Limits on the Detection of Planet Nine in the Dark Energy Survey

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ABSTRACT

Studies of the clustering of the most distant Kuiper belt objects in the outer solar system have hinted at the possible existence of a planet beyond Neptune referred to as Planet Nine (P9). Recent efforts have constrained the parameter space of the orbital elements of P9, allowing for the creation of a synthetic catalog of hypothetical P9s. By examining the potential recovery of such a catalog within numerous sky surveys, it is possible to further constrain the parameter space for P9, providing direction for a more targeted search. We examine the ability of the full six years of the Dark Energy Survey (DES) to recover a synthetic Planet Nine population presented in Brown & Batygin (2021a). We find that out of 100,000 simulated objects, 11,709 cross the wide DES survey footprint of which 10,187 (87.0\%) are recovered. This rules out an additional 5\% of the parameter space after accounting for Planets Nine that would have been detected by both the Zwicky Transient Facility and DES.

1. INTRODUCTION

The discovery of distant Kuiper belt objects (KBOs) on orbits that avoid Neptune’s gravitational perturbations, starting with Sedna by Brown et al. (2004), has spurred interest in scenarios of their origin and dynamical evolution. After the discovery of the Sednoid 2012 VP\textsubscript{113}, Trujillo & Sheppard (2014) pointed out an apparent clustering of the argument of perihelion of these most distant planets around zero and speculated that this clustering could be caused by a Kozai resonance with a distant planet, though no dynamical configuration that would cause such clustering could be found. Reexamining these objects, Batygin & Brown (2016) realized that the distant objects unaffected by Neptune are clustered in longitude of perihelion and pole position, rather than argument of perihelion. This clustering was found to be the natural outcome of a massive planet on an eccentric inclined orbit beyond Neptune.

Multiple lines of evidence now exist for the existence of a giant planet – here called Planet Nine – on a distant eccentric and inclined orbit, including the clustering in longitude of perihelion and pole position of the highest semimajor axis KBOs, the existence of a substantial detached population of KBOs with perihelia removed far from the planetary region, and the abundance of Centaurs on orbits essentially perpendicular to the plane of the solar system (Batygin et al. 2019). While alternative explanations have been invoked to explain subsets of the phenomena (Morbidelli & Levison 2004; Gladman et al. 2009; Shankman et al. 2017; Napier et al. 2021; Sefilian & Touma 2019; Scholtz & Unwin 2020), the existence of Planet Nine remains the simplest hypothesis unifying these disparate observations. Under the hypothesis that Planet Nine does exist, Brown & Batygin (2021a) sought to provide direction for more targeted searches. The work combined information on observational biases with a likelihood model to constrain the orbital elements and mass of P9. Using the full orbital elements of 11 distant KBOs with \( q > 42\text{AU} \) and associated observational biases, Brown & Batygin (2021a) write a likelihood function for detecting each KBO with a certain \((i, \varpi, \Omega)\) with known fixed semi-major axis and eccentricity for each KBO, given a Planet Nine with a certain mass, semi-major axis, and orbital parameters. The product of these 11 likelihoods is then the full likelihood model for Planet Nine. From each of these likelihoods, a probability distribution for \((i, \varpi, \Omega)\) assuming an isotropic distribution of KBOs was derived, and then multiplied by a biased probability distribution for each object. These probabilities were then converted into a final likelihood for Planet Nine, which was sampled using a Gaussian Process Emulator and Markov Chain Monte Carlo (MCMC) with two different prior choices, producing 49,100 uncorrelated samples. These samples capture the likely range of orbital param-
Figure 1. The synthetic Planet Nine population overlaid on top of the DES footprint. Blue dots represent the location of each member of the reference population on June 1st 2018. The yellow dots represent the individual detections of all members of the reference population which intersect the DES footprint. In red is the outline of the DES footprint, with the black dashed line tracing the ecliptic plane.

eters within which Planet Nine may be located, with a mass range of $m_9 = 6.2^{+2.2}_{-1.3}$ and semi-major axis range of $a_9 = 380^{+140}_{-80}$ (Brown & Batygin 2021b). In order to aid the search for Planet Nine, this posterior sample for the parameter space was then sampled 100,000 times to create a synthetic catalog in Brown & Batygin (2021b). Effectively, each member of this catalog has a set of orbital elements that are randomly sampled from the parameter space derived in Brown & Batygin (2021a), along with a randomly assigned mean anomaly that places the simulated object in a specific place in its orbit. This catalog is therefore a reference population for P9, allowing it to be tested against existing surveys. Using radii derived from the sampled masses of these objects, the visual magnitudes of the synthetic P9s are found to range from 18 to 26, and thus are detectable by a multitude of sky surveys, such as Pan-STARRS, the Zwicky Transient Facility (ZTF), and Dark Energy Survey (DES). The ZTF search for Planet Nine in Brown & Batygin (2021b) constrained 56% of the parameter space for the orbital elements of P9 by detecting 56,173 members of the synthetic catalog injected into the ZTF survey.

