Systematic KMTNet Planetary Anomaly Search. VIII. Complete Sample of 2019 Subprime Field Planets

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Abstract

We complete the publication of all microlensing planets (and “possible planets”) identified by the uniform approach of the KMT AnomalyFinder system in the 21 KMT subprime fields during the 2019 observing season, namely, KMT-2019-BLG-0298, KMT-2019-BLG-1216, KMT-2019-BLG-2783, OGLE-2019-BLG-0249, and OGLE-2019-BLG-0679 (planets), as well as OGLE-2019-BLG-0344 and KMT-2019-BLG-0304 (possible planets). The five planets have mean log mass ratio measurements of \((-2.6, -3.6, -2.5, -2.2, -2.3\)) M_Jup, and median distance estimates of \((6.7, 5.9, 6.4, 5.6)\) kpc, respectively. The main scientific interest of these planets is that they complete the AnomalyFinder sample for 2019, which has a total of 25 planets that are likely to enter the statistical sample. We find statistical consistency with the previously published 33 planets from the 2018 AnomalyFinder analysis according to an ensemble of five tests. Of the 58 planets from 2018–2019, 23 were newly discovered by AnomalyFinder. Within statistical precision, half of the planets have caustic crossings, while half do not; an equal number of detected planets result from major- and minor-image light-curve perturbations; and an equal number come from KMT prime fields versus subprime fields.

Unified Astronomy Thesaurus concepts: Exoplanet detection methods (489); Gravitational microlensing (672); Gravitational microlensing exoplanet detection (2147)

1. Introduction

We present the analysis of all planetary events that were identified by the KMTNet AnomalyFinder algorithm (Zang et al. 2021, 2022) and occurred during the 2019 season within the 21 subprime KMTNet fields covering \(\sim 84\) deg\(^2\) that lie in the periphery of the richest microlensing region of the Galactic bulge and are observed with cadences \(\Gamma = 0.2\)–1 hr\(^{-1}\). This
work follows the publication of complete samples of the 2018 prime (Gould et al. 2022; Hwang et al. 2022; Wang et al. 2022) and subprime (Jung et al. 2022) AnomalyFinder events, the 2019 prime (Zang et al. 2021; Hwang et al. 2022; Zang et al. 2022) events, and a complete sample of all events from 2016 to 2019 with planet–host mass ratios $q < 10^{-4}$ (Zang et al. 2023). The above references are (ignoring duplicates) Papers I, IV, II, III, V, VI, and VII in the AnomalyFinder series. The locations and cadences of the KMTNet fields are shown in Figure 12 of Kim et al. (2018a). Our immediate goal, which we expect to achieve within a year, is to publish all AnomalyFinder planets from 2016 to 2019. Over the longer term, we plan to apply AnomalyFinder to all subsequent KMT seasons, beginning with 2021.

For the 2019 subprime fields, AnomalyFinder identified a total of 182 anomalous events (from an underlying sample of 1895 events), which were classified as “planet” (9), “planet/binary” (10), “binary/planet” (18), “binary” (136), and “finite source” (9). Among the 136 in the “binary” classification, 56 were judged by eye to be unambiguously nonplanetary in nature. Among the nine in the “planet” classification, four were previously published (including two AnomalyFinder discoveries), while one had been recognized but remained unpublished. Among the 10 in the “planet/binary” classification, three were either published planets (one) or had been recognized by eye (two), and among the 18 in the “binary/planet” classification, one was a previously published planet. Among the 136 classified as “binary,” one (a two-planet system) was published. Thus, in total, AnomalyFinder recovered nine planets that had been previously found by eye, including six that were published and three others that had not been published. The latter are KMT-2019-BLG-1216, OGLE-2019-BLG-0249, and OGLE-2019-BLG-0679.

Our overall goal is to present full analyses of all events with mass ratios $q < 0.03$. To this end, we carry out systematic investigations of all of the AnomalyFinder candidates (other than the 56 classified by eye as unambiguously nonplanetary) using end-of-season pipeline data. Any (unpublished) event that is found to have a viable solution with $q < 0.06$ is then reanalyzed based on tender loving care (TLC) re reductions. If there are viable planetary solutions ($q < 0.03$), we report a detailed analysis regardless of whether the planetary interpretation is decisively favored. If the TLC analysis leads to viable solutions with $0.03 < q < 0.05$, we report briefly on the analysis but do not present all details. In the 2019 subprime sample, there was one such event, KMT-2019-BLG-0967, with $q = 0.040 \pm 0.004$. This event also has a competing solution in which the anomaly is generated by a binary source rather than a low-mass companion, so it could not be included in the final sample even if the sample boundary were moved upward. Finally, we note that one event, OGLE-2019-BLG-1352, that was selected as a “finite source” and hence required detailed investigation, may well have strong evidence of a planet based on extensive follow-up data. However, we find that, even if so, the signal from survey-only data is not strong enough to claim a planet. Hence, we do not include it in the present paper.

### 2. Observations

The description of the observations is nearly identical to that in Gould et al. (2022) and Jung et al. (2022). The KMTNet data are taken from three identical 1.6 m telescopes, each equipped with cameras of 4 deg$^2$ (Kim et al. 2016) and located in Australia (KMTA), Chile (KMTc), and South Africa (KMTS). When available, our general policy is to include Optical Gravitational Lensing Experiment (OGLE) and Microlensing Observations in Astrophysics (MOA) data in the analysis. However, none of the seven events analyzed here were alerted by MOA. OGLE data were taken using their 1.3 m telescope with a 1.4 deg$^2$ field of view at Las Campanas Observatory in Chile (Ulalski et al. 2015). For the light-curve analysis, we use only the $I$-band data.

As in those papers, Table 1 gives basic observational information about each event. The first column gives the event names in order of discovery (if discovered by multiple teams), which enables cross identification. The nominal cadences are given in the second column, and the third column shows the first discovery date. The remaining four columns show the event coordinates in the equatorial and galactic systems. Events with OGLE names were originally discovered by the OGLE Early Warning System (Ulalski et al. 1994; Ulalski 2003). Events with KMT names and discovery dates were first found by the KMT AlertFinder system (Kim et al. 2018b), while those listed as “postseason” were found by the KMT EventFinder system (Kim et al. 2018a).

| Event Name | $\gamma$ (hr$^{-1}$) | Alert Date | R.A.1200 | Decl.1200 | $l$ | $b$ |
|------------|---------------------|------------|----------|----------|-----|-----|
| KMT-2019-BLG-0298 | 1.0 | 2019 Apr 5 | 17:39:30.72 | −27:38:17.30 | +0.40 | +1.83 |
| OGLE-2019-BLG-0445 | 0.4 | | | | |
| KMT-2019-BLG-1216 | 0.4 | 2019 Jun 11 | 17:53:55.35 | −35:08:11.90 | −4.43 | −4.69 |
| OGLE-2019-BLG-1033 | 0.2 | | | | |
| KMT-2019-BLG-2783 | 1.0 | Postseason | 17:57:10.06 | −33:47:18.67 | −2.92 | −4.59 |
| OGLE-2019-BLG-0249 | 0.1 | 2019 Mar 9 | 17:41:36.84 | −34:42:06.30 | −5.35 | −2.30 |
| KMT-2019-BLG-0109 | 0.4 | | | | |
| OGLE-2019-BLG-0679 | 0.1 | 2019 May 5 | 17:42:57.70 | −27:46:22.37 | +0.69 | +1.11 |
| KMT-2019-BLG-2688 | 1.0 | | | | |
| OGLE-2019-BLG-0344 | 0.2 | 2019 Mar 20 | 17:23:52.38 | −29:32:48.59 | −3.08 | +3.66 |
| KMT-2019-BLG-0149 | 0.4 | | | | |
| KMT-2019-BLG-0304 | 1.0 | 2019 Apr 5 | 17:41:18.70 | −32:31:47.82 | −3.54 | −1.10 |

### Table 1

Event Names, Cadences, Alerts, and Locations
Two events, OGLE-2019-BLG-0249 and OGLE-2019-BLG-0679, were observed by Spitzer as part of a large-scale microlensing program (Yee et al. 2015), but these data will be analyzed elsewhere. As is generally the case, these Spitzer observations were supported by ground-based observations. In the case of OGLE-2019-BLG-0679, these observations consisted of 21 epochs, spread over 72 days, of $I/H$ observations on the ANDICAM camera at the SMARTS 1.3 m telescope in Chile, whose aim was to determine the source color. We will make use of these observations only for that purpose, i.e., not for the modeling. On the other hand, OGLE-2019-BLG-0249 was the object of intensive ground-based observations. These require special handling, as described in Section 3.5.1. To the best of our knowledge, there were no other ground-based follow-up observations of any of these events.

The KMT and OGLE data were reduced using difference image analysis (Tomany & Crots 1996; Alard & Lupton 1998) as implemented by each group, i.e., Albrow et al. (2009) and Woźniak (2000), respectively.

3. Light-curve Analysis

3.1. Preamble

With one exception that is explicitly noted below, we reproduce here Section 3.1 of Jung et al. (2022), which describes the common features of the light-curve analysis. We do so (rather than simply referencing that paper) to provide easy access to the formulae and variable names used throughout this paper. The reader who is interested in more details should consult Section 3.1 of Gould et al. (2022). Readers who are already familiar with these previous works can skip this section after first reviewing the paragraph containing Equation (9) below.

All of the events can be initially approximated by 1L1S models, which are specified by three Paczyński (1986) parameters, $(l_0, u_0, t_E)$, i.e., the time of lens-source closest approach, the impact parameter in units of $t_E$, and the Einstein timescale,

$$t_E = \frac{\theta_E}{\mu_{\text{rel}}}, \quad \theta_E = \sqrt{\kappa M \pi_{\text{rel}}},$$

where $M$ is the lens mass; $\pi_{\text{rel}}$ and $\mu_{\text{rel}}$ are the lens-source relative parallax and proper motion, respectively; and $\mu_{\text{rel}} = [\mu_{\text{rel}}]$. The notation “$n$L$mS$” means $n$ lenses and $m$ sources. In addition to these three nonlinear parameters, there are two flux parameters, $(f_S, f_B)$, that are required for each observatory, representing the source flux and the blended flux.

We then search for “static” 2L1S solutions, which generally require four additional parameters $(s, q, \alpha, \rho)$, i.e., the planet–host separation in units of $t_E$, the planet–host mass ratio, the angle of the source trajectory relative to the binary axis, and the angular source size normalized to $\theta_E$, i.e., $\rho = \theta_{\text{rad}}/\theta_E$.

We first conduct a grid search with $(s, q)$ held fixed at a grid of values and the remaining five parameters allowed to vary in a Markov Chain Monte Carlo (MCMC). After we identify one or more local minima, we refine these by allowing all seven parameters to vary.

We often make use of the heuristic analysis introduced by Hwang et al. (2022) and modified by Ryu et al. (2022) based on further investigation in Gould et al. (2022). If a brief anomaly at $t_{\text{anom}}$ is treated as due to the source crossing the planet–host axis, then one can estimate two relevant parameters,

$$s_{\perp} = \frac{\sqrt{4 + u_{\text{anom}}^2} \pm u_{\text{anom}}}{2}, \quad \tan \alpha = \frac{u_0}{\tau_{\text{anom}}},$$

where $u_{\text{anom}} = \tau_{\text{anom}} + \mu_0$ and $\tau_{\text{anom}} = (t_{\text{anom}} - t_0)/t_E$. Usually, $s_{\perp} > 1$ corresponds to anomalous bumps, and $s_{\perp} < 1$ corresponds to anomalous dips. This formalism predicts that if there are two degenerate solutions, $s_{\perp}$, they both have the same $\alpha$ and that there exists a $\Delta \ln s$ such that

$$s_{\perp} = s^\dagger \exp(\pm \Delta \ln s),$$

where $\alpha$ and $s^\dagger$ are given by Equation (2). To test this prediction in individual cases, we can compare the purely empirical quantity $s^\dagger \equiv \sqrt{\frac{\kappa}{\theta_E}}$ with the prediction from Equation (2), which we always label with a subscript, i.e., either $s^\dagger_0$ or $s^\dagger_1$. This formalism can also be used to find “missing solutions” that have been missed in the grid search, as was done, e.g., for the case of KMT-2021-BLG-1391 (Ryu et al. 2022).

For cases in which the anomaly is a dip, the mass ratio $q$ can be estimated,

$$q = \left(\frac{\Delta t_{\text{dip}}}{4 t_E}\right)^2 \frac{s^\dagger}{|u_0|} |\sin^3 \alpha|,$$

where $\Delta t_{\text{dip}}$ is the full duration of the dip. In some cases, we investigate whether the microlens parallax vector,

$$\pi_E \equiv \frac{\pi_{\text{rel}}}{\theta_E}, \quad \mu_{\text{rel}} \equiv \frac{\mu_{\text{rel}}}{\theta_E},$$

can be constrained by the data. When both $\pi_E$ and $\theta_E$ are measured, they can be combined to yield

$$M = \frac{\theta_E}{\kappa \pi_E}, \quad D_L = \frac{\text{au}}{\theta_E \pi_E + \pi_S},$$

where $D_L$ is the distance to the lens, and $\pi_S$ is the parallax of the source.

To model the parallax effects due to Earth’s orbital motion, we add two parameters $(\pi_{E,N}, \pi_{E,E})$ that are the components of $\pi_E$ in equatorial coordinates. We also add (at least initially) two parameters $\gamma = [(ds/dt)/s, \, d\alpha/dt]$, where $s$ and $\alpha$ are the first derivatives of projected lens orbital position at $t_0$, i.e., parallel and perpendicular to the projected separation of the planet at that time, respectively. In order to eliminate unphysical solutions, we impose a constraint on the ratio of the transverse kinetic to potential energy,

$$\beta = \frac{\text{KE}}{\text{PE}} = \frac{\kappa M m_2 \gamma^2 \pi_{E,N}^2}{8 \pi^2 \pi_E^2 \tau_{\text{anom}}^2 \left(\pi_E + \pi_S/\theta_E\right)^3} < 0.8.$$ (7)

It often happens that $\gamma$ is neither significantly constrained nor significantly correlated with $\pi_E$. In these cases, we suppress these 2 degrees of freedom (dof).

