THE SLOAN DIGITAL SKY SURVEY QUASAR LENS SEARCH. I. CANDIDATE SELECTION ALGORITHM

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ABSTRACT

We present an algorithm for selecting a uniform sample of gravitationally lensed quasar candidates from low-redshift (0.6 < z < 2.2) quasars brighter than i = 19.1 that have been spectroscopically identified in the Sloan Digital Sky Survey (SDSS). Our algorithm uses morphological and color selections that are intended to identify small- and large-separation lenses, respectively. Our selection algorithm only relies on parameters that the SDSS standard image processing pipeline generates, allowing easy and fast selection of lens candidates. The algorithm has been tested against simulated SDSS images, which adopt distributions of field and quasar parameters taken from the real SDSS data as input. Furthermore, we take differential reddening into account. We find that our selection algorithm is almost complete down to separations of 1″ and flux ratios of 10−0.5. The algorithm selects both double and quadruple lenses. At a separation of 2″, doubles and quads are selected with similar completeness, and above or below 2″ the selection of quads is better or worse, respectively, than for doubles. Our morphological selection identifies a nonnegligible fraction of single quasars; to remove these we fit images of candidates with a model of two point sources and reject those with unusually small image separations and/or large magnitude differences between the two point sources. We estimate the efficiency of our selection algorithm to be at least 8% at image separations smaller than 2″, comparable to that of radio surveys. The efficiency declines as the image separation increases because of larger contamination from stars. We also present the magnification factor of lensed images as a function of the image separation, which is needed for accurate computation of magnification bias.

Key words: gravitational lensing — quasars: general

Online material: color figures, machine-readable tables

1. INTRODUCTION

Gravitational lensing is a unique tool for exploring cosmology and the structures of astronomical objects. The well-understood physics of lensing makes it straightforward to use as a cosmological probe. Moreover, it is the only method that probes distributions of dark matter directly, since gravitational lensing is a purely gravitational phenomenon.

While detailed investigations of single intriguing lens systems are useful for cosmological and astrophysical applications, for some studies it is essential to do statistical analyses of a complete sample of lenses that are selected from a well-understood source population. For instance, strong lens probabilities are sensitive to the volume element of the universe, and thus the amplitude of the cosmological constant; thus, we can constrain cosmological models by comparing the observed lensed fraction in a catalog of distant sources to theoretical expectations (Turner 1990; Fukugita et al. 1990; Kochanek 1996; Chiba & Yoshii 1999). Strong lensing probabilities at the cluster mass scale probe the abundance and mass distribution of clusters (Narayan & White 1988; Maoz et al. 1997). The lens image separation distribution from galaxy to cluster mass scales can be used to study baryon cooling in dark halos (Kochanek & White 2001) and the connection between galaxy luminosities and masses of host halos (Oguri 2006).

A statistical sample is important not only for studies of lensing rates but also for interpretations of the results of individual lens modeling. It has been argued that lensing objects may be more or less atypical in their shapes, orientations (Hennawi et al. 2006a), and environments (Oguri et al. 2005a) because of strong dependences of lensing probabilities on these quantities. Therefore, we must take proper account of lensing biases to make a fair comparison between theory and observed results, which can only be accomplished for a well-defined statistical lens sample. A similar argument holds for constraints from stacked samples of strong lenses, including the structure and evolution of early-type galaxies (Treu & Koopmans 2004; Rusin & Kochanek 2005) and the fraction of substructures in lens galaxies (Dalal & Kochanek 2002). It is quite hard to estimate the degree of the lensing biases if one adopts a heterogeneous sample of lenses that have been discovered in a variety of ways.

The current largest complete sample of strong lenses was selected from a radio survey. The Cosmic Lens All-Sky Survey (CLASS; Myers et al. 2003; Browne et al. 2003) has discovered 22 gravitational lenses from ~16,000 radio sources. Thirteen of these lenses out of ~9000 flat-spectrum radio sources constitute a statistically well-defined sample. Recent progress on large-scale optical surveys has made it possible to construct comparable or even larger lens samples in the optical band. In particular, the Sloan Digital Sky Survey (SDSS; York et al. 2000) is expected to identify ~100,000 confirmed quasars (Schneider et al. 2005)
and ~500,000 photometrically selected quasars (Richards et al. 2004) when completed. In addition to the large number of sources, the optical lens survey has the advantage of a source population whose redshift and magnitude distributions are well understood, which is important when doing statistical analyses. On the other hand, optical lensed quasars are sometimes significantly affected by the light from foreground lens galaxies and dust reddening. Therefore, optical and radio lens surveys contain different systematics, and it is thus of great importance to compare the results from these two complementary lens samples.

