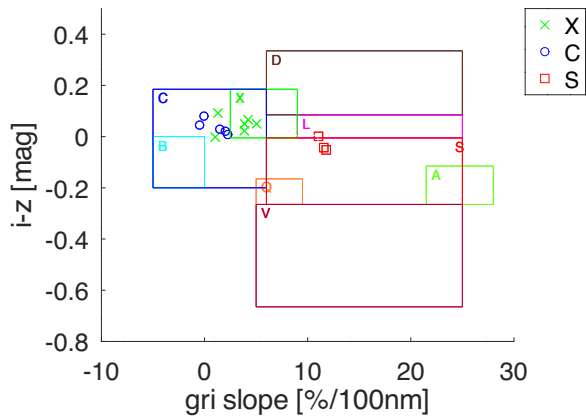


Figure 3. $i - z$ colours versus gri slope for asteroids with taxonomical data present in the DES data base. The coloured boxes represent the boundaries of the DeMeo & Carry (2013) taxonomical classes.

can be classified by DES data with purities, defined as the fraction of DES taxonomic labels correctly classified as such, as of 50 per cent or higher, while the other types have low purity percentages. Interestingly, 44.4 per cent of the misclassified V-type are found in the central and outer main belt, while no confirmed V-type has been found in these regions. This suggests that V-type candidates identified from DES data in these areas should be approached with caution. SV types can be easily misclassified as V-types using DES data, and none of the A-types were classified correctly. Based on this analysis, we propose a limited DES taxonomical classification consisting of three groups, the C- and S-complex, and the V-types. The C-complex will include X, D, and C-types, while the S-complex will encompass the remaining A, L, Q, K, and S-types. The V-types will remain as a separate class.

The new results for DES data are presented in the right panel of Fig. 4 and Table 4. The purities for both complexes are now above 75.0 per cent, suggesting that this simplified scheme could be applied more successfully. Following the analysis of Ivezić et al. (2002), we also plot in Fig. 5 the orbital distribution of C-complex, S-complex, and V-type objects as identified in both DES and SDSS-MOC4. Proper elements were obtained from AFP. As expected, C-complex asteroids are more common in the outer main belt, while S-complex asteroids are mostly found in the inner main belt, with some mixing of the two complexes. We can easily identify single asteroid families, like the S-complex Eunomia family in the central main belt, roughly at $a \simeq 2.6$ au, $\sin(i) \simeq 0.25$, or the Koronis family

in the pristine region, at $a \simeq 2.9$ au, $\sin(i) \simeq 0.05$. V-type asteroids are mostly associated with the Vesta family at $a \simeq 2.35$ au, $\sin(i) \simeq 0.12$, but are also found in the central and outer main belt. Because of the importance of V-type asteroids for early scenarios of the Solar System formation, we will further discuss their orbital distribution in Section 5.

Finally, we searched for objects in the DES data set with albedo data in the WISE and NEOWISE, AKARI, or IRAS databases (Ryan & Woodward 2010; Usui et al. 2011; Masiero et al. 2012), and we identified 1573 asteroids, as shown in Fig. 6. Albedo data correlates rather well with the position of asteroids in the plane of $i - z$ colour versus gri slope: 81.5 per cent of objects identified as C-complex have values of $p_V < 0.12$, and 94.6 per cent of S-complex asteroids (we include V-type asteroids as S-complex bodies for albedo purposes) have $p_V > 0.12$, as expected for these bodies. The lower value of C-complex asteroids with low albedo data, in contrast to the higher percentage of S-complex asteroids with high albedos can be attributed to the fact that C-complex asteroids include X-type, which, as discussed in Section (1), also include the M- and E-types, which have high albedos. Table 5 summarizes our findings.