The Dark Energy Survey is one of the deepest multi-epoch wide-field surveys to date and thus can provide useful constraints on the existence of Planet Nine. DES used the Dark Energy Camera (Flaugher et al. 2015) on the 4-meter Blanco Telescope in Cerro Tololo for 575 nights (2013–2019) to image a contiguous 5000 deg² patch of the Southern sky in five bands ($grizY$) over six years. Typical fields in the Wide Survey received around ten 90 second visits in each band in the 6 years, with typical limiting magnitudes of 23.8 in $r$ band. Although designed to probe the distribution of dark matter in the universe (DES Collaboration et al. 2021), the survey strategy has been proven to be capable of detecting hundreds of KBOs (The Dark Energy Survey Collaboration et al. 2016; Khain et al. 2020; Bernardinelli et al. 2020, 2022). The final search for KBOs in DES (Bernardinelli et al. 2022) yielded 814 objects, with magnitudes of 50% completeness at around 24.71 mag in DES $g$-band and a well characterized detection efficiency. The northern-most latitudes that DES has imaged intersect with the simulated population generated in Brown & Batygin (2021b), allowing the survey to further constrain the P9 parameter space. With the public release of both the DES survey simulator DESTNSIM and the accompanying Y6 data (Bernardinelli et al. 2022), a natural follow-up to Brown & Batygin (2021b) is to test DES for its ability to recover the simulated Planet Nine catalog, as the DES magnitude limits suggest most
members of the reference population can be recovered provided they intersect the DES footprint.

In this paper, we inject all 100,000 of the individual objects from the synthetic catalog of Brown & Batygin (2021b) into the survey simulator for DES from Bernardinelli et al. (2022) for the full six years of the Dark Energy Survey to determine which parts of the parameter space for Planet Nine DES can exclude. We present the main results with assumptions for color and albedo identical to those of BB21. To supplement this analysis, we examine how different hypotheses on the potential surface type and composition of P9 can impact our ability to detect it. Finally, we demonstrate the effect of the combined results from DES and ZTF on the P9 parameter space.

2. SIMULATED PLANET NINE POPULATION IN DES

We use the P9 reference population created in (Brown & Batygin 2021b, hereafter BB21) to determine what further constraints DES can place on P9. In order to inject the simulated population into the survey, we must first make certain assumptions regarding the albedo, radius-mass relation, and colors of the reference population. For the main analysis, we adopt the same parameters from BB21. They used a uniform prior on albedo ranging from 0.2 to 0.75, with the lower end being half of Neptune’s albedo, and the higher end derived from pure Rayleigh scattering based on Fortney et al. (2016). The radius of any given synthetic object was derived from its mass, estimated by \( r_9 = r_{\text{earth}}(m_9/3M_{\text{earth}}) \). This relationship for the radius and mass comes from a best fit for Kepler data, which holds well in the mass range of Neptune (Wu & Lithwick 2013). As DES observes in the griz bands (note that Bernardinelli et al. 2022 did not use the DES Y-band in their search), we must also make assumptions about the colors of the object. Using g-band as the reference band, we choose multiple models for the surface of Planet (listed in Table 1) to obtain a set of three colors. The fiducial model uses solar colors (Willmer 2018) derived from the assumption that P9’s atmosphere will be a perfect Rayleigh scattering atmosphere (Fortney et al. 2016). This fiducial set of colors is also selected in order to match the colors assumed in the analysis from Brown & Batygin (2021b). The other four models chosen for the analysis are the model presented in Fortney et al. (2016) assuming a temperature of 40K and 10% methane abundance, a Neptune-like P9, a “super-Ganymede”, and a “super-KBO”.