We are searching for optical gravitationally lensed quasars from the imaging and spectroscopic data of the SDSS to construct a statistical sample of lensed quasars. Indeed, this project, the SDSS Quasar Lens Search (SQLS), has already discovered 14 new lensed quasars and recovered several (~7) previously known lensed quasars (Inada et al. 2003a, 2003b, 2003c, 2005, 2006; Johnston et al. 2003; Morgan et al. 2003; Pindor et al. 2004, 2006; Oguri et al. 2004b, 2005b; S. Burles et al. 2006, in preparation). Although these lenses do not constitute a statistically well-defined sample, these discoveries promise that we can create a large complete sample of lensed quasars. In order to construct a statistical sample, we need (1) to select lens candidates by a homogeneous method and (2) to understand the completeness of the method. In reality, we must also keep the efficiency reasonably high to assure the feasibility of follow-up observations. We can take advantage of the deep SDSS imaging data in five broad bands for the efficient identification of lensed quasar candidates. Pindor et al. (2003, hereafter P03) presented a selection method for lensed quasars from the SDSS and quantified its selection function. Their selection algorithm compares the \( \chi^2 \) value of a two-component point-spread function (PSF) model of spectroscopically confirmed SDSS quasars with that of a one-component PSF model. From simulated SDSS images, they found that the algorithm recovers quasar pairs of similar fluxes down to separations of ~0.7 arcsec.

In this paper we present another efficient algorithm to identify low-redshift (\( z \leq 2.2 \)) lensed quasar candidates from the spectroscopic sample of SDSS quasars. Our selection algorithm uses both morphological and color selection. The former aims to select quasars that are extended and are not well fitted by a PSF; it is designed to identify small-separation (~1") lens candidates. The latter algorithm simply compares colors of close pairs of objects with those of source quasars and is focused on lenses whose image separations are large enough to be deblended in the SDSS imaging data reduction pipeline. Our selection algorithm relies only on the outputs of the SDSS imaging pipeline and therefore does not require independent analysis of the imaging data themselves. This allows easy and fast identification of lensed quasar candidates. We apply our algorithm to realistically simulated SDSS images to quantify the completeness in detail. For the simulation of lenses, we follow the methodology developed by P03. We adopt observed distributions of sky levels and seeing sizes, as well as the observed SDSS source quasar population in a real SDSS quasar catalog, to estimate the completeness of our algorithm in a realistic manner. The algorithm will be applied to the SDSS data in N. Inada et al. (2006, in preparation, hereafter Paper II).

This paper is organized as follows. In § 2 we discuss the source quasar sample suitable for lens statistics. Our lens selection algorithms are presented in § 3 and are tested against simulations in § 4. Section 5 is devoted to studying an additional selection that is intended to remove any point-source-like candidates. The efficiency of our algorithms is briefly discussed in § 6, and we summarize our results in § 7.

2. SOURCE QUASAR SAMPLE

The SDSS is a survey to image \( 10^4 \) deg\(^2\) of the sky and to conduct spectroscopy of galaxies, quasars, and stars that are selected from the imaging data (Eisenstein et al. 2001; Strauss et al. 2002; Richards et al. 2002). The tiling algorithm (Blanton et al. 2003) assigns targets to fibers.\(^{10}\) A dedicated 2.5 m telescope (Gunn et al. 2006) at Apache Point Observatory is equipped with a multi-CCD camera (Gunn et al. 1998) with five optical broad bands centered at 3551, 4686, 6166, 7480, and 8932 Å (Fukugita et al. 1996; Stoughton et al. 2002). The imaging data are automatically reduced by the photometric pipeline PHOTO (Lupton et al. 2001). The astrometric positions are accurate to about 0.1" for sources brighter than \( r = 20.5 \) (Pier et al. 2003), and the photometric calibration errors are typically less than 0.03 mag (Hogg et al. 2001; Smith et al. 2002; Ivezić et al. 2004). All data are being released on a regular schedule (Stoughton et al. 2002; Abazajian et al. 2003, 2004, 2005; Adelman-McCarthy et al. 2006).

We construct a statistical sample of lensed quasars from the spectroscopic SDSS quasar catalog (Schneider et al. 2002, 2003, 2005). The quasar sample is quite heterogeneous in the sense that the quasars are selected with several different methods and the completeness is strongly dependent on the properties of the quasars (e.g., magnitudes, redshifts, and morphologies; see Richards et al. 2006b). We make a subsample of source quasars, which are suitable for lens statistics, by posing the following restrictions:

1. We consider only quasars in the redshift range \( 0.6 \leq z \leq 2.2 \). Lower redshift quasars are often extended in the SDSS images, which makes the morphological selection (§ 3.1) quite difficult; in any case, these quasars have low lensing probabilities. At low redshift most quasars are targeted by UV excess, which can select both point and extended quasars efficiently, while at higher redshifts outliers from the \( griz \) stellar locus are targeted as quasar candidates, for which we had to restrict our attention to point sources because of large contamination from galaxies. This is clearly shown in Figure 1, which plots the completeness obtained from the simulation by Richards et al. (2006b). The completenesses are different at \( z \geq 2.2 \) between point and extended quasars, which introduces a nonnegligible antilens bias. In particular, extended quasars at \( z \geq 3 \) are not selected by the quasar target selection algorithm, which significantly reduces the number of ~1" separation high-redshift lenses in the SDSS spectroscopic quasar catalog (see also the Appendix). In addition, at \( z > 2.2 \) it becomes more difficult to use colors to distinguish quasars from stars; thus, the efficiency of lens candidate selections is significantly degraded, particularly at the large-separation region where contamination from stars is quite large.