What predictions can be made based on these results? Based on the results of the SDSS-MOC4 analysis, we can train machine learning (ML) algorithms. In order to select the best-performing ML methods and the combination of its free parameters, or hyperparameters, that work best for our data set, we use genetic algorithms (Chen, Wang & Lee 2004), according to the procedure described in Carruba, Aljbaae & Domingos (2021). We run the genetic algorithm procedure on a subset of asteroids with both DES and SDSS-MOC4 data three times, using a validation set corresponding to 20 per cent of the training data. The use of a validation set is recommended to avoid the issue of overfitting, which happens when the model is overly sensitive to the fine details of the training set, but may perform poorly when dealing with other sets of data. The best-performing algorithm was a Gaussian Naive Bayes (GNB) estimator (Chan, Golub & LeVeque 1979). GNB is a classification technique used in ML based on the probabilistic approach and Gaussian distribution. GNB assumes that each parameter (also called features or predictors) has an independent capacity of predicting the output variable. The combination of the prediction for all parameters is the final prediction, that returns a probability of the dependent variable to be classified in each group. The final classification is assigned to the group with the higher probability. Finally, since we are dealing with a multiclassification problem, the use of a f_1 score, defined as a harmonic of precision and recall ($f_1 = 2 \cdot \frac{\text{Precision} \cdot \text{Recall}}{\text{Precision} + \text{Recall}}$), is more appropriate to correctly estimate the efficiency of the algorithm,

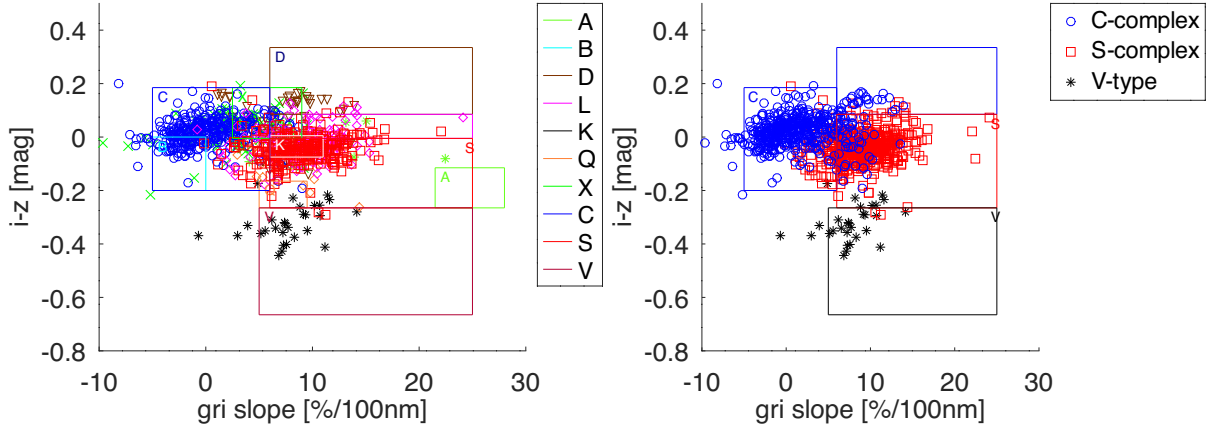


Figure 4. $i - z$ colour versus gri slope of asteroids with data in both the SDSS-MOC4 and DES data base. The boundaries of the DeMeo & Carry (2013) taxonomical classes are identified by the coloured boxes. The left panel displays the positions of these asteroids for all DeMeo & Carry (2013) available classes, while the right panel shows a simplified classification scheme that only identifies C-complex, S-complex asteroids, and V-type objects.

Table 3. Percentage of consistent taxonomic classifications, or purity, for asteroids in both the DES and SDSS-MOC4, the DES, and the Carvano et al. (2010), and the DES and the MOVIS data bases.

Ast. type	# of asteroids	Percentage of SDSS-MOC4 cons. classification	# of asteroids	Percentage of Carvano et al. (2010) cons. classification	# of asteroids	Percentage of MOVIS cons. classification
X	100	41.0	91	40.7	9	22.2
D	54	37.0	40	30.0	8	12.5
C	345	63.8	335	64.2	21	66.7
A	5	0.0	5	0.0	65	0.0
L	151	19.2	150	18.8	28	32.0
Q	29	34.5	33	30.3	0	–
S	268	38.8	270	38.5	55	34.5
V	32	65.6	25	64.0	21	52.4

Table 4. Percentage of consistent taxonomic classifications (purities) for asteroids in both the DES and SDSS-MOC4 data bases for a revised taxonomical scheme.