Particularly if there are no sharp caustic-crossing features in the light curve, 2L1S events can be mimicked by 1L2S events. Where relevant, we test for such solutions by adding at least three parameters $(t_{0,2}, u_{0,2}, f_{p,2})$ to the 1L2S models. These are the time of closest approach, the impact parameter of the
second source, and the ratio of the second to the first source flux in the $I$ band. If either lens-source approach can be interpreted as exhibiting finite-source effects, then we must add one or two further parameters, i.e., $\rho_1$ and/or $\rho_2$. And, if the two sources are projected closely enough on the sky, one must also consider source orbital motion.

In a few cases, we make kinematic arguments that solutions are unlikely because their inferred proper motions $\mu_{\text{rel}}$ are too small. If planetary events (or, more generally, anomalous events with planet-like signatures) traced the overall population of microlensing events, then the fraction with proper motions less than a given $\mu_{\text{rel}} < \sigma_\mu$ would be

$$p(\mu_{\text{rel}} \leq \sigma_\mu) = \frac{(\mu_{\text{rel}}/\sigma_\mu)^3}{6\sqrt{\pi}}$$

$$\rightarrow 4 \times 10^{-3} \left( \frac{\mu_{\text{rel}}}{1 \text{ mas yr}^{-1}} \right)^3 \quad \text{(old)},$$

where (following Gould et al. 2021) the bulge proper motions are approximated as an isotropic Gaussian with dispersion $\sigma_\mu = 2.9 \text{ mas yr}^{-1}$.

However, subsequent to the work of Gould et al. (2022) and Jung et al. (2022), Gould (2022) showed that the proper-motion distribution of observed planetary microlensing events scales $\propto d\mu \mu^2 \exp(-\mu/2\sigma_\mu^2)$, where $\sigma_\mu = 3.06 \pm 0.29 \text{ mas yr}^{-1}$, and $\mu = 1.02 \pm 0.29$. Hence, in place of Equation (8), we adopt

$$p(\mu_{\text{rel}} \leq \sigma_\mu) = \frac{(\mu_{\text{rel}}/\sigma_\mu)^{\nu+1}}{[(\nu + 1)/2]!} \times \frac{\mu_{\text{rel}}^2}{4\sigma_\mu^2}$$

$$\rightarrow 2.8 \times 10^{-2} \left( \frac{\mu_{\text{rel}}}{1 \text{ mas yr}^{-1}} \right)^2,$$

where we have evaluated at $\sigma_\mu = 3.0 \text{ mas yr}^{-1}$ and $\nu = 1$. For example, $p(0.5 \text{ mas yr}^{-1}) = 0.7\%$ and $p(0.1 \text{ mas yr}^{-1}) = 0.03\%$.

### 3.3. KMT-2019-BLG-1216

Figure 3 shows a low-amplitude ($\Delta I \approx 0.5$) microlensing event peaking at $t_0 = 8658.4$ and punctuated by a smooth bump at $t_{\text{anom}} \approx 8582$, i.e., $-39.4$ days before peak. Assuming that the source is unblended (as is reasonable for such a bright source), the remaining Paczyński (1986) parameters are $u_0 = 0.60$ and $t_{\text{E}} = 28$ days. Then $t_{\text{anom}} = -1.41$ and $u_{\text{anom}} = 1.53$. Because the source is bright (so, large), while the caustics are likely to be small (because $u_{\text{anom}} > 1$), we consider that the bump could be due to either a major- or a minor-image perturbation. For these, Equation (2) predicts $s^\alpha = 2.02$ and $\alpha = 158^\circ$ and $s^\beta = 0.49$ and $\alpha = 338^\circ$, respectively.

The grid search returns two solutions, whose refinements are shown in Table 2. The wide solution is substantially preferred by $\Delta \chi^2 = 19$. For this solution, the heuristic prediction of $\alpha$ is confirmed, while the fit value of $s = s_{\text{outer}} = 1.89$ indicates that there could be another solution near $s_{\text{inner}} = (s^\alpha)^2/s_{\text{outer}} = 2.16$. As a matter of due diligence, we seed an MCMC with this value and indeed find a local minimum at $s_{\text{inner}} = 2.22$. However, this solution is ruled out at $\Delta \chi^2 = 205$, which confirms its failure to be detected in the grid search. The reason that the degeneracy is decisively broken in this case is that the inner/outer degeneracy is most severe for angles $\alpha = \pm 90^\circ$ (Zhang et al. 2022), whereas in this case, $\alpha = (90 + 68)^\circ$.

Although there is no signature of finite-source effects in the light curve (i.e., all values $\rho < 0.046$ are consistent at $1\sigma$), the absence of a signal actually places significant constraints: $\rho < 0.061$ at $2\sigma$ and $\rho < 0.077$ at $3\sigma$. That is, sufficiently larger sources would be impacted by the caustic. See the inset in Figure 1. Hence, when we carry out a Bayesian analysis in Section 5.1, we will ultimately incorporate a $\rho$-envelope function to represent this constraint. For the present, however, we simply note that, in light of the source-radius estimate $\theta_{\text{src}} = 0.13 \text{ mas}$ and lens-source relative proper motions $\mu_{\text{rel}} > 1.7 \text{ mas yr}^{-1}$. These values imply that while there is no compelling reason to believe that $\theta_{\text{src}}$ is large, it could be relatively large, e.g., $\theta_{\text{src}} \sim 0.75 \text{ mas}$, and still be consistent with the typical range of microlensing proper motions, $\mu_{\text{rel}} \lesssim 10 \text{ mas yr}^{-1}$. If so, this would favor a nearby lens and thus a potentially large (hence, measurable) microlens parallax, $\pi_{\text{E}}$.

Therefore, as a matter of due diligence, and despite the relatively short Einstein timescale, $t_{\text{E}} = 28$ days, we undertake a parallax analysis. The results, given in Table 2, are consistent with $\pi_{\text{E}} = 0$, and in this sense, the parallax is “undetected.” Nevertheless, as shown in Figure 2, the fit does place one-dimensional (1D) constraints on $\pi_{\text{E}}$, and we will incorporate these into the Bayesian analysis in Section 5.1. However, because $\pi_{\text{E}}$ is poorly constrained in the orthogonal direction, the free fit allows for $\pi_{\text{E}}$ values that would be very unlikely in a posterior fit. These would unphysically broaden the errors in $q$ and other parameters. We note that (except for $u_0$) all of the parameter values from the parallax fits are consistent with those from the standard fits at $1\sigma$. Hence, we will finally report $(s, q)$ and the physical parameters that are derived from them based on the standard fit of Table 2.

Because the anomaly is a smooth, featureless bump, we must also consider the possibility that it is due to an extra source, i.e., 1L2S, rather than an extra lens. However, we find that such models are excluded by $\Delta \chi^2 = 125$.  

### 3.4. KMT-2019-BLG-3029

Figure 3 shows a low-amplitude, generally smooth microlensing event peaking at $t_0 = 8658.44$ except for six elevated points (from three observatories) over an interval of 5.2 hr. The maximum extent of the deviation, defined by the two limiting points that lie on the 1L1S curve, is 15.4 hr and centered at $t_{\text{anom}} \approx 8658.62$, i.e., just $+0.18$ days after peak. A 1L1S fit to the unperturbed parts of the light curve yields $u_0 = 0.18$ and $t_{\text{E}} = 90$ days. Then, $t_{\text{anom}} = +0.002$ and $u_{\text{anom}} = 0.18$. The first anomalous point is about 0.4 mag brighter than the others and almost 1 mag brighter than the point that precedes it by 2.3 hr. Therefore, the anomaly is very likely to be a caustic entrance that is followed by a caustic trough, but it is difficult to make a further assessment by eye. Within the heuristic framework of an on-axis (or near-axis) anomaly, Equation (2) predicts $s^\alpha = 1.094$ and $\alpha = 89.4^\circ$.

The grid search returns three solutions, whose refinements are shown in Table 3 and illustrated in Figure 3. Two of these are a classical inner/outer degeneracy (Gaudi & Gould 1997), in which (as is often the case; Yee et al. 2021) the outer-solution caustic has a resonant topology, with the source
intersecting its “planetary wing.” These two solutions have nearly identical values of $\alpha = 89.6^\circ$, which are in excellent agreement with the heuristic prediction, and $s = \sqrt{S_{\text{inner}} - S_{\text{outer}}} = 1.094$, which is also in near-perfect agreement.

However, there is also a third solution, which has a fully resonant topology and in which the source intersects an off-axis cusp. As illustrated by Figure 3, the degeneracy of the inner/outer solutions is intrinsic, while the off-axis solution degeneracy is only possible because of a lack of data during the latter part of the caustic perturbation. Nevertheless, because the off-axis solution is disfavored by $\Delta \chi^2 = 24$, we reject it.

As a matter of due diligence, we check for 1L2S solutions but find that these are rejected by $\Delta \chi^2 = 57$.

While $\rho$ is not well measured, there is a weak $\chi^2$ minimum at $\rho \sim 6 \times 10^{-4}$ and a secure 3$\sigma$ upper limit of $\rho < 11 \times 10^{-4}$. In Section 4.2, we will show that $\theta_{\text{ast}} \simeq 0.40$ mas. Hence, these $\rho$ values correspond to $\theta_{\text{E}} \approx 0.67$ and $>0.36$ mas. Thus, it is at least plausible that $\theta_{\text{E}}$ is relatively large, which would be consistent with a nearby lens, so a relatively large (hence,
expression is $s_{\text{anom,bump}} = 0.34, \nu_{\text{anom,bump}} = 0.10, s_{\text{t}} = 0.06$ and $t_{\text{anom,bump}} = 8758.0$. If the dip is regarded as the driving feature, then $\nu_{\text{anom,dip}} = -0.34, \nu_{\text{anom}} = 0.34, s_{\text{anom}} = 0.84$, and $\alpha_{\text{anom}} = 350^\circ$, while if the bump is regarded as the driving feature, then $\nu_{\text{anom,bump}} = -0.27, \nu_{\text{anom}} = 0.27, s_{\text{anom}} = 1.14$, and $\alpha_{\text{anom,bump}} = 168^\circ$. In either case, if the heuristic prediction is correct, the nondriving feature (bump or dip, respectively) would have to be naturally explained by the resulting geometry.

In fact, the grid search returns only a single solution whose refinement is shown in Table 5. The $s_{\text{anom}}$ prediction is qualitatively confirmed, while the $\alpha$ prediction is off by $\sim 10^\circ$. The $s_{\text{anom}}$ inaccuracy derives from the difficulty in judging the exact position of the dip in the presence of the bump. As can be seen from Figure 5, the error in the $\alpha$ estimate is due to the generic problem that minor-image caustics lie off-axis, which is exacerbated by the fact that $q$ is large, implying that the separation of the caustics is also large (Han 2006). As anticipated in the previous paragraph, the bump is then naturally explained by the fact that the source passes close to a cusp as it exits the trough between the two minor-image caustics. Indeed, there is also a bump before the dip, which is much weaker because the source passes much farther from the cusp. This bump is hardly noticeable in the data because of larger error bars, but it can be discerned in the model.

The source passes about 0.015 from the cusp, which creates a $3\sigma$ limit, $\rho < 0.01$. This is of relatively little interest because, given the estimate $\theta_{\text{rel}} = 0.39\text{mas}$ that is derived in Section 4.3, it corresponds to a limit $\theta_{\text{rel}} > 0.6\text{mas yr}^{-1}$, which excludes only a small fraction of parameter space. Nevertheless, we will include the constraints on $\rho$ via an envelope function when we carry out the Bayesian analysis in Section 5.3.

Due to the brevity of the event, the faintness of the source, and the absence of any data more than 10 days after $t_0$, we do not attempt a parallax analysis.

### 3.4. KMT-2019-BLG-2783

Figure 5 shows a smooth, somewhat complex perturbation on the rising wing of a microlensing event that peaks at $t_0 = 8764.4$, i.e., close to the end of the season. When this anomaly is excised, a 1L1S fit yields $u_0 = 0.06$ and $t_0 = 24$ days. The anomaly is characterized by a dip at $t_{\text{anom,dip}} \approx 8756.3$, followed by a bump at $t_{\text{anom,bump}} \approx 8758.0$. If the dip is regarded as the driving feature, then $\nu_{\text{anom,dip}} = -0.34, \nu_{\text{anom}} = 0.34, s_{\text{anom}} = 0.84$, and $\alpha_{\text{anom}} = 350^\circ$, while if the bump is regarded as the driving feature, then $\nu_{\text{anom,bump}} = -0.27, \nu_{\text{anom}} = 0.27, s_{\text{anom}} = 1.14$, and $\alpha_{\text{anom,bump}} = 168^\circ$. In either case, if the heuristic prediction is correct, the nondriving feature (bump or dip, respectively) would have to be naturally explained by the resulting geometry.