2. We adopt quasars with Galactic extinction-corrected (Schlegel et al. 1998) \( i \)-band magnitudes \( 15.0 \leq i_{\text{cor}} \leq 19.1 \). This is because at \( i_{\text{cor}} \leq 19.1 \) the low-redshift (\( z \leq 2.2 \)) quasar target selection is almost complete (Richards et al. 2002, 2006b), while fainter quasars are selected mainly with various “serendipity” algorithms (Stoughton et al. 2002), which are very incomplete and may introduce biases in estimating lensing rates. Moreover, the faint quasars have several disadvantages, including large photometric errors that reduce the efficiency of lens selection algorithms and contamination by lens galaxies, which become more important for less luminous quasars (e.g., Kochanek 1996). Fainter lenses are identified as lens candidates with lower efficiency because of their lower signal-to-noise ratios (S/Ns; see § 4.3).

\(^{10}\) Because of the physical size of fibers, no spectra of two targets that are closer than 55" are taken on a single plate. This “fiber collision” limitation makes it unlikely to obtain spectra of more than one component in a gravitational lens.
the same time have reasonable efficiency when applied to the SDSS data.

3.1. Morphological Selection

When the image separation between multiple images is small (\(\theta \lesssim 2.5\)), PHOT can not deblend the system into two components, and the objects are classified as single extended objects. Thus, small-separation lenses can be identified as lens candidates by searching for objects that are poorly fitted with the local PSF. To do so, we use the following two SDSS imaging parameters. One is \(\text{objc\_type}\), which describes the classification of the object as stars or galaxies, determined from the differences of magnitudes obtained by fitting PSF and galaxy profiles to the images in each band (e.g., Stoughton et al. 2002); \(\text{objc\_type} = 6\) indicates the object is a point source, while \(\text{objc\_type} = 3\) means the object is extended. The other parameter, \(\text{star\_L}\) (available in each band), is the probability that an object would have at least the measured value of \(\chi^2\) of a fit of the image to the PSF, if it really is well represented by a PSF. Put another way, the higher the value of \(\text{star\_L}\), the more likely it is fitted by a PSF.

The specific selection criteria of this morphology selection are as follows. First, even if a quasar is classified as a point source, we select it as a lens candidate if the PSF likelihood is small. Specifically, we select objects that satisfy all of the following four conditions:

\[
\begin{align*}
\text{M1: } & \quad \text{objc\_type} = 6, \\
& \quad \text{star\_L}(u) \leq 0.03, \quad \text{star\_L}(g) \leq 0.04, \\
& \quad \text{star\_L}(r) \leq 0.07 \text{ or } \text{star\_L}(i) \leq 0.07. 
\end{align*}
\]

We do not use the \(z\)-band \(\text{star\_L}\) parameter because of the relatively low S/N of \(z\)-band images. Although the S/Ns of \(u\)-band images are also low, we use the \(u\)-band parameter because quasars are UV-excess sources; thus, the \(u\) band is quite effective in discriminating quasar-quasar pairs from quasar-star pairs. However, we find that the condition on the \(u\)-band parameter is sometimes too strong, resulting in missing some lens candidates. As a separate cut, we adopt a relaxed range of \(u\)-band star likelihood but slightly tighter conditions on other parameters:

\[
\begin{align*}
\text{M2: } & \quad \text{objc\_type} = 6, \\
& \quad \text{star\_L}(u) \leq 0.06, \quad \text{star\_L}(g) \leq 0.04, \\
& \quad \text{star\_L}(r) \leq 0.04 \text{ or } \text{star\_L}(i) \leq 0.04. 
\end{align*}
\]

When a quasar image is classified as extended, we use a similar but relaxed criterion to identify a lens candidate; an object is selected when all of the following three conditions are met:

\[
\begin{align*}
\text{M3: } & \quad \text{objc\_type} = 3, \\
& \quad \text{star\_L}(u) \leq 0.45 \text{ or } \text{star\_L}(g) \leq 0.35, \\
& \quad \text{star\_L}(r) \leq 0.60 \text{ or } \text{star\_L}(i) \leq 0.60. 
\end{align*}
\]

We select quasars that satisfy one or more of the criteria M1–M3 as lens candidates. In practice, approximately half of small-separation gravitational lenses are classified as extended objects and selected by condition M3. Lenses that are classified as point sources tend to satisfy both conditions M1 and M2; the difference between M1 and M2 becomes important only for limiting cases in which lenses are close to the magnitude limit (\(i \sim 19.1\))

![Graph showing completeness of quasar spectroscopic target selection for point and extended quasars, computed from the simulation done by Richards et al. (2006b). The graph demonstrates the ratio of the two completenesses is shown in the bottom panel.](image-url)
and/or have very small image separations ($\theta \leq 1''$). These conditions are designed using both real SDSS data and simulations presented in § 4.