Asteroid complex	# of asteroids	Percentage of cons. classification
C	499	94.2
S	454	80.0
V	32	65.6

rather than the more commonly used accuracy. The model had an f_1 score of 91.6 per cent.

Fig. 7 displays the results of a prediction for 16516 asteroids with no previous labels. Predictions were made separately for numbered and multi-opposition asteroids, using the same training data. The diagonal line in Fig. 7 is caused by the presence of D-type objects that extends the boundaries of C-complex upwards and rightwards. In the ML model, this causes the boundary between C-complex and S-complex to be a diagonal line. Our classifier predicts that 10213 of the asteroids are likely to be C-complex, 5890 are likely S-complex asteroids, and 410 are new possible V-type objects.

4 DES DATA BASE: (G-R, G-I) PLANE

The number of asteroids with DES data increases significantly if we consider the $g - r$ and $g - i$ colours instead of the $i - z$ and gri slope, passing from 17 078 to 61493. Their distribution in the ($g - r$, $g -$

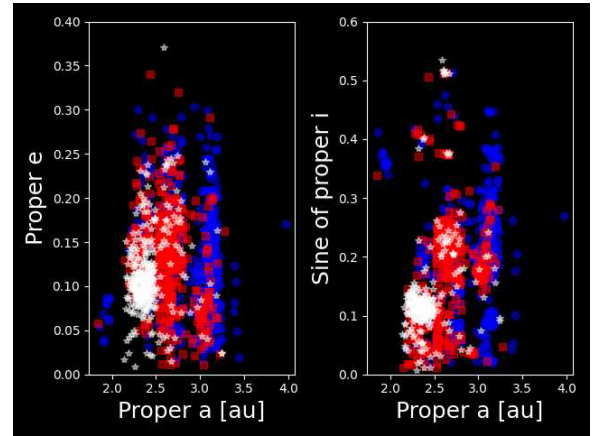


Figure 5. Projections in the proper (a , e) and proper (a , $\sin(i)$) of the C-complex (full circles), S-complex (full squares), and V-type objects (asterisks), identified in both the DES and SDSS-MOC4 data bases, and as listed in Table 4. Proper elements were obtained from *AFP*.

i) plane is shown in Fig. 8, with g , r , i being the photometric bands of the SDSS *'ugriz'* system. While some of the brightest objects can have negative values of these colours, most asteroids are found in an interval from 0.2 to 1.2 in both quantities. To eliminate these outliers, we again only consider asteroids within an interval of quantiles of 0.15 to 0.85 for both colour distributions, which corresponds to the interval $-0.266 < g - r < 1.393$ and $-0.326 < g - i < 1.746$.

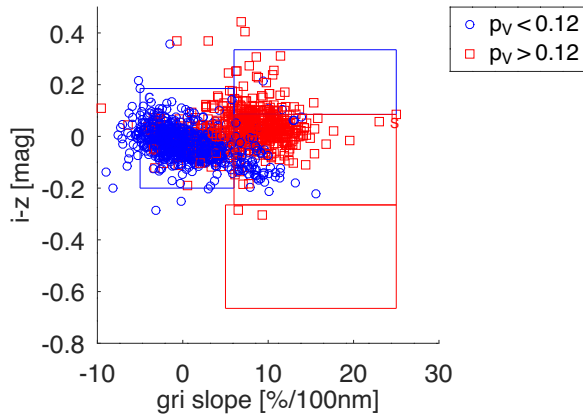


Figure 6. $i - z$ colour versus gri slope for asteroids with albedo data and present in the DES data base.

Table 5. Percentage of asteroids with C and S-complexes classifications with geometric albedo $p_v < 0.12$ and $p_v > 0.12$, respectively.