In fact, the grid search returns only a single solution whose refinement is shown in Table 5. The $s_{\text{anom}}$ prediction is
so-called “invariants,” i.e., combinations of semidegenerate variables that vary very little when $t_E$ is changed substantially, $t_{\text{eff}} \equiv u_0 t_E$ (for high-magnification events; Yee et al. 2012), as $\tan \alpha = t_{\text{eff}} / \Delta t_{\text{anom}}$. However, it does affect $s^1$, with the 24% blending fraction driving $s^1$ about 24% closer to unity.20

3.5.1. A Survey + Follow-up Event

In addition to survey data from OGLE and KMT, OGLE-2019-BLG-0249 was intensively observed by many follow-up observatories (see Figure 6), in part because it was a Spitzer target and in part because it was a moderately high magnification event ($A_{\text{max}} = 33$) in a low-cadence field (see Table 1). The current AnomalyFinder series of papers includes

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20 That is, in the limit $u_{\text{anom}} \ll 1$, $s^1 \to 1 \pm u_{\text{anom}}/2$, while $u_{\text{anom}}^2 = u_0^2 + \Delta t_{\text{anom}}^2 \to (u_0^2 + (\Delta t_{\text{anom}}^2))/t_E^2$. Hence, $s^1 \to 1 \pm \eta/t_E$, where $\eta \equiv 0.5 \sqrt{u_0^2 + (\Delta t_{\text{anom}}^2)}$, in which the first term is an invariant, and the second is a direct observable.

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the analysis of events only if they are (1) identified as anomalous by the AnomalyFinder algorithm, which is applied to KMT data alone, and (2) have a plausible planetary solution based on survey data alone. This “publication grade” analysis then (3) lays the basis for deciding whether the planet is ultimately included in the complete AnomalyFinder sample. From the standpoint of items (2) and (3), it is therefore essential to ask how the event would have been evaluated in the absence of follow-up data.

Nevertheless, if this evaluation determines that the planet should be in the AnomalyFinder papers and/or if it is included in the final sample, the follow-up data may be used to improve the characterization of the planet.

Planet OGLE-2019-BLG-0249 is the first planet with extensive follow-up data to be included in this series, following seven previous papers containing a total of 37 planets and “possible planets.” There have, of course, been other published planets that had extensive follow-up data and will ultimately...
enter the AnomalyFinder sample. For example, the planetary anomaly in OGLE-2019-BLG-0960 was originally discovered in follow-up data; therefore, Yee et al. (2021) carefully assessed that this planet could be adequately characterized based on survey data alone. Another relevant example is OGLE-2016-BLG-1195, for which the MOA group obtained intensive data over peak (including the anomaly) by using their survey telescope in follow-up mode (Bond et al. 2017). In principle, one should assess whether this anomaly would have been adequately characterized had MOA observed at its normal cadence. However, as a practical matter, this is unnecessary because the KMT and Spitzer groups showed that this planet could be adequately characterized based on an independent survey-only data set (Shvartzvald et al. 2017).

3.5.2. Survey-only Analysis

Thus, we began by analyzing the survey data alone. These results have already been reported above in Table 6, when we compared them to the heuristic predictions. We note that before making these fits, we removed the KMTC points during the 2 days [HJD' − 8706] < 1 due to saturation and/or significant nonlinearity of this very bright target. We also checked that if these excluded points were reintroduced, which we do not advocate, the parameters were affected by \( \lesssim 2\sigma \).

![Figure 3. Data (color-coded by observatory) together with the predictions and residuals for the three planetary models of KMT-2019-BLG-1216 specified in Table 3, plus the 1L2S model. The caustic topologies are shown in the insets for the inner, outer, and off-axis models. The first two are perfectly degenerate, but the last is decisively disfavored at \( \Delta \chi^2 = 24 \). The 1L2S model is excluded at \( \Delta \chi^2 = 57 \).](image)

| Parameters | Inner | Outer | Off-axis |
|------------|-------|-------|---------|
| \( \chi^2_{\text{red}} / \text{dof} \) | 1236.1/1374 | 1236.3/1374 | 1260.6/1374 |
| \( t_0 \) (HJD') | 8658.443 ± 0.239 | 8658.448 ± 0.235 | 8660.978 ± 0.292 |
| \( u_0 \) | 0.189 ± 0.036 | 0.175 ± 0.036 | 0.125 ± 0.007 |
| \( t_0 \) (days) | 88.729 ± 15.031 | 94.466 ± 16.881 | 138.780 ± 8.103 |
| \( s \) | 1.118 ± 0.018 | 1.074 ± 0.022 | 0.990 ± 0.003 |
| \( q \left( 10^{-4} \right) \) | 2.438 ± 0.538 | 2.323 ± 0.511 | 84.727 ± 10.535 |
| \( \langle \log q \rangle \) | −3.617 ± 0.099 | −3.637 ± 0.098 | −2.066 ± 0.085 |
| \( \alpha \) (rad) | 1.561 ± 0.014 | 1.561 ± 0.014 | 0.013 ± 0.032 |
| \( \rho \left( 10^{-4} \right) \) | 4.409 ± 1.897 | 4.184 ± 1.750 | 1.757 ± 0.662 |
| \( f_{\text{S,OGLE}} \) | 0.044 ± 0.010 | 0.040 ± 0.010 | 0.025 ± 0.002 |
| \( f_{\text{B,OGLE}} \) | 0.087 ± 0.010 | 0.090 ± 0.009 | 0.105 ± 0.002 |

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In addition to the two planetary solutions shown in Table 6, there are two local minima derived from the grid search that, when refined, have binary-star mass ratios, i.e., $q \sim 0.15$ and 0.25, with source trajectories passing roughly parallel to a side of a Chang & Refsdal (1979, 1984) caustic (not shown). This is a common form of planet-binary degeneracy for dip-type anomalies (Han & Gaudi 2008). However, in the present case, these binary solutions are rejected by $\Delta \chi^2 = 64$; see Table 6.

As a matter of due diligence, we also fit the data to 1L2S models. These usually give poor fits to dip-type anomalies, but there can be exceptions. However, in this case, we find that 1L2S is ruled out by $\Delta \chi^2 = 475$. 

| Parameters          | Inner $u_0 > 0$ | Inner $u_0 < 0$ | Outer $u_0 > 0$ | Outer $u_0 < 0$ |
|---------------------|----------------|----------------|----------------|----------------|
| $\chi^2_{\text{red}}$/dof | 1226.3/1370  | 1226.4/1370  | 1226.3/1370  | 1226.5/1370  |
| $t_0$ (HJD$^\dagger$) | 8659.089 ± 0.345 | 8659.053 ± 0.340 | 8659.115 ± 0.338 | 8659.048 ± 0.325 |
| $u_0$                | 0.110 ± 0.030  | -0.122 ± 0.035 | 0.119 ± 0.035 | -0.137 ± 0.037 |
| $r_e$ (days)         | 137.465 ± 33.221 | 134.147 ± 36.134 | 129.232 ± 36.368 | 120.992 ± 30.846 |
| $q$ ($10^{-4}$)      | 2.264 ± 0.593  | 2.197 ± 0.594  | 2.029 ± 0.616  | 2.319 ± 0.585  |
| $\langle \log q \rangle$ | -3.644 ± 0.116  | -3.661 ± 0.121  | -3.658 ± 0.120  | -3.639 ± 0.114  |
| $\sigma$ (rad)       | 1.60 ± 0.022   | -1.600 ± 0.020 | 1.603 ± 0.022  | -1.599 ± 0.020 |
| $\rho$ ($10^{-4}$)   | 2.537 ± 1.266  | 2.566 ± 1.266  | 2.772 ± 1.350  | 3.089 ± 1.514  |
| $\pi_{E,N}$          | 0.122 ± 0.101  | 0.381 ± 0.299  | 0.132 ± 0.103  | 0.453 ± 0.316  |
| $\pi_{E,E}$          | 0.543 ± 0.194  | 0.523 ± 0.238  | 0.535 ± 0.199  | 0.490 ± 0.255  |
| $d\pi/dt$ (yr$^{-1}$) | -0.010 ± 0.880 | -0.010 ± 0.920 | -0.050 ± 0.870 | 0.010 ± 0.890  |
| $d\sigma/dt$ (yr$^{-1}$) | 0.090 ± 0.860 | -0.190 ± 0.890 | 0.050 ± 0.870  | -0.290 ± 0.910 |
| $f_{\text{S,OGLE}}$  | 0.023 ± 0.007  | 0.025 ± 0.008  | 0.002 ± 0.008  | 0.002 ± 0.009  |
| $f_{\text{B,OGLE}}$  | 0.108 ± 0.007  | 0.106 ± 0.008  | 0.106 ± 0.008  | 0.102 ± 0.009  |

Figure 4. Scatter plot on the $\pi_{E}$ plane derived from the MCMC of the four parallax models of KMT-2019-BLG-1216 presented in Table 4, color-coded (red, yellow, green, cyan, blue) for $\Delta \chi^2 < (1, 4, 9, 16, 25)$. The black contours show the mean and covariances ($\Delta \chi^2 = 1$) that are used in Section 5.2.
3.5.3. Follow-up Data

The follow-up observations were all, directly or indirectly, initiated in response to an alert that this event would be monitored by Spitzer. Although the Spitzer observations themselves could not begin until July 9 (due to telescope-pointing restrictions), i.e., 66 days after \( t_0 \), OGLE-2019-BLG-0249 was chosen by the Spitzer team on April 29 (6 days before \( t_0 \) in order to “claim” any planets that were discovered (which would also ultimately require that the microlens parallax be measured at sufficient precision). See the protocols of Yee et al. (2015).

On April 30, the Tsinghua Microlensing Group, working with the Spitzer team, initiated observations on three 1 m telescopes from the Las Cumbres Observatory at the same locations as the KMT telescopes, which we designate in parallel as LCOC, LCOS, and LCOA, using an SDSS \( i \) filter.

Based on these observations, combined with ongoing survey observations by OGLE and KMT, these teams noted that the event was probably anomalous and, on this basis, alerted the microlensing community by email. Because this alert was triggered by an anomaly, such observations can be used only to characterize the planet, not to “claim” its detection according to the Spitzer protocols (Yee et al. 2015).

Four observatories in the Microlensing Follow Up Network (\( \mu \)FUN), which is composed mainly of small telescopes, responded to this alert, i.e., the Auckland, Farm Cove, Kumeu, and Turitea observatories, respectively, in

| Parameters     | Values                       |
|----------------|------------------------------|
| \( \chi^2 \)/dof | 2113.9/2103                  |
| \( e_0 \) (HJD') | 8765.395 \pm 0.037           |
| \( u_0 \)       | 0.057 \pm 0.003              |
| \( i_0 \) (days) | 23.669 \pm 1.065             |
| \( s \)         | 0.814 \pm 0.007              |
| \( q \) (10^{-3}) | 3.262 \pm 0.762             |
| \( \log q \)   | -2.483 \pm 0.101            |
| \( \alpha \) (rad) | 6.277 \pm 0.023            |
| \( \rho \) (10^{-3}) | 5.131 \pm 3.515            |
| \( f_s, \text{KMT} \) | 0.138 \pm 0.008           |
| \( f_b, \text{KMT} \) | 0.062 \pm 0.007           |

Table 5: Standard 2L1S Model for KMT-2019-BLG-2783

Figure 5. Data (color-coded by observatory) together with the prediction and residuals for the model of KMT-2019-BLG-2783 specified in Table 5. The caustic topology is shown in the insets.
We found that the Kumeu observations were not of sufficient quality to include them in the analysis. In addition, there were two other observatories, both in Chile, i.e., the Danish 1.5 m and SMARTS 1.3 m, that began observations on May 10 and 11, respectively, i.e., 5–6 days after $t_0$. We do not include these observations because they were taken too late to help constrain any of the event parameters.

![Figure 6](image-url) Data (color-coded by observatory) together with the predictions and residuals for the two planetary models of OGLE-2019-BLG-0249 specified in Table 7 with the caustic topologies shown as insets. Also shown are the 1L1S and 1L2S models.

| Parameters          | Planetary | Wide | Binary | Wide |
|---------------------|-----------|------|--------|------|
| $\chi^2_{\text{red}}$/dof | 1672.5/1682 | 1670.7/1682 | 1733.2/1682 | 1751.1/1682 |
| $t_0$ (HJD)         | 8607.442 ± 0.004 | 8607.436 ± 0.004 | 8607.446 ± 0.004 | 8607.488 ± 0.004 |
| $u_0$ (10^{-3})     | 30.934 ± 0.288 | 31.221 ± 0.290 | 32.388 ± 0.341 | 28.418 ± 0.245 |
| $r_0$ (days)        | 76.387 ± 0.624 | 76.596 ± 0.618 | 76.912 ± 0.615 | 84.974 ± 0.689 |
| $s$                 | 0.581 ± 0.010 | 1.663 ± 0.030 | 0.231 ± 0.003 | 5.657 ± 0.110 |
| $q$ (10^{-3})       | 5.918 ± 0.320 | 6.166 ± 0.335 | 153.281 ± 8.331 | 247.573 ± 17.120 |
| $\log q$            | -2.226 ± 0.023 | -2.209 ± 0.024 | -0.813 ± 0.024 | -0.603 ± 0.030 |
| $\alpha$ (rad)      | 5.287 ± 0.003 | 5.292 ± 0.003 | 2.974 ± 0.005 | 2.960 ± 0.004 |
| $\rho$ (10^{-3})    | 9.635 ± 1.582 | 10.692 ± 1.466 | 17.631 ± 0.751 | 16.683 ± 0.645 |
| $f_{S,OGLE}$        | 0.718 ± 0.007 | 0.722 ± 0.007 | 0.720 ± 0.007 | 0.715 ± 0.007 |
| $f_{B,OGLE}$        | 0.219 ± 0.007 | 0.216 ± 0.007 | 0.217 ± 0.006 | 0.220 ± 0.007 |
Figure 7. Scatter plot on the $\pi_b$ plane derived from the MCMC of the four parallax models of OGLE-2019-BLG-0249 presented in Table 7, color-coded (red, yellow, green, cyan, blue) for $\Delta \chi^2 < (1, 4, 9, 16, 25)$. The contours are effectively 1D, with typical widths $\sigma_1 \sim 0.0080$ and 0.0065, for the close and wide solutions, respectively. The black contours show the mean and covariances that are used in Section 5.4.