2.1. Generating Synthetic Observations in DES

We begin by determining which members of the synthetic population will intersect the DES footprint. Using the publicly available DES survey simulation software, DESTNOSIM\(^1\) (Bernardinelli et al. 2022), we determine the detectability of our population of synthetic objects. The software accomplishes this task by finding which DES exposures intersect the object’s orbit, followed by determining the likelihood of detecting the object individually in each exposure, and then checking against the recoverability of objects in the linking process. Bernardinelli et al. (2022) reported the search for KBOs to be > 99% complete for objects at distances 29 < d < 2500 AU (see also Figure 14 of Bernardinelli et al. 2020) that appear in at least 7 nights of data in at least 2 opposition seasons with at least a triplet of detections per season. We also use this set of criteria to determine which members of the synthetic population would have been detected.

For all synthetic objects, we use their orbital parameters to determine whether they intersect the footprint, and for those that do we obtain the RA/DEC coordinates and time of detection. In Figure 1, we show the locations of all members of the synthetic catalog at the specified epoch of June 1st, 2018, in blue, with the detections simulated by DES in yellow. Note that only a small fraction of the survey footprint is intersected by the P9 search region.

\(^1\) https://github.com/bernardinelli/DESTNOSIM
In order to determine whether any given exposure would have detected an object with mass and color of any object in the reference population, we determine apparent magnitudes for each simulated P9. First, for each object in the reference population, we determine an apparent magnitude in DES $griZ$ bands derived from its diameter $D$, assumed albedo $p$, colors relative to $g$-band, solar apparent magnitude $m_\odot$, and the object’s distance from the sun and earth. The assumptions for colors and albedo are given in Table 1. In order to determine the apparent magnitude, we use the absolute magnitude of a body,

$$H = m_\odot - 2.5 \log_{10} \left( \frac{pD^2}{9 \cdot 10^{16} \text{ km}^2} \right)$$

(1)

where the numerical factor is obtained from the definition $^2$ of geometric albedo. The absolute magnitude relates to the apparent magnitude as:

$$m = H + 5 \log_{10}(D_{ob}D_{sb}) - 2.5 \log_{10} \phi(\alpha)$$

(2)

where $D_{ob}$ is the observer-body distance and $D_{sb}$ is the sun-body distance. DES only observes at very small phase angles, so the phase function is $\approx 1$, thus combining the two equations, we obtain:

$$m = m_\odot - 2.5 \log_{10} \left( \frac{pD^2}{9 \cdot 10^{16} \text{ km}^2} \right) + 5 \log_{10}(D_{ob}D_{sb})$$

(3)

In Figure 2, we find that for solar colors and the albedos given in BB21, most members of the synthetic population are around 20-21st magnitude in DES $g$-band, with a long tailing end extending to 25th magnitude.

### 2.2. Recovery

We now determine which synthetic observations would have actually been detected by checking the apparent magnitude of each detection against the limiting magnitude of each exposure. The completeness function for all exposures in DES was determined in Bernardinelli et al. (2022), where the probability of a source being detected is modeled by a logit function:

$$p(m) = \frac{c}{1 + \exp(k(m - m_{50}))}$$

(4)

where $m$ is the detection’s apparent magnitude, $m_{50}$ the magnitude where a source has a detection probability of 50%, $k$ is a transition sharpness for the curve and $c \leq 1$

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$^2$ Starting from equation 5 and the following derivations in Russell (1916), convert to absolute magnitude – a factor of $(2 \text{ AU})^2$ in kilometers gives the $9 \cdot 10^{16} \text{ km}^2$.
Table 1. Table showing percent detected of the reference population that crosses the DES footprint, given different albedos and colors selected based on various scenarios for the composition and surface of Planet Nine.

| Model             | Albedo | $g - r$ | $g - i$ | $g - z$ | % det |
|-------------------|--------|---------|---------|---------|-------|
| Fiducial Model    | BB21   | 0.44    | 0.53    | 0.55    | 87.0% |
| Neptune-like      | 0.5    | -0.3    | -1.35   | -2.15   | 80.3% |
| 40K, 0.1 CH₄      | 0.75   | 0.6     | -0.3    | -0.2    | 88.1% |
| Super-Ganymede    | 0.43   | 0.72    | 0.88    | 0.87    | 85.8% |
| Super-KBO         | 0.1    | 1.0     | 1.25    | 1.5     | 76.8% |

ARCCUT is the shortest ARC that remains when any given detection is removed from the orbit. An ARCCUT of at least 180 implies there exist two detections at least 180 days after the rest.