Asteroid complex	# of asteroids	Percentage of albedo values
C	1188	81.5
S	385	94.6

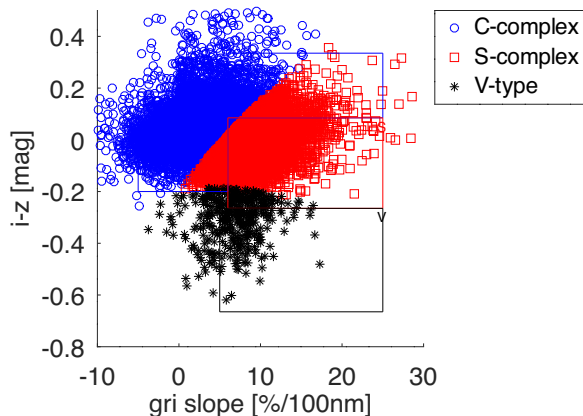


Figure 7. $i - z$ colour versus gri slope of the ML predicted data for asteroids in the DES data base with no SDSS-MOC4 data.

Outliers are displayed as blue full circles in the figure, while red full circles show the positions of the rest of the data set. We will exclude outliers from our analysis, henceforth. After removing the outliers, we ended up with a data set of 61 142 asteroids. Fig. 8 shows a joint plot and histograms of $g - r$ and $g - i$. Both distributions are single-peaked, slightly skewed, with skewness of -0.17 and -0.16 , respectively, and leptokurtic, with kurtosis values of 3.37 and 3.08.

First, we investigate where asteroids in this new data set with known taxonomies could be located. Our results are shown in Fig. 9. We identified 71 asteroids with taxonomical data: 4 B-, 14 C-, 3 D-, 15 X-, 29 S-, 2 L-, and 3 V-types. B, C, D, and X asteroids are darker objects, associated with the so-called C-complex, while the other types are brighter objects associated with the S-complex. Our data show that the two complexes are fairly separated in this domain, with C-complex asteroids located in the left part of the ellipsoidal distribution of $(g - r, g - i)$ values, and S-complex asteroids concentrating in the right part. However, different taxonomies inside

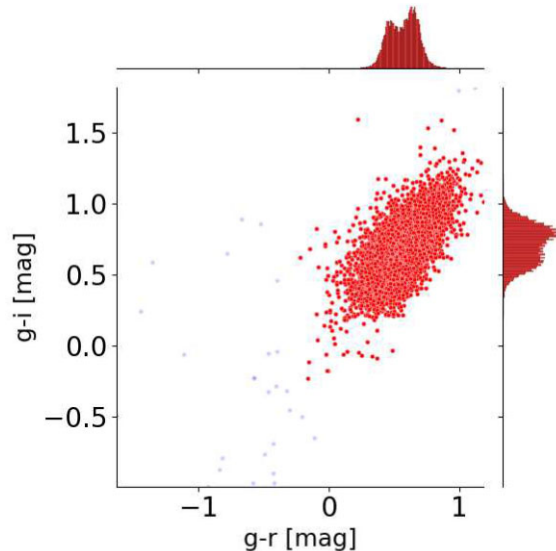


Figure 8. A joint plot of the $(g - r, g - i)$ colours for asteroids with available DES data. The blue full transparent circles indicate outlier data in the distributions of both colours. The histograms of the two colours distributions, without the outliers, are displayed on the top and right sides, respectively.

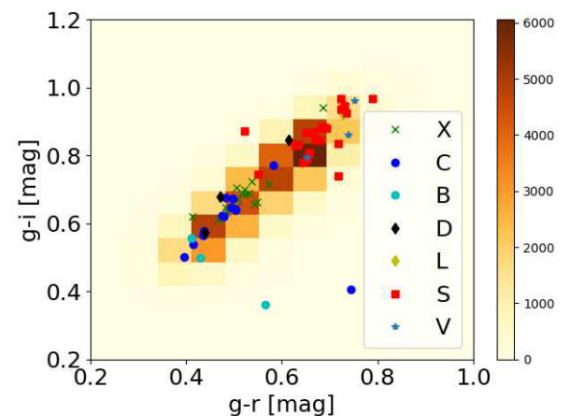


Figure 9. A 2D histogram in the $(g - r, g - i)$ plane of the asteroids with DES data. The symbols are objects for which taxonomical information is available in the Bus & Binzel (2002), Lazzaro et al. (2004), and DeMeo et al. (2009) surveys, according to the code shown in the legend.

these complexes overlap with each other. While DES data could help discriminate between C- and S-complexes, we do not have enough resolution in the $(g - r, g - i)$ domain to perform a more in-depth taxonomical analysis, including the V-type asteroids.