Table 7: Survey + Follow-up 2LIS Models for OGLE-2019-BLG-0249

| Parameters                  | Standard | Close $u_0 > 0$ | Close $u_0 < 0$ | Wide $u_0 > 0$ | Wide $u_0 < 0$ |
|-----------------------------|----------|----------------|----------------|----------------|----------------|
| $\chi^2_{\text{tot}}/\text{dof}$ | 2101.7/2081 | 1828.1/2077 | 1828.2/2077 | 2100.7/2081 | 1831.4/2077 |
| $v_0$ (HJD)                 | 8607.430 ± 0.003 | 8607.415 ± 0.011 | 8607.414 ± 0.011 | 8607.426 ± 0.003 | 8607.807 ± 0.017 | 8607.805 ± 0.019 |
| $u_0$ (10$^{-1}$)           | 31.133 ± 0.269 | 30.859 ± 0.366 | -30.911 ± 0.369 | 31.363 ± 0.277 | 39.021 ± 0.490 | -39.064 ± 0.512 |
| $v_b$ (days)                | 75.628 ± 0.585 | 76.264 ± 0.645 | 76.309 ± 0.687 | 75.923 ± 0.597 | 76.592 ± 0.598 | 76.649 ± 0.647 |
| $s$                         | 0.554 ± 0.004 | 0.543 ± 0.010 | 0.545 ± 0.010 | 1.755 ± 0.014 | 1.777 ± 0.014 | 1.775 ± 0.015 |
| $q$ (10$^{-1}$)             | 6.643 ± 0.195 | 7.461 ± 0.270 | 7.455 ± 0.268 | 6.983 ± 0.214 | 7.819 ± 0.229 | 7.805 ± 0.237 |
| $\langle \log q \rangle$   | -2.177 ± 0.013 | -2.127 ± 0.015 | -2.127 ± 0.016 | -2.156 ± 0.013 | -2.106 ± 0.013 | -2.107 ± 0.013 |
| $\omega$ (rad)              | 5.295 ± 0.002 | 5.298 ± 0.012 | 5.298 ± 0.012 | 5.302 ± 0.003 | 5.302 ± 0.004 | 5.302 ± 0.005 |
| $\rho$ (10$^{-3}$)          | 5.138 ± 1.308 | 5.925 ± 1.231 | 6.084 ± 1.197 | 6.375 ± 1.057 | 7.369 ± 0.803 | 7.413 ± 0.804 |
| $\pi_b N$                   | 0.016 ± 0.076 | 0.008 ± 0.082 | -0.003 ± 0.068 | -0.005 ± 0.075 | 0.056 ± 0.019 | 0.056 ± 0.018 |
| $\pi_b E$                   | 0.061 ± 0.019 | 0.059 ± 0.020 | 0.056 ± 0.019 | 0.056 ± 0.019 | 0.056 ± 0.019 | 0.056 ± 0.019 |
| $ds/dt$ (yr$^{-1}$)         | -0.339 ± 0.688 | -0.249 ± 0.744 | 0.048 ± 0.528 | 0.051 ± 0.527 | 0.052 ± 0.527 | 0.052 ± 0.527 |
| $d\sigma/dt$ (yr$^{-1}$)    | 0.419 ± 1.419 | -0.569 ± 1.469 | -0.002 ± 0.303 | -0.026 ± 0.308 | 0.026 ± 0.308 | 0.026 ± 0.308 |
| $f_b, \text{OGLE}$          | 0.727 ± 0.007 | 0.719 ± 0.008 | 0.720 ± 0.008 | 0.730 ± 0.007 | 0.724 ± 0.007 | 0.724 ± 0.007 |
| $f_b, \text{OGLE}$          | 0.211 ± 0.006 | 0.218 ± 0.007 | 0.218 ± 0.007 | 0.208 ± 0.006 | 0.213 ± 0.007 | 0.213 ± 0.007 |

3.5.4. Survey + Follow-up Analysis

Table 7 shows the parameters after incorporating the follow-up data into the fit. The values of $q$ increase by about 10%, corresponding to $\sim 2\sigma$, which is not surprising given that the additional data are concentrated on the anomaly. The changes in $s$ are similar. From the comparison of Tables 6 and 7, the most puzzling (and potentially most consequential) change is that $\rho$ drops by a factor of $\sim 2$ without much change in the error bar. We investigate this and find that these three parameters are tightly correlated, which is very plausible given that they are all derived from the same short feature in the light curve, so that the three parameter changes are all expressions of the same additions to the data set.

3.5.5. Parallax Analysis

Because the event is long ($v_b \sim 75$ days) and reaches relatively high magnification ($A_{\text{max}} \sim 30$), and because the source is relatively bright ($I_s \sim 18$), it is plausible that
substantial parallax information can be extracted. We therefore add four parameters, i.e., $\pi_E$ and $\gamma$, and report the results in Table 7. A scatter plot of the MCMC on the $\pi_E$ plane is shown in Figure 7 for each of the four solutions. As in the case of KMT-2019-BLG-0298, the contours are essentially 1D, with axis ratios of $\sim 10$. However, contrary to that case, even the long axes of the error ellipses are relatively small, $\sigma_\perp \sim 0.08$, which is comparable to the offsets of the 1D contours from the origin. Hence, the argument given in Section 3.2 for adopting the standard-model parameters (but incorporating the $\pi_E$ constraints) does not apply, and we therefore use the full parallax solutions from Table 7 when we carry out the Bayesian analysis in Section 5.4.

3.6. OGLE-2019-BLG-0679

Figure 8 shows a roughly 10 day bump, which peaks at $t_{\text{anom}} \approx 8707.6$ and is itself punctuated by a shorter 2 day bump on its falling wing, all on the falling wing of a microlensing event that peaks at $t_0 = 8660.7$. When this anomaly is excised, a 1L1S fit (assuming no blending, as is plausible for such a
bright source) yields $u_0 = 0.87$ and $t_E = 31$ days. Hence, $\tau_{\text{anom}} = 1.51$, $u_{\text{anom}} = 1.74$, $s_1 = 2.20$, and $\alpha = 30^\circ$.

The grid search returns only one solution, whose refinement is shown in Table 8. The value of $\alpha$ is in good agreement with the heuristic prediction, while the fitted value of $s_{\text{inner}} = 2.22$ is in surprising agreement with $s_1$, given that the anomaly does not appear to be caustic-crossing. Figure 8 shows that the solution has an “inner” topology. In fact, if we had used the fit values for $u_0$ and $t_E$ (as opposed to those assuming no blending), we would have derived $s_1 = 2.13$, which would suggest that there might be another solution at $s_{\text{outer}} = (s_1)^2/s_{\text{inner}} = 2.04$. However, it is clear from the caustic topology in Figure 8 that the peak of the bump is due to the source passing the on-axis cusp, and the shorter, postpeak bump is due to passage of the off-axis cusp. Hence, in a hypothetical “outer” solution, this extra bump would occur before the peak of the main bump. Thus, there is no degeneracy.

Although $\rho$ is not measured, the constraints on $\rho$ are of some interest. That is, we will show in Section 4.5 that $\theta_\text{ast} \approx 7.0$ $\mu$as, so the $3\sigma$ limit, $\rho < 0.03$, rules out $\mu_{\text{rel}} < 2.8$ mas yr$^{-1}$, which is a reasonably well-populated part of parameter space. We will therefore incorporate the $\rho$ envelope function when we carry out the Bayesian analysis in Section 5.5.

Because the source is relatively bright ($I_S \approx 17.3$) and the anomaly is long after the peak and has two features that are separated by 4 days, we attempt a parallax analysis. That is, while in many cases, the change in source trajectory induced by parallax could be compensated for (or mimicked) by lens orbital motion, this is much more difficult when the model must accommodate additional light-curve features; see, e.g., An & Gould (2001).

The results are shown in Table 8 and illustrated in the $\pi_E$ scatter plot from the MCMC in Figure 9. Including parallax and orbital motion improves the fit $\Delta \chi^2 = 29$. Nevertheless, as we explain in some detail in Section 5.5, we will adopt the standard-model parameters for the purposes of this paper. However, we document the details of the parallax fit here in anticipation that they will be useful when the ground- and space-based parallax fits are later integrated. While we do not know what the space-based parallax fits will reveal, we do note that preliminary reduction of the Spitzer data shows a fall of $\sim 20$ flux units over 37 days, which should be enough to strongly constrain the parallax. In brief, when quoting parameters from this paper, only the “standard” column in Table 8 should be used.

3.7. OGLE-2019-BLG-0344

Figure 10 shows a moderately high magnification microlensing event peaking at $t_0 = 8567.53$ and punctuated by a short dip that is almost exactly at peak. A LL1S fit to the data (with the anomaly excluded) yields $u_0 = 0.10$ and $t_E = 14$ days. Hence, $\tau_{\text{anom}} = 0$, $u_{\text{anom}} = 0.1$, $s_1 = 0.95$, and $\alpha = 270^\circ$.

A grid search does indeed return two planetary solutions whose refinements are shown in Table 9 and that are in good agreement with these predictions, i.e., $s_1 = \sqrt{s_{\text{inner}} s_{\text{outer}}} = 0.95$ and $\alpha = 270^\circ$. However, it also returns six other solutions. Before discussing these, we first note that the planetary solutions are somewhat suspicious in that they have relatively large values of $\rho \approx 0.06$. We will show in Section 4.6 that $\theta_\text{ast} \approx 1.03$ $\mu$as. If these solutions are correct, they would therefore imply $\theta_E = 17$ $\mu$as and $\mu_{\text{rel}} = 0.44$ mas yr$^{-1}$. The first of these falls in the category of “exciting if true,” while the second has a relatively implausible $p = 0.5\%$ probability.
according to Equation (9). Therefore, we also show for comparison the solutions with $\rho = 0$, which are disfavored by $\Delta \chi^2 = 8$.

The six other solutions come in three pairs, which each approximately obey the close/wide degeneracy (Dominik 1999). We label these pairs (A, D), (B, E), and (C, F). The close solutions are given in Table 10 and illustrated in Figure 11. One of these also has an implausibly large $\rho$, so we show the $\rho = 0$ solutions in all cases. The bottom line is that if we consider the free $\rho$ case, then local B is preferred over either planetary

![Data (color-coded by observatory) together with the predictions and residuals for the two planetary models of OGLE-2019-BLG-0344 specified in Table 9, as well as the 1L2S model specified in Table 11. All three models can explain the data equally well. In addition, there are six other nonplanetary 2L1S models, most of which can also explain the data equally well; see Figure 11. Hence, this event cannot be cataloged as “planetary.”](image-url)

**Table 9**

Planetary Models for OGLE-2019-BLG-0344

| Parameters | Inner $\rho = 0$ | Outer $\rho = 0$ |
|------------|------------------|------------------|
| $\chi^2$/dof | 1612.8/1621 | 1620.2/1622 |
| $t_0$ (HJD) | 8567.532 ± 0.014 | 8567.529 ± 0.014 |
| $\nu_0$ | 0.103 ± 0.005 | 0.102 ± 0.004 |
| $r_e$ (days) | 14.413 ± 0.470 | 14.108 ± 0.366 |
| $q$ | 3.73 ± 0.528 | 2.938 ± 0.457 |
| $\langle \log q \rangle$ | −1.627 ± 0.096 | −1.530 ± 0.067 |
| $\alpha$ (rad) | 4.714 ± 0.023 | 4.706 ± 0.021 |
| $\rho$ ($10^{-5}$) | 5.809 ± 1.166 | 6.269 ± 1.097 |
| $f_{S, OGLE}$ | 0.365 ± 0.021 | 0.380 ± 0.018 |
| $f_{B, OGLE}$ | 0.063 ± 0.021 | 0.048 ± 0.018 |
solution by $\Delta \chi^2 = 6$, while if we consider the $\rho = 0$ case, then local A is within $\Delta \chi^2 < 1$ of either planetary solution. Hence, there is no reason to believe that the companion is a planet rather than another star. To avoid clutter, we do not present a table or figure for the three wide solutions, but the situation is qualitatively similar.

Finally, we investigate the 1L2S models, which are shown in Table 11. In this case, the apparent “dip” is the result of two
sources of nearly equal brightness successively passing the lens with nearly equal impact parameters and an interval of 2.0 days. The values of $\rho_1$ and $\rho_2$ are each poorly measured, and if we were to take them at face value, the two stars would nearly overlap in projection. Hence, we also consider the $\rho_1 = \rho_2 = 0$ case. This has the best $\chi^2$ for any of the $\rho = 0$ cases.

We conclude that the lens-source system could be either 1L2S or 2L1S, and, if the latter, the lens could equally well be planetary or binary in nature. Hence, we strongly counsel against classifying this event as “planetary.”