3. At least one triplet exists whose pairs are within 90 days of each other (as all of the objects here are farther than 50AU).

Note that we do not fully inject these detections into DES – we do not search for these objects using a full pipeline that would recover orbits. The detection efficiency determined in Bernardinelli et al. (2022) accounts not only for incompleteness of the search at the faintest magnitudes, but also at the brightest ones. The suppression of detectability at 19-20 $g$-band mag is clearly visible in Figure 3, as detection of the reference population never reaches 100%. This is due to a higher density of transients and stars in the northern part of the DES footprint, which is accounted for within the completeness function of each exposure, as $c < 1$. Moreover, DESTNOSIM accounts for moving object retrieval by checking detections against the probability of transient detection, thus making it unnecessary to fully inject into the DES pipeline. Such injection would also be prohibitively expensive, as it would require the reprocessing of 80,000 sets of 62 CCD images, along with corresponding coadds. Additionally, the DES search was complete to a magnitude of $r \approx 23.8$ for distances $29 < d < 2500$ AU and 100% efficiency for objects in the catalog above 200AU, which covers the entire range of semi-major axes relevant to the reference population (see also Bernardinelli et al. 2020). For simplicity, we therefore assume that all P9 candidates that meet the three above criteria are recovered. Thus, as the efficiency of detection for distant KBOs has been determined for every exposure in DES by Bernardinelli et al. (2022), it becomes simple to test for the completeness of P9 recovery in the survey.

Figure 4. 2D histogram in albedo vs. $g-r$ space, colored by the proportion of the reference population detected. Redder and higher albedo objects are more easily detected, while the bluer, low albedo range significantly depresses detection.

3. RESULTS

We find that of the 100,000 objects in the P9 reference population, 11,709 cross the DES footprint as seen in Figure 1, and of these, we are able to recover 87.0%, or 10,187 objects. Many of the objects we fail to recover do not cross the footprint on six nights, even before transient efficiency or magnitude limits imposed by the detection probability per exposure are taken into account. This therefore suppresses the maximum efficiency at the lower magnitude range in Figure 3. In Figure 3 we show the completeness of the P9 search in DES in DES $g$-band, finding an $m_{50}$ of around 24.1 apparent magnitude. This value is lower than the DES $g_{50}$ for a simulated TNO population determined in Bernardinelli et al. (2022), likely due to differences in completeness for exposures in the ecliptic plane and unfavorable geometry in the northern latitudes of DES. In total, the extra depth of DES as compared to the ZTF search allows for the detection of 5032 objects not recovered by Brown & Batygin (2021b) allowing us to rule out an additional 5% of the reference population, and thus narrow possible parameter space for P9. It is worth noting that 98% of the objects uniquely detected by DES have $g$-band magnitude greater than 20.75, which is the ZTF $m_{50}$.

While the standard synthetic population assumes solar colors and a range of albedos, we explore how different choices of color and albedo affect the detectability of the population. We repeat our analysis, now test-
ning a spectrum of color and albedo instead of the BB21 values. We allow the $g - r$ color to vary from 0 to 2.0 and calculate the other two colors using the spectral slope defined by $g - r$. We also vary albedo from 0.1 to 0.75, thus creating a 2D space where for each combination of color and albedo we determine the proportion of the reference population detected. In Figure 4 we plot the dependence of detectability on $g - r$ color and albedo. At the lower albedo, bluer range, we detect up to 25% fewer members of the reference population, implying that a Neptune-like P9 would prove significantly harder to detect than a redder, higher albedo P9.

We next explore how differing models for the nature of Planet Nine would affect the detectability. First, we assume a Neptune-like scenario for P9, the results for which are given in the second row of Table 1. Substituting the exact values for albedo and color of Neptune while using the same fiducial model for the mass-radius relation, we recover 7% fewer members of the reference population, dropping to 80% recovery. The extremely blue colors of Neptune suppress recoverability despite the higher albedo, as confirmed by results in Figure 4.

We next consider Neptune-like interior compositions, but more realistic atmospheres for a distant planet. The potential range of optical characteristics for P9 was analyzed in a series of models by Fortney et al. (2016), in which assumptions for the temperature and methane abundance of P9 were used to determine its possible colors (see Table 2 of Fortney et al. 2016). Of these models we choose the one with a 40K temperature (the results in Fortney et al. 2016 do not significantly vary on temperature, and 40K is a middle-ground) and the maximum methane abundance. The resulting $g - r$ color is similar to the solar colors, however the other two colors are much bluer, though not nearly as blue as Neptune. We find that the effect of this model on the recovery of the synthetic population is minimal (third row of Table 1), and in fact increases the amount recovered by 1% as compared to the fiducial model. The albedo from Fortney et al. (2016) is at the maximum of the range of albedos selected for the fiducial model, thus increasing the proportion of the population detected.