To further confirm this hypothesis, we use the method of DeMeo & Carry (2013) to obtain taxonomical information for asteroids listed in the SDSS-MOC4 (Ivezić et al. 2001). Results for the Carvano et al. (2010) and Popescu et al. (2018) data sets are similar and will not be presented or discussed for the sake of brevity. We identify 3347 asteroids having data in both data bases, of which 1514 are within the C-complex, and 1833 are within the S-complex. Fig. 10 shows the $(g - r, g - i)$ distribution of these objects, which confirms the results from the spectroscopical surveys data: there is a clear separation between C-complex and S-complex asteroids in this domain. Again, the data resolution is not sufficient to distinguish between different asteroid types, as previously observed, and this is not shown in Fig. 10 for simplicity.

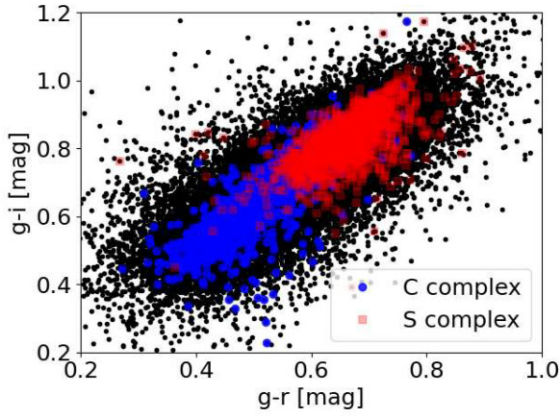


Figure 10. Projection in the $(g - r, g - i)$ plane of the asteroids with DES data and SDSS-MOC4 taxonomies. Circles identify asteroids likely to belong to the C-complex, while squares display asteroids with more likely to belong to the S-complex. Points are asteroids with no SDSS-MOC4 data.

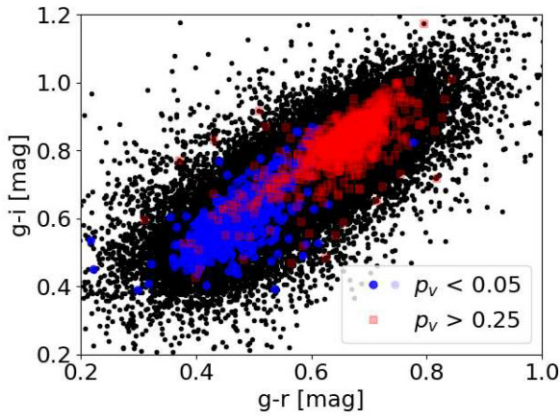


Figure 11. Projection in the $(g - r, g - i)$ plane of the asteroids with DES data and albedo values. Circles identify asteroids with albedo p_V lower than 0.05, while squares display asteroids with $p_V > 0.25$.

Finally, while the distribution of albedo values may vary among a single spectral type, C-complex asteroids tend to have lower values of geometric albedos p_V , while S-complex ones have larger p_V . We searched for objects in the DES data set with albedo data in the *WISE* and *NEOWISE*, *AKARI*, or *IRAS* data bases, and identified 5122 asteroids, as shown in Fig. 11. While asteroids with intermediate albedos ($0.05 < p_V < 0.25$) cover the $(g - r, g - i)$ domain more or less uniformly, very dark ($p_V < 0.05$), and very bright ($p_V > 0.25$) asteroids are quite separated in the $(g - r, g - i)$ domain, which confirms the trends observed for taxonomic data.

Using taxonomic information obtained from the SDSS analysis, we utilized genetic algorithms to determine the optimal ML methods for predicting the complex type of asteroids based on the $(g - r, g - i)$ DES data. Our analysis found that the Linear Support Vector Classifier (Linear SVC; Cortes & Vapnik 1995) performed the best.