### 3.8. KMT-2019-BLG-0304

In many ways, KMT-2019-BLG-0304 is very similar to OGLE-2019-BLG-344 (Section 3.7), except that the anomaly near peak is a bump rather than a dip. Figure 12 shows a moderately high magnification microlensing event peaking at $t_0 = 8574.0$ and punctuated by a short bump at $t_{\text{anom}} = 8574.5$, i.e., just $\Delta t_{\text{anom}} = 0.5$ days after peak. The source is extremely faint, $I_S \sim 23$, which implies (e.g., Yee et al. 2012) that in the 1L1S and 2L1S fits, the parameter combinations $t_{\text{eff}} \equiv u_0t_E$, $t_{\text{ast}} \equiv \rho t_E$, and $t_p \equiv q t_E$ will be much better determined than $(u_0, t_E, \rho, q)$. For the 1L1S fit, we find $t_{\text{eff}} = 13.7$ days. This implies $\alpha = \tan^{-1}(t_{\text{eff}}/\Delta t_{\text{anom}}) = 88^\circ$, which is independent of $t_E$. On the other hand, the prediction for $s_1$ does depend on $t_E$. Noting that $u_{\text{anom}} = u_0/\sin \alpha \simeq u_0$, 

| Parameters          | Free $\rho$ | $\rho = 0$ |
|---------------------|-------------|------------|
| $\chi^2_{\text{red}}$/dof | 1609.7/1620 | 1613.7/1622 |
| $t_0,1$ (HJD')     | 8566.544 ± 0.071 | 8566.528 ± 0.069 |
| $t_0,2$ (HJD')     | 8568.511 ± 0.069 | 8568.506 ± 0.065 |
| $u_{0,1}$           | 0.082 ± 0.006  | 0.077 ± 0.004   |
| $u_{0,2}$           | 0.092 ± 0.012  | 0.080 ± 0.005   |
| $t_E$ (days)       | 15.215 ± 0.440 | 15.394 ± 0.430  |
| $\rho_1$ ($10^{-2}$) | 4.432 ± 2.775 | 4.32 ± 2.775    |
| $\rho_2$ ($10^{-2}$) | 7.221 ± 2.885 | 7.121 ± 2.885   |
| qF                  | 1.032 ± 0.189  | 1.054 ± 0.180   |
| $f_{S, \text{OGLE}}$ | 0.328 ± 0.015  | 0.320 ± 0.014   |
| $f_{B, \text{OGLE}}$ | 0.101 ± 0.015  | 0.108 ± 0.014   |

Figure 12. Data (color-coded by observatory) together with the predictions and residuals for the two planetary models of KMT-2019-BLG-0304 specified in Table 12, as well as the 1L2S model specified in Table 13. All three models can explain the data equally well. Hence, this event cannot be cataloged as “planetary.”
\[ s^t = \frac{1}{2} \left( \sqrt{\frac{u^2_0}{\sin^2 \alpha}} + \frac{u^0_0}{\sin \alpha} \right) \rightarrow 1 + \frac{u^0_0}{2} + \frac{u^2_0}{8}. \] (10)

| Parameters | Close | \( \rho = 0 \) | Wide | \( \rho = 0 \) |
|------------|-------|----------------|------|----------------|
| \( \chi^2_{\text{red}}/\text{dof} \) | 2394.3/2387 | 2394.9/2388 | 2393.8/2387 | 2394.8/2388 |
| \( t_0 \) (HJD') | 8574.014 ± 0.346 | 8574.020 ± 0.361 | 8574.019 ± 0.341 | 8574.039 ± 0.327 |
| \( u_0 \) | 0.083 ± 0.024 | 0.098 ± 0.048 | 0.099 ± 0.029 | 0.093 ± 0.030 |
| \( t_E \) (days) | 166.981 ± 45.725 | 144.948 ± 62.216 | 140.494 ± 38.089 | 151.043 ± 45.234 |
| \( s \) | 0.713 ± 0.033 | 0.722 ± 0.052 | 1.546 ± 0.051 | 1.569 ± 0.051 |
| \( q \) (10^{-3}) | 1.220 ± 0.340 | 1.415 ± 0.445 | 1.494 ± 0.422 | 1.496 ± 0.435 |
| \( \langle \log q \rangle \) | −1.917 ± 0.122 | −1.863 ± 0.138 | −1.838 ± 0.139 | −1.828 ± 0.127 |
| \( \alpha \) (rad) | 1.536 ± 0.025 | 1.537 ± 0.024 | 1.538 ± 0.024 | 1.536 ± 0.023 |
| \( \rho \) (10^{-3}) | 1.667 ± 1.008 | 1.617 ± 0.633 | 1.954 ± 1.211 | 0.011 ± 0.004 |
| \( f_{\text{fs, KMT}} \) | 0.009 ± 0.003 | 0.011 ± 0.006 | 0.011 ± 0.004 | 0.010 ± 0.003 |
| \( f_{\text{fs, KMTC}} \) | 0.180 ± 0.002 | 0.177 ± 0.004 | 0.176 ± 0.002 | 0.177 ± 0.002 |
| \( t_{\text{rel}}(\rho t_E) \) | 14.098 ± 0.796 | 14.581 ± 1.025 | 14.393 ± 0.887 | 14.492 ± 0.820 |
| \( t_{\text{rel}}(\rho t_E) \) | 0.299 ± 0.164 | 0.299 ± 0.164 | 0.299 ± 0.164 | 0.299 ± 0.164 |
| \( f_{\text{fs}} \) | 2.044 ± 0.411 | 2.018 ± 0.466 | 2.142 ± 0.447 | 2.305 ± 0.463 |
| \( f_{\text{fs}} \) | 1.578 ± 0.084 | 1.631 ± 0.137 | 1.633 ± 0.101 | 1.635 ± 0.094 |

If \( \rho \) can be measured from the light curve, then one can use standard techniques (Yoo et al. 2004) to determine the angular source radius, \( \theta_{\text{ast}} \), and so infer \( \theta_E \) and \( \mu_{\text{rel}} \):

\[ \theta_E = \frac{\theta_{\text{ast}}}{\rho}, \quad \mu_{\text{rel}} = \frac{\theta_{\text{E}}}{t_{\text{E}}} \]

However, in contrast to the majority of published by-eye discoveries (but similar to most of the new AnomalyFinder discoveries reported in Zang et al. 2021, 2022, 2023; Gould et al. 2022; Hwang et al. 2022; Jung et al. 2022), most of the planetary events reported in this paper have only upper limits on \( \rho \), and these limits are mostly not very constraining. As discussed by Gould et al. (2022), in these cases, \( \theta_{\text{ast}} \) determinations are not likely to be of much use, either now or in the future. Nevertheless, the source color and magnitude measurements that are required inputs for these determinations may be of use in the interpretation of future high-resolution observations, either by space telescopes or AO on large ground-based telescopes (Gould 2022). Hence, like Gould et al. (2022), we calculate \( \theta_{\text{ast}} \) in all cases.

**4. Source Properties**

As in Section 3.1, above, we begin by reproducing (with slight modification) the preamble to Section 4 of Jung et al. (2022). Again, this is done for the convenience of the reader. Readers who are familiar with Jung et al. (2022) may skip this preamble.
Our general approach is to obtain pyDIA (Albrow 2017) reductions of KMT data at one or possibly several observatory/field combinations. These yield the microlensing light curve and field-star photometry on the same system. We then determine the source color by regression of the $V$-band light curve on the $I$-band light curve. For the $I$-band source magnitudes, we adopt the values and errors from the parameter tables in Section 3 after aligning the reporting system (e.g., OGLE-IV or KMT pySIS) to the pyDIA system via regression of the $I$-band light curves. While Gould et al. (2022) were able to calibrate the KMT pyDIA color–magnitude diagrams (CMDs) using published field-star photometry from OGLE-III (Szymański et al. 2011) or OGLE-II (Udalski et al. 1997; Udalski et al. 2002; Szymański 2005), only three of the seven subprime field events in this paper are covered by these catalogs. Hence, for the remaining four, we work directly in the KMT pyDIA magnitude system. Because the $\theta_{\text{ast}}$ measurements depend only on photometry relative to the clump, they are unaffected by calibration. In the current context, calibration is only needed to interpret limits on lens light. Where relevant, we carry out an alternative approach to calibration.

We then follow the standard method of Yoo et al. (2004). We adopt the intrinsic color of the clump $(V - I)_{0, \text{cl}} = 1.06$ from Bensby et al. (2013) and its intrinsic magnitude from Table 1 of Natálf et al. (2013). We obtain $[(V - I), I_{S,0}] = [(V - I), I_{S}] + [(V - I), I_{L,0}] - [(V - I), I_{L}]$. We convert from $V/I$ to $V/K$ using the $VIK$ color–color relations of Bessell & Brett (1988) and then derive $\theta_{\text{ast}}$ using the relations of Kervella et al. (2004a, 2004b) for giant and dwarf sources, respectively. After propagating the errors, we add 5% in quadrature to account for errors induced by the overall method. These calculations are shown in Table 14. Where there are multiple solutions, only the one with the lowest $\chi^2$ is shown. However, the values of $\theta_{\text{ast}}$ can be inferred for the other solutions by noting the corresponding values of $I_{S}$ in the event-parameter tables and using $\theta_{\text{ast}} \propto 10^{-I_{S}/5}$. In any case, these are usually the same within the quoted error bars.

Where relevant, we report the astrometric offset of the source from the baseline object.

Comments on individual events follow, where we also note any deviations from the above procedures.

### 4.1. KMT-2019-BLG-0298

The positions of the source and clump centroid are shown in blue and red, respectively, in Figure 13, together with the background of neighboring field stars. The blended light is consistent with zero, so it is not represented in the CMD. The source position (derived from difference imaging) is offset from the baseline object by 24 mas, which is consistent with the measurement error.

On the other hand, the 1σ error on the blended flux is about 7% of the source flux, which would correspond to $I_B \sim 20.4$. According to the KMT website, there are $A_I \sim 2.9$ mag of extinction toward this line of sight, while the mean distance modulus of the bar is 14.50 (Natálf et al. 2013). Hence, a bulge lens star that saturated this 1σ limit would have $M_I \sim 3.2$, implying that no useful limit can be placed on flux from the lens.

While the normalized source size is not well measured, it is constrained to be $\rho < 0.075$ at 3σ. From the values of $\theta_{\text{ast}} = 6.12$ mas and $t_E = 27.7$ days in Tables 14 and 2, we therefore obtain $\theta_E > 0.082$ mas and $\mu_{\text{rel}} > 1.1$ mas yr$^{-1}$. As can be seen from Equation (9), this is only marginally constraining. Nevertheless, we will use the $\rho$ envelope function in Section 5.1 to constrain the Bayesian analysis.

Finally, we note that Gaia (Gaia Collaboration et al. 2016, 2018) reports a source proper motion of

$$\mu_S(N, E) = (-5.91 \pm 0.41, -4.20 \pm 0.64) \text{mas yr}^{-1} \quad \text{(Gaia).}$$

There are two reasons for mild caution regarding this result. First, the same solution yields a 3.6σ negative parallax, $\pi_S = -1.696 \pm 0.469$ mas. Second, the Gaia RUWE parameter is 1.25. It is extremely unlikely that the large negative parallax is due to normal statistical fluctuations if one interprets the error bars naively. Jung et al. (2022) showed that high RUWE numbers are indicative of spurious source proper motions in microlensing events. While these high values were all above 1.7, i.e., far above the RUWE value of 1.25 for KMT-2019-BLG-0298, it is still the case that this RUWE value is somewhat above average. Noting that Rybicki et al. (2022) found that Gaia errors in microlensing fields are typically underestimated by a factor of 2, we accept the estimate of Equation (12), but double the error bars. With this revision, the negative parallax becomes $<2\sigma$.

We note that the source is a typical bulge clump giant, both from its proper motion and its position on the CMD.

### 4.2. KMT-2019-BLG-1216

The positions of the source and clump centroid are shown in blue and red, respectively, in Figure 14, while the blended light is shown in green.

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**Table 14**

| Name                  | $(V - I)_S$ | $(V - I)_L$ | $(V - I)_{S,0}$ | $I_S$     | $I_L$     | $I_{L,0}$ | $I_{S,0}$ | $\theta_{\text{ast}}$ (mas) |
|-----------------------|-------------|-------------|-----------------|-----------|-----------|-----------|-----------|--------------------------|
| KMT-2019-BLG-0298     | 3.67 ± 0.06 | 3.56 ± 0.03 | 1.11 ± 0.07     | 17.69 ± 0.06 | 17.53 ± 0.03 | 14.42     | 14.58 ± 0.07 | 6.117 ± 0.53       |
| KMT-2019-BLG-1216     | N.A.        | 1.91 ± 0.03 | 1.04 ± 0.14     | 21.52 ± 0.25 | 15.78 ± 0.05 | 14.61     | 20.35 ± 0.25 | 0.396 ± 0.046       |
| KMT-2019-BLG-2783     | 1.31 ± 0.11 | 1.70 ± 0.03 | 0.67 ± 0.11     | 20.18 ± 0.06 | 15.32 ± 0.05 | 14.59     | 19.45 ± 0.08 | 0.387 ± 0.051       |
| OGLE-2019-BLG-0249    | 2.71 ± 0.01 | 2.65 ± 0.02 | 1.10 ± 0.02     | 18.47 ± 0.01 | 16.75 ± 0.04 | 14.62     | 16.34 ± 0.04 | 2.695 ± 0.155       |
| OGLE-2019-BLG-0679    | N.A.        | 4.20 ± 0.03 | 0.94 ± 0.10     | 17.48 ± 0.06 | 17.99 ± 0.05 | 14.41     | 13.90 ± 0.08 | 7.004 ± 0.958       |
| OGLE-2019-BLG-0344    | 1.93 ± 0.04 | 2.29 ± 0.04 | 0.70 ± 0.06     | 18.80 ± 0.03 | 15.99 ± 0.05 | 14.59     | 17.40 ± 0.06 | 1.026 ± 0.090       |
| KMT-2019-BLG-0304     | N.A.        | 4.58 ± 0.05 | 0.74 ± 0.07     | 22.89 ± 0.36 | 19.07 ± 0.08 | 14.60     | 18.42 ± 0.37 | 0.680 ± 0.128       |

Note. $(V - I)_{L,0} = 1.06$. 

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21 This site uses the $A_K$ map of Gonzalez et al. (2012) and assumes $A_I = 7 A_K$. 