3.2. Color Selection

When the image separation is large enough for PHOTO to deblend two components, we can select lens candidates by comparing the colors between spectroscopically confirmed quasars and candidate companions. At $\theta > 7''$ the candidates’ companions must be point sources, but for smaller separations we also allow extended sources to be candidate companions, because for such small separations lensing galaxies may be superposed on the fainter components (e.g., Q0957+561). The selection method presented here is somewhat similar to those given by Oguri et al. (2004a) and Bennabi et al. (2006b), but we follow P03 to extend the methods to select differentially reddened lenses.

We conduct the color selection on the basis of the color difference:

$$D(i - j) = (i - j)_\text{quasar} - (i - j)_\text{companion} \quad \text{or} \quad (i - j)_\text{companion} - (i - j)_\text{quasar}, \quad (4)$$

where $(i, j) = (u, g), (g, r), (r, i), \text{ and } (i, z)$ and “quasar” and “companion” indicate the SDSS spectroscopic quasar in the source quasar sample and the candidate companion, respectively. We use PSF magnitudes and their errors throughout the paper. Although the definition of $D(i - j)$ contains a sign ambiguity, we regard candidate pairs as lens candidates if they pass the criteria below using either one of the above definitions. This becomes important when we consider differential reddening, because at first sight we do not know which object is reddened more. When the color difference comes from reddening, $D(i - j)$ for all sets of $(i, j)$ must have the same sign. On the other hand, the color difference of two unrelated objects could have both positive and negative signs for different $(i, j)$. This is the reason we do not define $D(i - j)$ as the absolute value of the color difference.

Since gravitational lensing does not change the color of objects, the color difference should be small if two components are lensed images of a single quasar. This criterion can be written as follows:

$$C1: |D(i - j)| < \Delta D(i - j), \quad (5)$$

which must be met for all four colors. The limit $\Delta D(i - j)$ describes an acceptable error of the color difference, for which we adopt the following fixed values

$$\Delta D(u - g) = 3 \times 0.20, \quad \Delta D(g - r) = 3 \times 0.09, \quad \Delta D(r - i) = 3 \times 0.09, \quad \Delta D(i - z) = 3 \times 0.30. \quad (6)$$

These values correspond to typical 3 $\sigma$ errors of the color differences of faint ($\sim$19–20 mag) objects in the SDSS images. We find that color differences contain very large errors when image separations are small because of the SDSS deblending algorithm. To account for this, we replace $\Delta D(i - j)$ with $3 \sigma_{D(i-j)}$ for $2'' \leq \theta < 3''$ and with $6 \sigma_{D(i-j)}$ for $\theta \leq 2''$, where $\sigma_{D(i-j)}$ is defined by

$$\sigma_{D(i-j)} = \sqrt{(\sigma^2_i + \sigma^2_j)_\text{quasar} + (\sigma^2_i + \sigma^2_j)_\text{companion}}, \quad (7)$$

where values of $\sigma_i$ are PSF magnitude errors estimated by PHOTO. Since it is quite rare that objects lie so close to the quasar, the large errors have little effect on the efficiency of our lens selection algorithm.

While gravitational lensing is independent of wavelength, significant color differences between lensed components may be caused by reddening. For instance, Falco et al. (1999) measured the distribution of differential reddening $\Delta E(B - V)$ from a number of gravitationally lensed quasar systems and found that it is well described by a Gaussian with a zero mean and standard deviation of 0.1 mag. We estimate how the differential reddening affects the color difference $D(i - j)$ as follows (see also P03): First, we use the composite quasar spectrum of Vanden Berk et al. (2001) as an input quasar spectral energy distribution. Next, we adopt the Galactic extinction law of Cardelli et al. (1989) and compute the color difference as a function of quasar and lens redshifts and $\Delta E(B - V)$, assuming $R_V = 3.1$. We show estimated reddening vectors for several sets of quasar and lens redshifts in Figure 2. As expected, the effect of the differential reddening is more significant for bluer bands. From this estimation, we prepare another color-selection algorithm in color-color space that is designed to select lens systems with large differential reddening:

$$C2: y > -\frac{\Delta y}{\Delta x} x - \Delta y,$$

$$y < \frac{y_2 - y_1 - \Delta y}{x_2 - x_1 + \Delta x} \left( x - x_1 \right) + y_1 + \Delta y,$$

$$y > \frac{y_2 + \Delta y}{x_2 + \Delta x} x - \Delta y, \quad y < \frac{y_1 + \Delta y}{x_1 + \Delta x} (x + \Delta x),$$

$$x < 0.2 + \Delta x \text{ if } x = D(u - g), \quad (8)$$

where the condition must be satisfied for all sets of $(x, y) = [D(u - g), D(g - r)], [D(g - r), D(r - i)], [D(r - i), D(i - z)]$, and

$$(x_1, y_1, x_2, y_2) = \begin{cases} (0.3, 0.4, 0.4, 0.4), & (x, y) = [D(u - g), D(g - r)], \\ (0.2, 0.2, 0.4, 0.1), & (x, y) = [D(g - r), D(r - i)], \\ (0.1, 0.2, 0.4, 0.1), & (x, y) = [D(r - i), D(i - z)]. \end{cases} \quad (9)$$

The color difference errors, $\Delta x$ and $\Delta y$, were defined in equation (6) and the text that follows it. Figure 2 shows the color selection regions defined by C1 and C2. In particular, Figure 2 indicates that condition C2 can identify lens systems with very large differential reddening, $\Delta E(B - V) \sim 0.3$. The cut at $D(u - g) = 0.2 + \Delta D(u - g)$ is added to reduce the contamination of quasar-pair systems. We select quasars that satisfy one or both criteria C1 or C2 as lens candidates.