Linear SVC is a type of machine learning algorithm used for classification tasks, where the goal is to assign each input to one of a set of predefined categories or classes. Linear SVC is a variant of the support vector machine (SVM) algorithm that uses a linear kernel function. In Linear SVC, the algorithm tries to find the hyperplane that separates the different classes with the largest possible margin. The hyperplane is defined as the set of points in the feature space

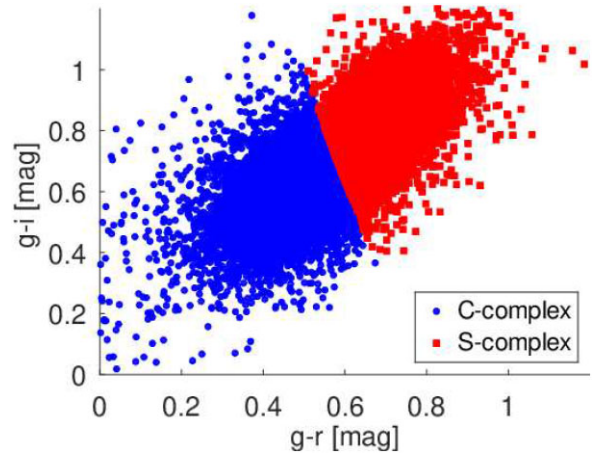


Figure 12. $(g - r, g - i)$ colours of the ML predicted data for asteroids in the DES data base.

where the decision boundary between classes lies. The margin is the distance between the hyperplane and the closest points from each class. The larger the margin, the more robust the classification will be to noise and outliers.

For our model, we used a C parameter, which controls the trade-off between achieving a low training error and a low testing error, of 0.1. a penalty parameters equal to $1/l$, which uses the $l1$ regularization method, and a tolerance parameter, which specifies the tolerance for stopping the optimization algorithm, of 0.0001. Other parameters were the standard choices for the Linear SVC algorithm. Our model achieved an accuracy of 90.9 per cent on the validation set, which, as in the previous section, was 20 per cent of the original training set.

Fig. 12 shows predictions for 58 118 new asteroids with no prior complex information. Our analysis suggests that 28871 of these objects belong to the C-complex, while 29 247 are more likely to be S-complex asteroids. The higher fraction of S-complex asteroids, as opposed to C-complex in the $(g - r, g - i)$ domain with respect to the $(gri$ slope, $i - z$) plane can be explained by the fact that, in the $(g - r, g - i)$ domain V-type asteroids are considered as part of the S-complex.

5 V-TYPE ASTEROIDS: DES CANDIDATES

In Section 3, we observed that the accuracy of predicting V-type objects in the gri slopes and $i - z$ colour plane was higher (65.6 per cent) than for other classes. V-type objects are important because of their association with differentiated parent bodies and basaltic composition. Identifying their orbital location and physical properties may provide insights about the early phases of our Solar System formation. To date, asteroid 4 Vesta is the only confirmed differentiated body in the main belt, but other possible differentiated parent bodies have been suggested in the past. Using the DES data in the gri slopes and $i - z$ colour plane, we identified 410 new potential V-type objects with available proper elements from the Asteroid Family Portal AFP, <http://asteroids.matf.bg.ac.rs/fam/> (Novaković et al. 2022), 85 of which are located in the central and outer main belt.

Fig. 13 displays their projection onto a proper $(a, \sin(i))$ plane. The dynamical evolution of V-type objects suggests that there may be six possible regions where injected material, either from a local or remote source, can evolve due to non-gravitational forces. Carruba et al.

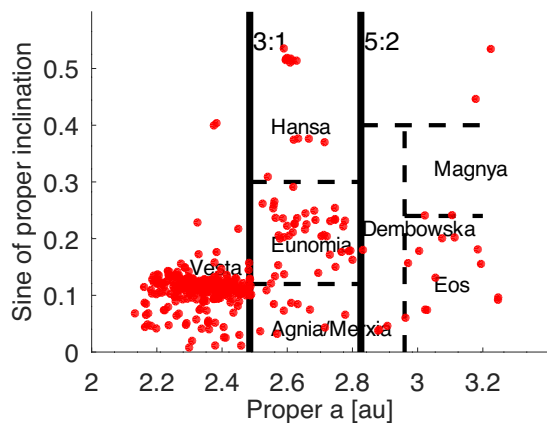


Figure 13. A proper (a , $\sin(i)$) projection of the V-type candidate asteroids identified using DES data. Vertical full lines identify the dynamical boundaries of the inner, central, and outer main belt. The dashed lines show the location of the dynamical regions associated with possible local sources of V-type materials, as identified by Carruba et al. (2014) and Huaman, Carruba & Domingos (2014).