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Our procedures differ substantially from most other events in this paper. First, we do not obtain a reliable source color from regression because the $V$-band signal is too weak. Therefore, to determine the source position on the CMD, it is unnecessary to make use of the pyDIA reductions. Instead, we go directly from the OGLE-IV value and error shown in Table 3 to the calibrated OGLE-III system by finding the $I$-band offset between OGLE-III and OGLE-IV from comparison stars. We then find the offset relative to the clump (Table 14) and infer from this offset the $(V-I)_{S,0}$ intrinsic color using the Hubble Space Telescope (HST) CMD from Baade’s Window (Holtzman et al. 1998).

To find $I_B$, we subtract this source flux from the flux of the baseline object in the OGLE-III catalog, $I_{\text{base}} = 20.15$. Unfortunately, there is no color measurement for this object in the OGLE-III catalog. Therefore, to estimate its color, we first identify its counterpart in the KMTC pyDIA catalog. After transforming the photometry to the OGLE-III system, we find agreement for $I_{\text{base}}$ within 0.03 mag. Therefore, we transform the pyDIA $(V-I)_{\text{base}}$ into the OGLE-III system and then proceed to find $(V-I)_B$ in the usual way.

The baseline object is offset from the source by 120 mas, which means that it cannot be the lens and, in fact, cannot be dominated by the lens. If there were no errors in the estimates of $I_{\text{base}}$ and $I_S$, this would imply that the blend flux would place a very conservative upper limit on the lens flux. In fact, $I_S$ has a 0.25 mag error from the modeling, although this has only a small effect on $I_B$ because $>70\%$ of the baseline light comes from the blend. The error in the DoPhot (Schechter et al. 1993) photometry of the baseline object is of greater concern. While it is encouraging that OGLE-III and KMTC pyDIA agree closely on this measurement, both could be affected by the mottled background of these crowded fields (Park et al. 2004).

When calculating $\theta_{\text{ast}}$, we take into account the correlation between $I_{S,0}$ and $(V-I)_{S,0}$ in the above-described color–magnitude relation method. That is, at each possible offset (i.e., taking into account the 0.25 mag error in $I_{S,0}$), we allow for a 0.1 mag spread in $(V-I)_{S,0}$, centered on the value for that $I_{S,0}$. Then we consider the ensemble of all such estimates within the quoted error of $I_{S,0}$.

4.3. KMT-2019-BLG-2783

The positions of the source and clump centroid are shown in blue and red, respectively, in Figure 14. After transforming the source flux to the OGLE-III system and comparing to the OGLE-III baseline object, we find that the blended light is consistent with zero. From the pyDIA analysis, we find that the source is offset from the baseline object by only 14 mas, which is consistent with zero within the measurement errors.

We find that all values of $\rho < (8, 12, 15) \times 10^{-3}$ are consistent at $(1, 2, 3)\sigma$. Given the values $\theta_{\text{ast}} = 0.39$ mas from Table 14 and $t_E = 23.6$ days from Table 5, these values correspond to $\mu_{\text{red}} > (0.75, 0.50, 0.40)$ mas yr$^{-1}$. These are virtually unconstraining according to Equation (9). Nevertheless, we will include a $\rho$-envelope function in the Bayesian analysis of Section 5.3.
Based on the absence of blended light, we set the limit on lens flux at half the source flux, i.e., \( I_L > I_S + 0.75 = 20.93 \).

### 4.4. OGLE-2019-BLG-0249

The positions of the source and clump centroid are shown in blue and red, respectively, in Figure 13. The blended light (green) cannot be determined from the KMT pyDIA analysis because there is no true “baseline” during 2019. Rather, we find the blended flux from OGLE-IV and transform to the pyDIA system, \( I_B = 19.75 \pm 0.05 \). While the source color is determined with high precision from this very bright event, the error in the blend color is very large, \( (V - I)_B = 2.70 \pm 0.22 \). Nevertheless, the color plays no significant role because, as we will show, the blend is unlikely to be related to the event.

Using a special pyDIA reduction with a late-season-based template, we find that the source position (derived from difference images) is offset from the baseline object by 53 mas. Taking into account the fact that these late-season images are still magnified by \( A \approx 1.06 \), this separation should be corrected to \( \Delta \theta_{S,\text{base}} = 56 \) mas. This implies that the separation of the source (and thus lens) from the blend is \( \Delta \theta_{S,B} = (1 + f_S/f_0) \Delta \theta_{S,\text{base}} = 240 \) mas. Hence, on the one hand, it cannot be the lens and is very unlikely to be a companion to either the source or the lens. On the other hand, it lies well within the point-spread function, so it is undetectable in seeing-limited images.

When \( \pi_E \) and \( \gamma \) are included in the fits, there are well-defined minima in \( \rho \) for the wide solutions but less so for the close solutions; see Figure 15. Hence, we will use \( \rho \)-envelope functions in the Bayesian analysis of Section 5.4 in all cases.

We set the limit on lens flux as that of the blended flux, i.e., \( I_L > I_{BC,\text{Cousins}} \approx I_{B,\text{pyDIA}} - 0.1 = 19.65 \), where we have estimated (based on other events) a 0.1 mag offset between the pyDIA and standard systems. Note that while it is true that the blended flux could be underestimated due the mottled background of crowded bulge fields, it is also the case that the lens flux can comprise no more than three-fourths of all of the blended flux; otherwise, the remaining light would be so far from the source as to be separately resolved.

Finally, we adopt the Gaia proper-motion measurement, 
\[
\mu_s(N, E) = (-5.29 \pm 0.39, -3.92 \pm 0.77) \text{mas yr}^{-1} \quad \text{(Gaia)},
\]

noting that it has a RUWE value of 1.00.

### 4.5. OGLE-2019-BLG-0679

The positions of the source and clump centroid are shown in blue and red, respectively, in Figure 13. According to the better \( u_0 < 0 \) solution in Table 8, blended light is detected at 2.5\( \sigma \). For Gaussian statistics, this would have a low false-alarm probability, \( p \approx 0.7\% \). Nevertheless, as we now discuss, we treat this detection cautiously.

The key point is that the offset between the source and the baseline object is only 17 mas, which is consistent with zero within the measurement error. This would naturally be explained if there were no blended light or, as a practical matter, much less than is recorded in Table 8. For principle, it might also be explained by the blend being associated with the event, either the lens or a companion to the lens or the source. However, this possibility is itself somewhat problematic. That is, this is a heavily extinguished field, \( A_I = 3.8 \), so if this blend is
behind most of the dust, then $I_{B,0} \sim 15.7$. Hence, if it is in the bulge (e.g., as a companion to the source or part of the lens system), then it is a giant, i.e., $M_I \sim 1.3$. And, to be an unevolved main-sequence lens (or companion to the lens), it would have to be at $D_L \lesssim 2\, \text{kpc}$. Of course, this is not impossible, but it is far from typical.

In addition, there is much experience showing that microlensing photometry does not obey Gaussian statistics, so the $p < 1\%$ false-alarm probability cannot be taken at face value. The combination of this reduced confidence with the low prior probability for so much blended light so close to the lens is what makes us cautious about this interpretation. While we adopt the source flux as measured by the fit (i.e., less than the baseline flux), we do not claim to have detected blended light; therefore, we do not show an estimate of the blend in Figure 13.

For both the standard and parallax fits, $\rho$ is poorly constrained. Hence, we will apply the $\rho$-envelope function in the Bayesian analysis of Section 5.5; see Figure 15.

The color measurement, which is tabulated in Table 14 and illustrated in Figure 13, presented some difficulties because the event is low amplitude and suffers heavy extinction. Both factors contribute to low flux variation in the $V$ band. Our usual approach, based on regression of the magnified event in KMTC data, yields $(V - I)_S = 4.01 \pm 0.12$. By comparison, the color of the baseline object, which has more than twice the flux of the difference object even at the peak of the event, has an identical central value but substantially smaller error, $(V - I)_\text{base} = 4.01 \pm 0.06$. This coincidence of color values would be a natural consequence of the zero-blending hypothesis, but the errors are too large to draw any strong conclusions. Note that both central values place the source $(V - I) = -0.19$ blueward of the clump.

We made two further efforts to clarify the situation. First, we made independent reductions of KMTC data. These produced a similar color but with an error bar that was more than twice as large. Combining KMTC and KMTS yields $(V - I)_S = 4.01 \pm 0.11$, which is not a significant improvement.

Second, we found the offset from the clump in $(I - H)$, making use of ANDICAM $H$-band data for the light curve and VVV data for the baseline-object and field-star $H$-band photometry, and we compared these to the offsets in $(V - I)$ using the color–color relations of Bessell & Brett (1988). For the baseline object, we find $\Delta(I - H)_\text{base} = -0.12 \pm 0.02$, corresponding to $\Delta(V - I)_\text{base} = -0.10 \pm 0.02$, which is marginally consistent with the direct $V/I$ measurement at 1.4σ. On the other hand, the regression of the $H$-band light curve leads to $\Delta(I - H)_S = +0.17 \pm 0.06$. This is inconsistent at 4.6σ with the color offset of the baseline object. In principle, $\Delta(I - H)$ need not be the same for the source and the baseline because the baseline can have a contribution from blended light of a different color. However, to explain such a large offset from just 15% of the $I$-band light would require an extraordinarily red blend. Considering, in particular, that the

![Figure 15. Envelope functions for $\chi^2$ vs. $\rho$ for all five planetary events in this paper.](image)
blend lies just 1.3 mag below the clump, we consider this to be very unlikely.

In the face of this somewhat contradictory evidence, we adopt \( \Delta(V-I)_S = -0.12 \pm 0.09 \). That is, first, given that the baseline light is dominated by the source, it provides the best first guidance to the source color. We then adopt a compromise value between the \( V/I \) and \( I/H \) determinations that is consistent with both at \( \sim 1\sigma \). This value is also well within the \( 1\sigma \) interval of the source-color determination in \( V/I \). For the error bar, we adopt the offset between these two baseline-object determinations, in recognition of the fact that they disagree by more than \( 1\sigma \). We consider that the \( I/H \) determination of the source color is most likely spurious.

Although unsatisfying, any errors in our adopted resolution of this issue do not have significant implications for the results reported in this paper. The source color (as well as the degree of blending) only impacts the \( \theta_{\text{tot}} \) determination, and at \( \lesssim 15\% \). This would be of some concern if we had a precise \( \rho \) measurement, in which case, it would impact \( \theta_{\text{tot}} \) at the same level. However, we basically have only an upper limit on \( \rho \), and this fact completely dominates the uncertainty in \( \theta_{\text{tot}} \). We have presented a thorough documentation of this issue mainly for reference, in case it becomes relevant to the interpretation of Spitzer data. That is, when the Spitzer data do not cover the peak of the light curve (as appears to be the case for OGLE-2019-BLG-0679), the parallax measurement can sometimes be substantially improved if the Spitzer source flux is independently constrained via a ground–Spitzer color–color relation, together with a ground-based color measurement. Hence, a thorough understanding of potential uncertainties in the latter can be of direct relevance.

We do not attempt to place any limit on the lens light. At the \( 2\sigma \) level, \( I_R > 18.8 \), which (assuming the blend lies behind most of the dust) corresponds to \( I_{R,0} > 15.0 \) and so is not constraining.

Finally, we note that Gaia reports a proper-motion measurement

\[
\mu_s(N, E) = (-6.04 \pm 0.28, -6.35 \pm 0.51) \text{mas yr}^{-1} \quad \text{(Gaia DR3).}
\]

However, it also reports a RUWE number, 1.75. Based on a systematic investigation of Gaia proper motions of microlensed sources, Jung et al. (2022) concluded that such high RUWE measurements were often spurious or at least suspicious. In the present case, caution is further indicated by the fact that the Gaia DR2 measurement, \( \mu_s(N, E) = (-6.98 \pm 0.79, -3.42 \pm 1.12) \text{mas yr}^{-1} \), is inconsistent with the DR3 measurement, even though they are mostly based on the same data.

These discrepancies lead us to make our own independent measurement of \( \mu_s \) based on almost 10 yr of OGLE-IV data, which yields

\[
\mu_s(N, E) = (-5.32 \pm 0.37, -8.36 \pm 0.16) \text{mas yr}^{-1} \quad \text{(OGLE - IV).}
\]

This measurement is strongly inconsistent with the Gaia DR3 measurement, casting further doubt upon the latter. However, as we discuss in Section 5.5, the OGLE-IV measurement is, similar to Gaia DR3, in significant tension with other information about the event.

Therefore, we do not incorporate any \( \mu_s \) measurement into the Bayesian analysis of Section 5.5. Nevertheless, as we will discuss in that section, the various estimates of \( \mu_s \) raise enough concerns about the microlensing \( \pi_\text{E} \) measurement as to convince us to report the “standard” (i.e., nonparallax) solution for our final values.

4.6. OGLE-2019-BLG-0344

As discussed in Section 3.7, this event has planetary solutions, but it cannot be claimed as a planet. Hence, the CMD analysis is presented solely for completeness. Because the source is consistent with being unblended in the planetary fits, we simply adopt the parameters of the OGLE-III baseline object as those of the microlensed source. These are shown as a blue circle in Figure 14, while the source centroid is shown as a red circle.

4.7. KMT-2019-BLG-0304

Due to heavy extinction, the red clump on the CMD is partially truncated by the \( V \)-band threshold. Therefore, we determine the height of the clump in the \( I \) band by matching the pyDia to the VVV catalog (Minniti et al. 2010, 2017) and then determining the \( (V-I) \) color from the portion of the red clump that survives truncation. This is shown as a red circle in Figure 13. We then determine the offset from the clump in the \( I \) band (Table 14) and apply the HST color–magnitude relation, as in Section 4.2. Note that the color of the red clump centroid plays no role in this calculation, and it is shown in Figure 13 only to maintain a consistent presentation with other events.

5. Physical Parameters

To make Bayesian estimates of the lens properties, we follow the same procedures as described in Section 5 of Gould et al. (2022). We refer the reader to that work for details. Below, we repeat the text from Section 5 of Jung et al. (2022) for the reader’s convenience.