4. TESTING THE SELECTION ALGORITHM

4.1. Simulation Method

To quantify the completeness of our selection algorithm, we perform a series of simulations that mimic SDSS observations. We follow the methodology developed by P03, but in this paper we implement multicolor (ugriz) simulations to test our selection based on both the morphology and color match.

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11 We note that our result is not strongly dependent on the assumed reddening law. For instance, we have repeated the analyses adopting SMC-like reddening, which Hopkins et al. (2004) claim fits the observed reddening of quasars, and have confirmed that the difference of completeness (§ 4.3) is indeed small.
First, we prepare object fields in which simulated lensed images are placed. We assign the characteristics of each field (sky level, seeing, and magnitude zero point) from a randomly selected field in the real SDSS data. Figure 3 shows the sky counts and seeing sizes (PSF_WIDTH) of 1000 random fields. Following the discussion in §2, we reject fields with PSF_WIDTH > 1.2 in the i band in our simulations.

For each field we also randomly choose a quasar from the subset of the DR3 quasar catalog with $0.6 \leq z \leq 2.2$ and $15.0 \leq i_{cor} \leq 19.1$ and adopt its redshift and magnitudes in the simulation. However, the redshift and magnitude distributions of the lensed population are quite different from those of the original unlensed population, because the lensing probability is a strong function of redshift and magnitude (e.g., Turner et al. 1984). We account for this difference by assigning to each object a weight proportional to its lensing probability, computed from a singular isothermal sphere model for the mass distribution of lensing galaxies and the velocity function of early-type galaxies derived from the SDSS (Sheth et al. 2003; Mitchell et al. 2005). The velocity function in comoving units is assumed to be constant with redshift. The magnification bias is computed by assuming the luminosity function of quasars obtained from the combination of the Two Degree Field and SDSS results (the “2SLAQ+Croom et al.” model; Richards et al. 2005). Although the lensing probability is sensitive to cosmological parameters, their main effect is to change the overall amplitude of the lensing probability and to have little effect on the dependence of lensing probability on the source redshift and magnitude, on which we are focusing. Figure 4 plots redshifts versus $i$-band magnitudes for randomly selected quasars with and without lensing probability weighting. Because of magnification bias, gravitational lensing preferentially chooses brighter and higher redshift quasars.

We convert magnitude to the number of counts with the following equations (Lupton et al. 1999):

$$ m = - \frac{2.5}{\ln(10)} \left[ \sinh \left( \frac{F/F_0}{2b} \right) + \ln(b) \right], \quad (10) $$

$$ \frac{F}{F_0} = \frac{\text{counts}}{\exp. \text{ time (s)}} 10^{0.4(Z+kX)}, \quad (11) $$

where $b$ is the softening parameter, $Z$ is the magnitude zero point, $k$ is the extinction coefficient, $X$ is air mass, and the exposure time is 53.91 s. Gravitationally lensed images are created in the simulated image by placing two point-source objects that
are separated by an angle $\theta$ and have a flux ratio $f_i$. For the PSF we use an analytic PSF composed of two Moffat functions with $\beta_1 = 7$ and $\beta_2 = 2$ and widths given for each field (Racine 1996; P03). The PSF magnitude of each system is normalized to that of the input quasar PSF magnitude in each band. With this procedure, we can simulate lensed quasars with realistic colors and field conditions.

We also include differential reddening in our simulation. The value of differential extinction $\Delta E(B - V)$ is randomly distributed as a Gaussian with a standard deviation of 0.1 mag (Falco et al. 1999). Again, the reddening vectors are computed from the composite quasar spectrum of Vanden Berk et al. (2001) and the extinction curve of Cardelli et al. (1989) assuming $R_V = 3.1$. The lens redshifts are also assigned randomly in the range $[0.2, z_{\text{max}}]$ with $z_{\text{max}} = \min(0.8, z_s - 0.2)$, where $z_s$ is the redshift of the source quasar. The distribution has the average redshift $z = 0.5$, which is consistent with the value $z \sim 0.51$, the mean of 30 existing gravitationally lensed quasar systems with known lens redshifts. We note that the flux ratio $f_i$ depends on the band when reddening is included; hereafter we refer to $f_i$ as the flux ratio of the $i$-band image.