(2014) identified three such regions in the central main belt, named after three possible local differentiated bodies: Hansa, Eunomia, and the Agnia/Merxia parent body. A similar analysis by Huaman, Carruba & Domingos (2014) found three comparable dynamical regions in the outer main belt: the Dembowska, Eos, and Magnya regions. DES V-type candidates are primarily located where they are expected, near the Vesta family and in the densely populated Eunomia and Eos regions. Further investigation of their physical properties and understanding their dynamical evolution will remain a challenge for future studies.

6 SUMMARY AND CONCLUSIONS

In this study, we aimed to explore the feasibility of using DES data to infer physical properties of main belt asteroids. We initially focused on using the gri slopes and $i - z$ colours of asteroids, as this plane allows us to employ the taxonomic classification scheme of DeMeo & Carry (2013). After removing outliers, we identified a population of 17 135 asteroids. However, with the exception of V-type and possibly C-type asteroids, DES data in this plane are insufficient to identify all the DeMeo & Carry (2013) taxonomies. None the less, we can still distinguish between C- and S-complex taxonomies. We utilized machine learning approaches, optimized using genetic algorithms, to predict complex labels for 10 213 asteroids that previously had no taxonomic information.

A much larger sample of objects is available if we consider the $(g - r, g - i)$ plane, with a sample of 61 493 asteroids with such data obtained after outlier removal. We employed machine learning algorithms to predict the complex labels of 58 118 new asteroids with no prior taxonomic information, using the asteroids with SDSS-MOC4 known taxonomies as a training sample.

Lastly, we used DES data to identify 410 possible new V-type objects. Their distribution in the proper (a , $\sin(i)$) domain is consistent with the location of known V-type bodies. Future research could concentrate on further examination of the physical characteristics and dynamical evolution of the asteroids identified in this study.

CODE AVAILABILITY

All codes are available from the authors, upon reasonable request.

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DATA AVAILABILITY

The taxonomical complexes data, as well as the data on DES V-type candidates, are available at the NASA Planetary Data System (PDS): Carruba, V., Camargo, J. I. B., Aljbaae, S, and the Dark Energy Survey Team (2024). Taxonomy, colours, and slope parameters for asteroids from the Dark Energy Survey V1.0. urn:nasa:pds:gbo.ast.des.taxonomy::1.0. NASA Planetary Data System; <https://doi.org/10.26033/m95p-bn08>.

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SUPPORTING INFORMATION

Supplementary data are available at *MNRAS* online.

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¹School of Natural Sciences and Engineering, São Paulo State University (UNESP), 12516.410-Guaratingueta-SP, Brasil.

²Laboratório Interinstitucional de e-Astronomia - LIneA, 20765-000-Rio de Janeiro-RJ, Brazil

³Observatório Nacional, Rua Gal. José Cristino 77, Rio de Janeiro RJ-20921-400, Brazil

⁴National Space Research Institute (INPE), Division of Space Mechanics and Control, C.P. 515, São José dos Campos SP-12227-310, Brazil

⁵Department of Physics, University of Michigan, Ann Arbor, MI 48109, USA

⁶IGeo - Universidade Federal do Rio de Janeiro, Av. Athos da Silveira Ramos 274, Rio de Janeiro RJ-21941-916, Brazil

⁷Cerro Tololo Inter-American Observatory, NSF's National Optical-Infrared Astronomy Research Laboratory, Casilla 603, La Serena, Chile

⁸Fermi National Accelerator Laboratory, P. O. Box 500, Batavia, IL 60510, USA

⁹Astronomy Department, University of Washington, Box 351580, Seattle, WA 98195, USA