In Table 15, we present the resulting Bayesian estimates of the host mass \( M_{\text{host}} \), planet mass \( M_{\text{planet}} \), distance to the lens system \( D_L \), and planet–host projected separation \( a_\perp \). For three of the five events, there are two or more competing solutions. For these cases (following Gould et al. 2022), we show the results of the Bayesian analysis for each solution separately, and we then show the “adopted” values below these. For \( M_{\text{host}}, M_{\text{planet}} \), and \( D_L \), these are simply the weighted averages of the separate solutions, where the weights are the product of the two factors at the right side of each row. The first factor is simply the total weight from the Bayesian analysis. The second is \( \exp(-\Delta \chi^2/2) \), where \( \Delta \chi^2 \) is the \( \chi^2 \) difference relative to the best solution. For \( a_\perp \), we follow a similar approach, provided that either the individual solutions are strongly overlapping or one solution is strongly dominant. If neither condition were met, we would enter “bimodal” instead. However, in practice, this condition is met for all three events for which there is potentially an issue. Note that in all cases (including those with only one solution), we have provided symmetrized error bars in the “adopted” solution for simplicity of cataloging. The reader interested in recovering the asymmetric error bars can do so from the table.

We present Bayesian analyses for five of the seven events but not OGLE-2019-BLG-0344 and KMT-2019-BLG-0304, for which we cannot distinguish between competing interpretations of the event; see Sections 3.7 and 3.8. Figures 16 and 17 show histograms of \( M_{\text{host}} \) and \( D_L \) for these five events.
5.1. KMT-2019-BLG-0298

As discussed in Section 3.2, we accept the event parameters from the standard (seven-parameter) solution in Table 2 but incorporate the $\pi_E$ constraints from the parallax–plus–orbital motion solution. Again, the reason for this is that the $\pi_E$ constraints are essentially 1D, so the parallax MCMC explores regions of very high $|\pi_E|$, which would be highly suppressed after incorporating Galactic priors.

In the Bayesian analysis, there are four constraints, i.e., on $t_E$, $\mu_S$, $\rho$, and $\pi_E$: The first is $t_E = 27.71 \pm 0.62$ days from Table 2. The second is $\mu(S,E) = (-5.91 \pm 0.82, -4.20 \pm 1.28) \text{ mas yr}^{-1}$ from Section 4.1. The third is given by $\exp(-\Delta \chi^2(\rho)/2)$, where $\Delta \chi^2(\rho)$ is the envelope function that is shown in Figure 15. For the fourth, we represent the $\pi_E$ scatter plots shown in Figure 2 as Gaussian ellipses (also illustrated in this figure) with means and covariance matrices derived from the MCMC. These have central values and error bars similar to those shown in Table 2 (based on medians) and correlation coefficients of 0.95 and 0.60 for the $u_0 > 0$ and $u_0 < 0$ solutions, respectively. They are highly linear structures with minor axes $\sigma_{\pi} = (0.039, 0.046)$ and axis ratios of $\sigma_1/\sigma_2 = (11.0, 12.3)$ for the respective cases.

The Bayesian estimates (Table 15 and Figure 16) favor $M \sim 0.7 M_\odot$ hosts that are in or near the bulge, i.e., small $\pi_E$. This preference is due to the $\pi_E$ constraint, which is, effectively, a 1D structure passing through the origin. Hence, for randomly oriented $\rho$ (so $\pi_E$), the fraction of surviving simulated events scales $\propto \pi_E^{-1}$, while the very weak constraints on $\rho$ (so $\pi_E$) imply that typical $\pi_E \sim 5 \text{ mas yr}^{-1}$ are favored, so $0 \lesssim \pi_E \lesssim 0.4 \text{ mas}$. Low $\pi_E$ then drives $\pi_E \approx \theta_E/\pi_E$ to the higher range of the available mass function. Nevertheless, the fact that the ($u_0 > 0$) solution for $\pi_E$ closely tracks the direction of Galactic rotation (i.e., $\sim 30^\circ$ north through east), combined with the fact that the source is measured to be moving at $\mu_S \sim 7.3 \text{ mas yr}^{-1}$ at $\sim -145^\circ$ (north through east, i.e., almost antirotation), permits disk hosts with very small $D_L$; see Figure 16.

5.2. KMT-2019-BLG-1216

As discussed in Section 3.3, we adopt a cautious attitude toward incorporating the $\pi_E$ measurement. That is, given the relatively high value of $\pi_E \sim 0.6$, the $\rho = 3\%$ false-alarm probability of this measurement would be too high to accept it for typical microlensing events, for which $\pi_E$ is generally much closer to zero. Therefore, we begin the Bayesian analysis using the standard (seven-parameter) solution in Table 3. There are then three constraints, i.e., on $t_E$ (from Table 3), $\rho$ (from the envelope function in Figure 15), and the lens flux, $I_L > 19.76$ from Section 5.2. The results are shown in Table 15 and illustrated in Figure 16.

The results favor nearby lenses $D_L \sim 3.5 \text{ kpc}$, corresponding to $\pi_{rel} \sim 0.17 \text{ mas}$. The reason is that while $\rho$ is not measured, it is constrained at, e.g., $2\sigma$ to be $\rho_{rel} \lesssim 5 \times 10^{-4}$, corresponding to $\theta_E > 0.47 \text{ mas}$. Because this is a long event, this threshold corresponds to $\mu_{rel} > 1.9 \text{ mas}$, which is moderately low. Hence, somewhat bigger $\theta_E$ are favored by Galactic kinematics, e.g., $\theta_E \sim 0.8 \text{ mas}$, which would also correspond to the weak minimum of the $\rho$-envelope function. Considering the “effective top” of the mass function $M \lesssim M_\odot$, these values respectively imply $\pi_{rel} \gtrsim 0.03$ and $\lesssim 0.08 \text{ mas}$. For $M \sim 0.5 M_\odot$, i.e., closer to the peak of the mass function, these values are doubled. Hence, nearby lenses are strongly favored, while a broad range of masses is permitted.

The Bayesian results do not give any reason to be suspicious of the $\pi_E$ measurement. The main takeaway from Figure 16 is that despite the powerful Galactic priors favoring bulge lenses (e.g., Batista et al. 2011), which tend to “override” the $\rho$ constraint, disk lenses are strongly favored. It is notable that the direction of $\pi_E$ for ($u_0 < 0$) is consistent with that of Galactic rotation at 1$\sigma$. While the central value of this solution, $\pi_E = 0.7 \pm 0.2$, is substantially higher than would be naively
indicated by the Bayesian analysis, the error is large. Therefore, we incorporate this result.

We find that the main effect of incorporating the $\pi_E$ measurement is to effectively eliminate the bulge and near-bulge lenses, which (as explained above) were previously allowed due to the Galactic priors “overriding” the $\rho$ constraint.

5.3. KMT-2019-BLG-2783

There is only one solution, upon which there are three constraints, i.e., on $t_E = 23.6$ days (from Table 5), $\rho$ (from the envelope function in Figure 15), and the lens flux, $I_L > 20.93$ from Section 4.3. However, given that the $2\sigma$ limit, $\rho < 0.011$, corresponds to $\mu_{\text{rel}} > 0.5 \text{ mas yr}^{-1}$, the $\rho$ constraint effectively plays no role. On the other hand, the lens-flux constraint combined with the low extinction ($A_V \sim 0.73$; see Table 14) eliminates solar-type lenses even in the bulge and then progressively eliminates increasingly less massive stars for increasingly nearby disk lenses. The net result is that the Bayesian results are compatible with a very broad range of distances but a mass distribution that is sharply curtailed at the high end; see Figure 16.

5.4. OGLE-2019-BLG-0249

There are four solutions (two parallax solutions for each of the close and wide topologies) on which there are five constraints, i.e., on $t_E$, $\rho$, $I_L$, $\mu_{\text{rel}}$, and $\pi_E$. The first comes from Table 7, the second is from the $\rho$-envelope functions discussed in Section 4.4 and shown in Figure 15, the third is $I_L > 19.65$ (from Section 4.4), and the fourth is from Gaia (Equation (13)). Finally, we characterize the $\pi_E$ constraints as 2D Gaussian distributions, whose $\Delta \chi^2 = 1$ contours are shown as black ellipses in Figure 7. These have central values and error bars similar to those shown in Table 7 (based on medians) and with correlation coefficients (0.91, 0.92, 0.95, 0.94) for the (close, $u_0 > 0$; close, $u_0 < 0$; wide, $u_0 > 0$; wide, $u_0 < 0$) solutions, respectively. They are highly linear structures with minor axes
σ_⊥ = (0.0080, 0.0079, 0.0064, 0.0067) and axis ratios of σ_∥/σ_⊥ = (9.9, 11.0, 12.0, 12.1) for the respective cases.

The result is that the host is very well constrained to be an upper main-sequence star that is in or near the bulge; see Table 15 and Figure 17. The reason that these constraints are much tighter than for any other event analyzed in this paper is that, while neither θ_E nor (the vector) π_E are well measured, both θ_E and (the scalar) π_E are reasonably well constrained.

In the case of θ_E, the ρ-envelope functions have relatively broad, but nonetheless well-defined, minima. It is true that these functions turn over for ρ ≲ 0.001 for the close solution. However, these values typically result in masses M ≳ 5 M_⊙ and so are excluded by the mass function. Regarding π_E, despite the high axis ratios mentioned above, the (scalar) π_E are well constrained (and to very similar values) in the four cases because the lines from the origin that are perpendicular to these linear structures all pass though the 1σ contour, with σ_∥ ≈ π_E, best.

5.5. OGLE-2019-BLG-0679

As foreshadowed in Sections 3.6 and 4.5, we ultimately decided to report final results (for this paper) based on the “standard” solution of Table 8. We detail our reasons for this decision at the end of this subsection.

Hence, there is only one solution on which there are two constraints, i.e., on f_0 and ρ. The first comes from Table 8, and the second comes from the ρ-envelope function discussed in Section 4.5 and shown in Figure 15.

Table 15 and Figure 17 show that the posterior distributions of both mass and distance are very broad, and the lens system can almost equally well reside in the bulge or disk. This is a consequence of the fact that the only measured constraint is f_0, while θ_E effectively has only a lower limit.

We made the decision to adopt the “standard” solution as follows. We first carried out Bayesian analyses for both the “standard” and “parallax” solutions to understand how they differ not only with respect to the parameters that we normally report (in Table 15) and display (in Figure 17) but also for the source proper motion, μ_S, which is normally considered a nuisance parameter. Before continuing, we note that including the parallax measurement somewhat reduced the estimates of the host mass and distance but left broad distributions for both.

We found that, regardless of which solution (μ_0 > 0 or μ_0 < 0) was correct, and regardless of whether the host was assumed to be in the disk or the bulge, both the Gaia DR3 and OGLE-IV measurements of μ_S were inconsistent at ≳ 2σ with the posterior μ_S distributions; see Figure 18. These tensions can be understood by considering the example of disk lenses in the u_0 > 0 solution (red), for which π_E, best ≳ 0.3 ± 0.1. Because μ_S and π_E have the same direction, this implies that μ_S, rel should also be positive. As the prior distributions of μ_S, hel, and μ_S, L are basically symmetric in b, while μ_S, hel,best = μ_S, hel,best - μ_S, hel,S, the posterior distributions are driven to positive and negative values for the lenses and sources, respectively.

This “conflict” may well have a perfectly reasonable explanation. As discussed in Section 4.5, the Gaia DR3 measurement may simply be wrong, as signaled by both its high RUWE number and its strong disagreement with both Gaia DR2 and OGLE-IV; see Figure 18. Similarly, the OGLE-IV measurement may be wrong. Alternatively, it may be that either Gaia DR3 or OGLE-IV is correct (or basically correct), that the host lies in the bulge (blue and cyan ellipses), and that the event characteristics are 2σ outliers. In addition, it could be that the π_E measurement suffers from unrecognized systematics. Given that any of these three explanations is possible in principle, they can lead to very different results, and the matter will probably be resolved within a year by the Spitzer parallax measurement, the most prudent course is to defer judgment.

22 Actually, what is directly relevant is μ_S, hel,best, but the difference, which is relatively small, is ignored here in the interest of simplicity.
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until the Spitzer data can be properly evaluated. This course will also minimize the possibility that confusion will propagate through the literature.

5.6. OGLE-2019-BLG-0344

Because there is no compelling reason to believe that the planetary solution is correct, we do not present a Bayesian analysis.

5.7. KMT-2019-BLG-0304

Because there is no compelling reason to believe that the planetary solution is correct, we do not present a Bayesian analysis.

6. Discussion

Here we have analyzed all five of the previously unpublished planets found by the KMT AnomalyFinder algorithm toward the 21 KMT subprime fields. We also analyzed the two events that have nonplanetary solutions but are consistent with planetary interpretations. Such events are rarely published, but they are a standard feature of the AnomalyFinder series because they can be important for understanding the statistical properties of the sample as a whole. A total of five such events were previously published from the 2018 season (Gould et al. 2022; Jung et al. 2022).

6.1. Summary of All 2019 Subprime AnomalyFinder Planets

Table 16 shows these five planets and two “possible planets” in the context of the ensemble of all such 2019 subprime AnomalyFinder events. The horizontal line distinguishes between objects that we judge as likely to enter the final statistical sample and those that we do not. Note that among the latter, OGLE-2019-BLG-1470 is definitely planetary in nature, but it has a factor of \(\sim 3\) discrete degeneracy in its mass ratio, \(q\). On the other hand, KMT-2019-BLG-0414 has an alternate, orbiting binary-source (xallarap) solution that is disfavored by only \(\Delta \chi^2 = 4\), so it cannot be claimed as a planet.