We next pass these simulated fields through the PHOTO pipeline. Here we briefly summarize the procedure; for more details, see P03. We begin by preparing a PSF frame in which bright PSF stars (constructed from two Moffat functions) are displayed. This PSF field is passed through the fake stamp pipeline and postage stamp pipeline; this produces tables of the measured object parameters ($f_{\text{Obj}}$), as well as the cropped atlas images ($f_{\text{Atlas}}$). We do not need to pass the result through the photometric calibration pipeline, since we can convert which the result of the PSF fit is stored. The object field is passed through the fake stamp pipeline and frames pipeline; this produces tables of the measured object parameters ($f_{\text{Obj}}$), as well as the cropped atlas images ($f_{\text{Atlas}}$). We do not need to pass the result through the photometric calibration pipeline, since we can convert

\[\text{Fig. 3.—Width of PDFs vs. sky levels for 1000 randomly selected fields in the SDSS for each band. Open circles indicate those fields that are rejected because of poor seeing.}\]

\[\text{Fig. 4.—Redshifts vs.} \ i\text{-band magnitudes of randomly selected quasars at} \ 0.6 \leq z \leq 2.2 \text{and} \ 15.0 \leq i_{\text{Ox}} \leq 19.1. \text{Circles are selected from the quasar sample without any weight, and triangles use a lensing probability weight. [See the electronic edition of the Journal for a color version of this figure.]}\]
counts to magnitudes again by using equations (10) and (11). The magnitude error is estimated from that of the counts as

$$\Delta m = - \frac{2.5}{\ln(10)} \frac{\Delta \text{(counts)}}{\text{exp. time (s)}} \frac{1}{2b} \frac{10^{0.4(z+k)}}{\sqrt{1 + [(F/F_0)/2b]^2}}.$$  \hspace{1cm} (12)

The image parameters used to select the lens candidates are extracted from the fpObjc file. When PHOTO deblends two images, we regard the brighter image as targeted by the quasar spectroscopic target selection and use the image parameters of the brighter image.

We repeat the simulation for different sets of field parameters, redshifts and magnitudes of quasars, and reddening and compute the completeness of our selection algorithm as a function of the image separation $\theta$ and flux ratio $f_i$. Each simulation run consists of simulations of $25 \times 21$ quasars; the separation is changed from $0''$ to $4''8$ with a step of $0''2$ and the flux ratio from $10^{-1}$ to 1 with a step of a factor of $10^{0.05}$. Since the maximum image separation $\theta = 4''8$ is large enough compared with the seeing sizes of the SDSS images, we think the completenesses of larger separation lenses can be approximated by those of $\theta = 4''8$. We note that only a few known lensed quasars have flux ratios $f_i < 0.1$. We simulate 1000 runs to make an accurate estimation of the completeness.

We note that our simulations differ from those of P03 in several ways. In P03 the selection algorithm was tested for a few choices of seeing sizes and magnitudes of quasars, while in this paper we distributed them in a realistic way. We also simulated images in all five bands, while P03 did not consider differences between different bands.

### 4.2. Star Likelihood

Before showing the completeness, we check the distribution of $\star L$ to see the validity of using this parameter to identify lens candidates. All single quasars should be classified as point sources (objc_type = 6); thus, we can choose quite loose conditions for selecting lens candidates, without a significant decrease of efficiency, when they are classified as extended objects (objc_type = 3; see M3 in § 3.1). Therefore, in this section we concentrate our attention on lenses with objc_type = 6, for which we need to adopt a good discriminator between lenses and single quasars.

In Figure 5 we show the distributions of $\star L$ in $u, g, r,$ and $i$ for both lensed and single quasars from our simulations. For

![Figure 5](image-url)
lensed quasars we only consider the image separation range of \( \theta = 1'' - 1''4 \) and the flux ratio \( f_i > 10^{-0.5} \). It is clear that the distribution of \( \text{star}_L \) for lensed quasars is markedly different from that for single quasars. We find that lensed quasars have quite small star likelihoods \( \text{star}_L \sim 0 \), even when lenses are classified as point sources. On the other hand, the distribution of star likelihoods for single quasars peaks at \( \text{star}_L \sim 1 \) and rapidly decreases as \( \text{star}_L \) decreases. This figure assures that conditions M1 and M2 presented in § 3.1 serve as an excellent method to distinguish gravitational lenses from normal quasars.

4.3. Completeness

We show the completeness in the \( \theta - f_i \) plane in Figure 6. As expected, lenses with larger image separations and flux ratios closer to unity are more likely to be selected. In particular, our selection algorithm is \( \geq 95\% \) complete at \( \theta \geq 1'' \) and \( f_i \geq 10^{-0.5} \) (Fig. 6, dotted lines). Therefore, we propose to construct a statistical sample within this region of parameter space.

Figure 7 shows the completenesses for three different ranges of redshifts and magnitudes. We find that the completeness is almost independent of the quasar redshift. On the other hand, the completeness shows a weak but significant trend with magnitude, in the sense that brighter quasars are more complete. In particular, the completeness at small \( f_i \) is larger for brighter quasars. This is reasonable, since companion pairs, which are often hidden when \( f_i \) is small, are more easily detected for brighter quasars. However, even for the faintest quasars \( (i > 18.7) \), the completeness at \( \theta \geq 1'' \) and \( f_i \geq 10^{-0.5} \) is very high. Therefore, we conclude that our selection does not affect the distributions of redshifts and magnitudes of lensed quasars very much.