¹⁰CNRS, UMR 7095, Institut d'Astrophysique de Paris, F-75014 Paris, France

- ¹¹*Sorbonne Universités, UPMC Univ Paris 06, UMR 7095, Institut d'Astrophysique de Paris, F-75014 Paris, France*
- ¹²*Department of Physics and Astronomy, University College London, Gower Street, London WC1E 6BT, UK*
- ¹³*Instituto de Astrofísica de Canarias, E-38205 La Laguna, Tenerife, Spain*
- ¹⁴*Universidad de La Laguna, Dpto. Astrofísica, E-38206 La Laguna, Tenerife, Spain*
- ¹⁵*Institut de Física d'Altes Energies (IFAE), The Barcelona Institute of Science and Technology, Campus UAB, E-08193 Bellaterra (Barcelona), Spain*
- ¹⁶*Hamburger Sternwarte, Universität Hamburg, Gojenbergsweg 112, D-21029 Hamburg, Germany*
- ¹⁷*School of Mathematics and Physics, University of Queensland, Brisbane QLD 4072, Australia*
- ¹⁸*Centro de Investigaciones Energéticas, Medioambientales y Tecnológicas (CIEMAT) 28040, Madrid, Spain*
- ¹⁹*Department of Physics, IIT Hyderabad, Kandi, Telangana 502285, India*
- ²⁰*Institute of Theoretical Astrophysics, University of Oslo. P.O. Box 1029 Blindern, NO-0315 Oslo, Norway*
- ²¹*Center for Astrophysical Surveys, National Center for Supercomputing Applications, 1205 West Clark St., Urbana, IL 61801, USA*
- ²²*Kavli Institute for Cosmological Physics, University of Chicago, Chicago, IL 60637, USA*
- ²³*Instituto de Física Teórica UAM/CSIC, Universidad Autónoma de Madrid, E-28049 Madrid, Spain*
- ²⁴*Department of Physics and Astronomy, University of Pennsylvania, Philadelphia, PA 19104, USA*
- ²⁵*University Observatory, Faculty of Physics, Ludwig-Maximilians-Universität, Scheinerstr. 1, D-81679 Munich, Germany*
- ²⁶*Department of Astronomy, University of Illinois at Urbana-Champaign, 1002 W. Green Street, Urbana, IL 61801, USA*
- ²⁷*Santa Cruz Institute for Particle Physics, Santa Cruz, CA 95064, USA*
- ²⁸*Center for Astrophysics, Harvard and Smithsonian, 60 Garden Street, Cambridge, MA 02138, USA*
- ²⁹*Australian Astronomical Optics, Macquarie University, North Ryde, NSW 2113, Australia*
- ³⁰*Lowell Observatory, 1400 Mars Hill Rd, Flagstaff, AZ 86001, USA*
- ³¹*George P. and Cynthia Woods Mitchell Institute for Fundamental Physics and Astronomy, and Department of Physics and Astronomy, Texas A&M University, College Station, TX 77843, USA*
- ³²*Institució Catalana de Recerca i Estudis Avançats, E-08010 Barcelona, Spain*
- ³³*Department of Physics, Carnegie Mellon University, Pittsburgh, PA 15312, USA*
- ³⁴*Kavli Institute for Particle Astrophysics and Cosmology, P. O. Box 2450, Stanford University, Stanford, CA 94305, USA*
- ³⁵*SLAC National Accelerator Laboratory, Menlo Park, CA 94025, USA*
- ³⁶*Laboratoire de Physique des 2 Infinis Irène Joliot-Curie, CNRS Université Paris-Saclay, Bât. 100, Faculté des sciences, F-91405 Orsay Cedex, France*
- ³⁷*Instituto de Física, UFRGS, Caixa Postal 15051, Porto Alegre, RS - 91501-970, Brazil*
- ³⁸*School of Physics and Astronomy, University of Southampton, Southampton SO17 1BJ, UK*
- ³⁹*Computer Science and Mathematics Division, Oak Ridge National Laboratory, Oak Ridge, TN 37831, USA*
- ⁴⁰*Lawrence Berkeley National Laboratory, 1 Cyclotron Road, Berkeley, CA 94720, USA*

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