Thus, of the 11 events (containing 12 planets) that are “above the line,” almost half are published here. This is the main accomplishment of the present work. Among these five planets, none is truly exceptional in its own right, although OGLE-2019-BLG-0679 has a relatively large normalized projected separation, \(s = 2.18\). Indeed, among the 53 previously published (or summarized) AnomalyFinder planets from 2018 and 2019 (Gould et al. 2022; Jung et al. 2022; Zang et al. 2022, 2023), only one had a larger separation, i.e., OGLE-2018-BLG-0383, with \(s = 2.45\) (Wang et al. 2022).

6.2. 2018+2019 Planets: Four Discrete Characterizations

Because the AnomalyFinder planets for the 2019 prime fields (Zang et al. 2022), as well as all of the 2018 fields (Gould et al. 2022; Jung et al. 2022), have previously been published (or summarized), our work permits several types of comparison between different seasons, classes of planets, and methods and conditions of discovery. At the highest level, we can compare the 2018 and 2019 seasons in terms of the number of planets found by field type (prime versus subprime), method of discovery (by eye versus AnomalyFinder), source trajectory (caustic crossing or not caustic crossing), and type of perturbation (major image, minor image, or central caustic). Table 17 presents these comparisons and summaries.
6.2.1. Statistical Consistency of 2018 and 2019

The first point is that the 2018 and 2019 seasons are consistent with respect to all of these breakdowns. For example, there were 33 and 25 total detections, respectively, i.e., a difference of $8 \pm \sqrt{58}$ according to Poisson statistics. Of all of the various comparisons that one could make among the various subcategories, the most “discrepant” is in the difference between the fraction of events identified by the AnomalyFinder, 52% versus 37%, i.e., a difference of $15\% \pm 13\%$, according to binomial statistics. Similarly, the fraction of planets found via major- versus minor-image perturbations is 55% versus 42%, i.e., a difference of 13%. Combining five tests, i.e., the Poisson test of total detections and the four binomial tests of Table 14, we find $\chi^2 = 3.37$ for 5 dof.

6.2.2. AnomalyFinder Yielded 40% of All Detections

Given that the two seasons are statistically consistent, we should ask what can be learned from their combined statistics. In particular, with 58 planets, this is a factor more than 2.5 times larger than any other homogeneously detected planetary microlensing sample (Suzuki et al. 2016). Perhaps the most striking feature of Table 17 is that 23 of the 58 planets (40%) were initially identified by AnomalyFinder, despite the fact that KMT’s publicly available data (the same as are input to AnomalyFinder) had previously been systematically searched by several experienced modelers. This may indicate the difficulty of by-eye searches in the era of massive microlensing data sets. It also shows that samples derived from by-eye searches alone are not even approximately complete.

At the same time, AnomalyFinder has not replaced by-eye searches; the two actually work hand in hand. AnomalyFinder typically identifies of order 250 candidates (after human review of a much larger candidate list) that each require detailed investigation to various levels. The first step in these massive reviews is to consult the summaries of systematic by-eye investigations, particularly those of C. Han, thereby reducing the number that require new or additional investigations by a factor of 3–5. The by-eye searches also serve as a check on the AnomalyFinder completeness. In fact, for 2021, we deliberately accelerated the by-eye searches with three new “mass production” papers (Ryu et al. 2022, 2023; Shin et al. 2023), as well as many other papers on individual planets (see Ryu et al. 2023 for a list), so that about 18 planets that are suitable for statistical studies were identified and prepared for publication prior to running the AnomalyFinder algorithm.

In contrast to the other three statistical indicators that are discussed below, the AnomalyFinder fraction of planets depends on a human factor. For example, when the 2016–2017 data are analyzed, the fraction could go down simply because there has been more time to apply the by-eye approach. And, going forward, the rate could go down because humans have learned more about planetary signatures based on the results from 2018–2019. On the other hand, the rate could go up if humans become less diligent, knowing that the planets will “eventually” be found anyway.

6.2.3. 50% ± 7% of Planets Have Caustic Crossings

Zhu et al. (2014) predicted that for about half of the planets detected in a KMTNet-like survey, the source would cross a caustic. These crossings are important because they allow the normalized source radius, $\rho$, to be measured, which in turn enables measurement of $\theta_E$ and $\mu_{\text{ep}}$. In addition to helping to characterize the planet, these measurements allow one to predict when the source and lens will be sufficiently separated to resolve them using AO on large telescopes, which can lead to measurements of the host and planet masses and the system distance. The 2018–2019 AnomalyFinder statistical sample confirms this prediction of Zhu et al. (2014) at relatively high statistical precision.

We note that a large minority of planetary events that do not have caustic crossings nevertheless yield good $\rho$ measurements because the planet is detected when the source passes over a magnification “ridge” that extends from the tip of a cusp; see, for example, OGLE-2016-BLG-1195 (Bond et al. 2017; Shvartzvald et al. 2017). Gould (2022) showed, based on a larger (but inhomogeneous) sample of 102 planetary events that substantially overlaps the current one, that about two-thirds yield $\rho$ measurements, even though only about half have caustic crossings.

Of the five planetary events analyzed in the present work, only one (KMT-2019-BLG-1216) has a caustic crossing. Yet, due to inadequate data over the caustic, $\rho$ is not well measured. This problem is likely to be much more common in subprime fields, particularly those that (like KMT-2019-BLG-1216) have cadences of $\Gamma = 0.4$ hr$^{-1}$. None of the four planetary events that lacked caustic crossings yielded precise $\rho$ measurements, although for OGLE-2019-BLG-0249, $\rho$ was reasonably well constrained.

6.2.4. 50% ± 7% of Major-/Minor-image Perturbations Are Major

It has long been known that for microlensing events with high or even moderate sensitivity to planets, the $(\log s, \log q)$ sensitivity diagrams are nearly symmetric about zero in $\log s$. One aspect of this symmetry is understood at a very deep level, while another aspect remains, to the best of our knowledge, completely unexplored.

Griest & Safizadeh (1998) showed that, for low $q$, there is a deep symmetry in the lens equation for $s \approx s^{-1}$ in the immediate neighborhood of the host (or, more accurately, the “center of magnification”). For example, the very first planet to exhibit such a degeneracy, OGLE-2005-BLG-071 ( Udalski et al. 2005), has a nearly identical $\chi^2$ for the two solutions (Dong et al. 2009). Hence, because the magnification pattern
is nearly identical for the two cases, a given source trajectory will generate very similar light curves and hence nearly equal detectabilities. As a result, all published sensitivity diagrams for high-magnification events (whose planet sensitivity is completely dominated by the source passage close to the center of magnification) are nearly perfectly symmetric; see, for example, OGLE-2007-BLG-050 (Batista et al. 2009) and OGLE-2008-BLG-279 (Yee et al. 2009).

By contrast, for source trajectories that pass closer to the planetary caustics than the central caustics, the magnification structures, and hence the resulting light-curve morphologies, are completely different. Major images generally have much larger caustics that are flanked by narrow magnification ridges, while minor images have smaller caustic pairs that are threaded by broad magnification troughs. Because of these two very different morphologies, one might expect the symmetry in the sensitivity profiles to break down.

In the very first systematic study of such sensitivity, Gaudi et al. (2002) presented \((\log s, \log q)\) plots for 43 microlensing events. Despite the fact that they span a very broad range of peak magnifications, many of these events display rough symmetry in their sensitivity profiles. However, in detail, many individual events also have an asymmetry in the minimum detectable \(\log q\) for \(\pm \log s\), with more sensitivity for \(\log s < 0\).

On the other hand, their Figure 13, which combines the sensitivities of these 43 events, shows a slight deviation from symmetry toward positive \(\log s\). However, they did not comment upon either effect.

Here we investigate detections in the 2018–2019 Anomaly-Finder sample from the standpoint of image perturbations rather than planet-host separation. As will become clear, these represent orthogonal perspectives. Figure 19 shows a scatter plot of \(\log q\) versus \(\log s^1\) for which positive and negative values correspond to major- and minor-image perturbations, respectively; see Equation (2).

There are three notable features. First, a majority (35/58) of the planets lie within \(|\log s^1| < 0.05\). In this regime, there is essentially no correlation between the signs of \(\log s^1\) and \(\log s\) because either light-curve morphology can be almost equally generated by positive or negative \(\log s\), including with large absolute values. Hence, there is no correspondence between the signs of \(\log s^1\) and \(\log s\) in this regime. By contrast, for \(|\log s^1| > 0.2\) (outer dashed lines), the anomalies are generally associated with the planetary caustics, so that \(\log s^1\) and \(\log s\) have the same sign. In this regime, there is suggestive evidence (six vs. two detections) that there is more overall sensitivity to wide-separation planets.

The second notable feature is that for \(|\log s^1| > 0.2\) (outer dashed lines), there are substantially more (six versus two) major-image than minor-image planets. In this regime, the anomalies are generally closely associated with the planetary caustics (and this is so for all eight cases from Figure 19). Moreover, both of the minor-image perturbations are caustic-crossing, whereas this is the case for only half of the major-image perturbations. Because of small number statistics, no strong conclusions can be drawn from either of these two comparisons. However, both conform to our naive impression that for planets that are far from the Einstein ring, it should be easier to detect the isolated bump due to a wide-separation planet than the weak dip of a close-separation planet, unless the source actually interacts with one of the two small caustics. Again, this issue can be more precisely explored from detailed simulations than from current planet samples due to small number statistics.

The third notable feature is that in the intermediate region, i.e., the transition between the central and planetary caustic regimes, there are about an equal number (seven versus eight) of major- and minor-image perturbations. This suggests that in this regime, the substantially different light-curve morphologies lead to about equal sensitivity. This is again deserving of systematic study via simulations. One might also note that all but one of the major-image perturbations in this regime are from 2018, while all but two of the minor-image perturbations are from 2019. However, as we cannot imagine any physical cause for this near dichotomy, we ascribe it to the random “noticeable effects” that one often discovers when viewing scatter plots.
The 6D scatter plot of 58 planets from 2018–2019. Abscissa: log mass ratio. Ordinate: source magnitude of unperturbed event at time of anomaly. The remaining four dimensions are shown in the legend and are the same as in Figure 14 of Jung et al. (2022).

6.2.5. 55% ± 7% of Detections Are from Prime Fields

The KMT devotes approximately half23 of its observing time to the six prime fields and the other half to the 21 subprime fields. While there are many considerations that go into this division, such as sensitivity to the Galactic distribution of planets, probing planets in a broad range of mass ratios, and probing other types of dim or dark objects like black holes, one consideration is certainly “return of planets on observing-time investment.” Prior to the start of KMT’s commissioning observations in 2015, Henderson et al. (2014) had already shown that there would be diminishing returns from concentrating all observations on the “most productive” fields. This understanding, as well as the experience of OGLE, which pioneered a multitiered observing approach, contributed to KMT adopting this strategy. Thus, it is of some interest that the planet return is in fact approximately proportional to the invested observing time.

6.3. 6D Distribution

In Figure 20, we show a six-dimensional (6D) representation of the 58 planets from 2018 and 2019 AnomalyFinder searches that are discussed in this section. It is an update to Figure 14 of Jung et al. (2022), which includes the subset of 33 planets from that paper. To recapitulate their description, it includes two continuous dimensions (given by the axes) and four discrete dimensions that are represented by colors and point types. The abscissa and ordinate are log \( q \) and \( I_{\text{S,anom}} = I_s - 2.5 \log[A(t_{\text{anom}})] \), with the latter being the source brightness in the unperturbed event at the time of the anomaly. The description of the symbols is identical to that of Jung et al. (2022), and they are also given in the legend.

Previously, Jung et al. (2022) had noted a “paucity of by-eye detections of non-caustic-crossing events (open bluish symbols) at low \( q \): i.e., 1 out of 5 for log \( q < -3 \) compared to 7 out of 12 for log \( q > -3 \).” This trend is strongly confirmed by the larger sample: 2 out of 12 for log \( q < -3 \) compared to 9 out of 16 for log \( q > -3 \). They also noted that 14 out of their 16 caustic-crossing planets were discovered by eye and that the remaining two were both in prime fields and at low log \( q < -3 \). They suggested that this was “a regime where machines may do better than people because the relatively weak signals of low-\( q \) events are spread out over a greater number of data points.” In the 2019 sample, almost equal numbers of caustic-crossing planets were found by each method, so that total is now 23 out of 29 caustic-crossing planets, i.e., still heavily favoring by-eye detections. Moreover, the other trend is strongly confirmed; now, out of six AnomalyFinder caustic-crossing planets, none are from sub-prime fields, and only one had log \( q > -3 \). This strengthens the evidence for the Jung et al. (2022) conjecture that machines excel in the high-cadence, low-\( q \) regime for caustic-crossing planets.

One feature of this diagram noted by Jung et al. (2022) that is not confirmed is the apparent threshold of AnomalyFinder detection at \( I_{\text{S,anom}} = 18.75 \). There had been only one major exception (OGLE-2018-BLG-0962), which has \( I_{\text{S,anom}} = 20.4 \). While there are still no detections fainter than this (previous) outlier, the 1.5 mag, 18.75 < \( I_{\text{S,anom}} \) ≤ 20.25, are now “filled in” with a total of 10 planets. Thus, \( I_{\text{S,anom}} \sim 20.25 \) now appears to be the detection floor. This will be tested as additional seasons are analyzed in this series.

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23 In KMT’s nominal schedule, exactly half of the time is devoted to prime fields. However, during 2016–2019, the schedule alternated between this nominal schedule and an alternate one, according to the need to support Spitzer microlensing (Yee et al. 2015). During these alternate times, KMT kept to the nominal schedule, while KMTS and KMTA devoted five-eighths of their time to the prime fields. Given the better weather at KMT in 2018, the overall fraction of time devoted to prime fields during these alternate times was about 57%.
have contributed to the research results reported within this paper.

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