Figure 6 does not make it clear whether lenses are mainly chosen by the morphological or color selection at each point in the \( \theta - f_i \) plane. Figure 8 shows the completeness from the two selection algorithms separately. We also show the overlap of these two selections. At \( \theta \leq 2'' \) lenses are mostly selected morphologically, and at \( \theta \geq 2'' \) the color selection dominates. The overlap of these two selections is quite small. The completenesses from the two selection algorithms, as well as the total completeness, are listed in Table 1.

4.4. Magnification Factor

The magnification bias is one of the most important elements in predicting lensing probabilities. The SDSS quasar target selection (Richards et al. 2002) uses PSF magnitudes for the magnitude limit; therefore, we examine those of simulated lensed quasars with various image separations. Basically, when the image separation is quite large and two images are deblended successfully by PHOTO, the PSF magnitude of the targeted quasar should be roughly equal to that of the brighter image. However, when the image separation is too small to be distinguished from a point source, the PSF magnitude is that of the sum of the two images. Bearing this in mind, we define the following parameter:

\[
\tilde{\mu} \equiv \frac{\mu - \mu_{\text{bri}}}{\mu_{\text{fai}}},
\]

where \( \mu_{\text{bri}} \) and \( \mu_{\text{fai}} \) are the magnification factors for the brighter and fainter images, respectively. The magnification factor \( \mu \) indicates that measured in the SDSS (i.e., the PSF magnitude of the lens system divided by the original magnitude of the source quasar). The definition indicates that \( \mu = \mu_{\text{tot}} = \mu_{\text{bri}} + \mu_{\text{fai}} \) and \( \mu = \mu_{\text{bri}} \) correspond to \( \tilde{\mu} = 1 \) and 0, respectively. Therefore, the parameter should approach zero at large separations, and \( \tilde{\mu} \rightarrow 1 \) for \( \theta \rightarrow 0 \).

We derive \( \tilde{\mu} \) as a function of the image separation from our simulation runs. We only consider simulated lens systems with flux ratio \( f_i \geq 10^{-0.5} \), because the derived \( \tilde{\mu} \) contains large errors when flux ratios are very large, \( \mu_{\text{bri}} \gg \mu_{\text{fai}} \), and also because the completeness is low for large flux ratio systems. Moreover, \( \mu_{\text{tot}} \sim \mu_{\text{bri}} \) at \( f_i \ll 1 \) implies that the magnification factors do not change as a function of the image separation in any case. Figure 9 plots the result. As expected, \( \tilde{\mu} \) changes from \( \tilde{\mu} \sim 1 \) to \( \sim 0 \) with increasing image separations. We find that the curve is fitted roughly by the form

\[
\tilde{\mu} = \frac{1}{2} \left[ 1 + \tanh \left( 1.76 - 1.78\theta \right) \right],
\]

where the image separation \( \theta \) is in units of arcseconds.

We note that for a given image separation the magnification bias can be written as

\[
B = \tilde{\mu}B_{\text{tot}} + (1 - \tilde{\mu})B_{\text{bri}},
\]

where \( B_{\text{tot}} \) and \( B_{\text{bri}} \) are the magnification bias from the total magnification and magnification of the brighter image, respectively.

4.5. Quadruple Lens Systems

Thus far we have concentrated our attention on two-component lenses, but in reality a non-negligible fraction of gravitationally lensed quasars have four components. In this subsection we consider our selection algorithm’s effectiveness for such quadruple lenses.

The difficulty in studying the selection of quadruple lenses lies in their complexity; while double lenses are parameterized by flux ratio and image separation, we require eight parameters to characterize a quadruple lens, making a full exploration almost impossible. Instead, in this paper we adopt the image configurations and flux ratios from 20 known quadruple lenses taken
Fig. 7.—Completeness of our lens selection algorithm for different redshift and magnitude bins. The lines are the same as in Fig. 6. Top panels: Completenesses for restricted redshift ranges of $z < 1.5$ (left), $1.5 < z < 1.8$ (middle), and $1.8 < z$ (right). Bottom panels: Completenesses for restricted $i$-band magnitude ranges of $i < 18.2$ (left), $18.2 < i < 18.7$ (middle), and $18.7 < i$ (right). In both cases, the sizes of the bins are determined such that the (lensing probability weighted) number of quasars in each bin becomes roughly equal. [See the electronic edition of the Journal for a color version of this figure.]

Fig. 8.—Left: Completeness of the morphological selection only. Middle: Completeness of the color selection only. Right: Fraction of lenses selected from both morphological and color selections. The lines are the same as in Fig. 6. [See the electronic edition of the Journal for a color version of this figure.]
from the CASTLES Web site (see Table 2). We apply a similarity transformation to the quadruple lenses in order to derive the completeness as a function of the image separation. We define the image separation to be the maximum distance between multiple images. The differential reddening is included, again using the distribution of Falco et al. (1999). As in the case of double lenses, we simulate 1000 runs to make an accurate estimate of the completeness.