For a very different end-member, we examine a mixed ice-rock planet — a “super-Ganymede” — with composition similar to that of Ganymede at around equal parts rock and ice and with an albedo and color equal to those of Ganymede itself. We calculate the radius for each synthetic object by using an equation of state derived by Seager et al. (2007) for a Ganymede-like planet. Effectively, this change narrows the extreme ends of the radius ranges for the reference population, with a 6 $M_\oplus$ planet yielding the same $2 R_\oplus$ radius as previously assumed, while a 10 $M_\oplus$ planet results in a radius of 2.27 $R_\oplus$ rather than 3.3 $R_\oplus$. We assign the “super-Ganymede” colors roughly estimated based on those determined in Millis & Thompson (1975) and Fadden et al. (1980)\(^3\). We find that a Ganymede-like P9 is not significantly different from the fiducial model in terms of ability to detect, as the albedo of 0.43 and very red color offset the smaller size of the planet.

Finally, we consider a low albedo all-ice object that we term a “super-KBO.” We assign this body generically red colors typical of classical KBOS, and give it an albedo of 0.1. While the largest known KBOS have high albedos, we are more interested in exploring a limiting case of one of the most difficult to detect objects that we could envision. We use the same equation of state from Seager et al. (2007) as for the Ganymede-like P9, but with 100% water ice (see Table 4 of Seager et al. 2007). For this equation of state that planet would be slightly larger than the super-Ganymede, such that a 6 $M_\oplus$ planet has a radius of 2.23 $R_\oplus$, while a 10 $M_\oplus$ planet goes up to 2.55 $R_\oplus$. At the higher mass end, these radii are smaller than those of the fiducial model, however are significantly larger at the lower end. Despite the increased radius, the lower albedo of the icy

\(^3\) We transform the $B - V$ color for Ganymede in Millis & Thompson (1975) to SDSS $g - r$, and then use spectral data from Fadden et al. (1980) to extrapolate $g - i$ and $g - z$ colors.
“super-KBO” model causes 10% of the reference population to no longer be detectable. For both the KBO-like and Ganymede-like models, we find that the increase in radius at the lower mass ends offsets the recoverability loss from the decrease in radius at the higher mass end. We find a reduction in the fraction detected of only 10%, from 87% to just under 77%.

4. CONCLUSION

We inject a simulated catalog of 100,000 objects generated in Brown & Batygin (2021b) based on likely orbital parameters of Planet Nine into the Dark Energy Survey, recovering 10,187 of the synthetic objects (87% of all objects that intersect the DES footprint), demonstrating that DES itself is capable of ruling out more than 10% of the predicted parameter space of Planet Nine. Combining these results, DES and ZTF are able to rule out a significant portion of the reference population with perihelion distance under 300 as seen in Figure 5. We find that the median perihelion distance of the reference population after the two searches increases from 300 AU to 340 AU. However, DES itself does not significantly alter the distributions of orbital elements for the reference population as compared to ZTF, as it more evenly removes objects at all distances, eccentricities, and inclinations. The updated reference population, which now indicates which members of the population have been excluded by DES, can be found at https://data.caltech.edu/records/11545 (Brown 2022). We find that modifying assumptions of the albedo and color of P9 can have a noticeable impact on its recoverability, such that a Neptune-like or KBO-like P9 would be somewhat more difficult to recover for surveys which lack the magnitude depth of DES. By contrast, assuming a Ganymede-like P9, or one similar to the models proposed in Fortney et al. (2016), does not significantly alter our ability to recover members of the reference population. The methods used in this paper can be replicated for existing sky surveys, such as Pan-STARRS, Visible and Infrared Survey Telescope for Astronomy (VISTA), or SkyMapper to further constrain the parameter space for P9 and tighten the search. Additionally, the Rubin Observatory Legacy Survey of Space and Time (LSST) can be tested for its ability to recover the P9 reference population using simulated exposures in order to test how quickly the survey would be able to detect and rule out parts of the Planet Nine parameter space.

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