Figure 10 plots the completeness of each of the 20 quadruple lens configurations as a function of the image separation, as well as the average of them. We find that our algorithm can select quadruple lenses reasonably well: even in the limiting case, as the average of them. We find that our algorithm can select quadruple lenses as a function of the image separation, as well as the completeness as a function of the image separation. We define the image separation to be the maximum distance between multiple images. The differential reddening is included, again using the distribution of Falco et al. (1999). As in the case of double lenses, we simulate 1000 runs to make an accurate estimate of the completeness.

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Figure 11 shows the completeness for morphological and color selections separately. Unlike double lenses, quadruple lenses are small, but they might become important when discussing the relative number of quadruple and double lenses (Keeton et al. 1997; Rusin & Tegmark 2001; Chae 2003; Oguri & Keeton 2004).
are effectively chosen by the morphological selection even at large image separations, $\theta > 3\arcsec$. In addition, it is quite common that lenses are selected by both criteria. This is because quadruple lenses are more complex than double lenses. In particular, the complexity and larger number of images for quadruple lenses make the completeness at $\theta > 2\arcsec$ quite large. Again, the completenesses from the two selection algorithms, as well as the total completeness, are listed in Table 3.

We also study the completeness for different image configuration types: fold (two of the images lying close together), cross (almost symmetric), and cusp (three of the images lying close together). To do so, we divide the 20 quadruple lenses into these three categories by visual inspection and compute the average completeness for each category. The categorization is summarized in Table 2, and the results are shown in Figure 12. Our algorithm selects cross lenses very well but is less effective for cusp lenses; this is because the fourth image of a cusp lens, which lies far from the critical curve, is usually much fainter than the other images near the critical curve, and therefore the effective “flux ratio” of cusp lenses becomes small. To pursue this further, we consider the flux ratio between the brightest image and the image farthest from the brightest image (denoted by $f_{\text{quad}}$) and see its correlation with the completeness. Figure 13 plots $f_{\text{quad}}$ and the completeness at $\theta = 1\arcsec$ for all 20 quadruple lenses listed in Table 2. As expected, we see a clear correlation: all quadruple lenses with completeness $<0.8$ have $f_{\text{quad}} < 0.4$. Therefore, despite the complexity and large number of parameters involved in quadruple lenses, $f_{\text{quad}}$ gives a rough criterion of how well our selection algorithm can identify quadruple lenses.

5. REMOVING SINGLE QUASARS

The nonzero completeness at $\theta \rightarrow 0$ implies that our morphological selection algorithm falsely identifies “normal” quasars as lenses. Indeed, when we apply our algorithm to the DR3 quasar catalog, we find many lens candidates, many of which appear to be point sources (see §6), although most normal single quasars are removed by our morphological selection. Therefore, in this section we present an additional selection that aims to exclude such point sources.

![Fig. 11.—Completeness for quadruple lenses, with the morphological selection only (left), the color selection only (center), and both morphological and color selections (right). The lines are the same as in Fig. 10. [See the electronic edition of the Journal for a color version of this figure.]](image1)

**TABLE 3**

| NAME    | $\theta$ (arcsec) | Completeness |
|---------|-------------------|--------------|
|         | Total | Morph. | Color | Both |
| PG 1115+080 | 0.0   | 0.035  | 0.035 | 0.000 | 0.000 |
| PG 1115+080 | 0.2   | 0.030  | 0.030 | 0.000 | 0.000 |
| PG 1115+080 | 0.4   | 0.029  | 0.029 | 0.000 | 0.000 |
| PG 1115+080 | 0.6   | 0.059  | 0.059 | 0.000 | 0.000 |
| PG 1115+080 | 0.8   | 0.188  | 0.188 | 0.000 | 0.000 |
| PG 1115+080 | 1.0   | 0.490  | 0.490 | 0.000 | 0.000 |
| PG 1115+080 | 1.2   | 0.796  | 0.796 | 0.000 | 0.000 |
| PG 1115+080 | 1.4   | 0.937  | 0.937 | 0.000 | 0.000 |
| PG 1115+080 | 1.6   | 0.978  | 0.978 | 0.001 | 0.001 |
| PG 1115+080 | 1.8   | 0.987  | 0.987 | 0.007 | 0.007 |
| PG 1115+080 | 2.0   | 0.992  | 0.991 | 0.016 | 0.015 |

Note.—Table 3 is published in its entirety in the electronic edition of the Astronomical Journal. A portion is shown here for guidance regarding its form and content.

![Fig. 12.—Completeness for quadruple lenses of different image configuration types: fold (solid line), cross (dotted line), and cusp (dashed line). [See the electronic edition of the Journal for a color version of this figure.]](image2)
To do so, we fit systems selected by our morphological selection algorithm with two PSFs using the GALFIT package (Peng et al. 2002) and derive the image separation and flux ratio. We use stars in the same field as a PSF template in fitting. The fitting of any single object with two PSFs should result in either too-large magnitude differences or too-small image separations. Here we only consider gravitational lenses included in the proposed statistical sample, i.e., those with \( \theta > 1'' \) and \( f_i > 10^{-0.5} \). The point sources should have significantly smaller image separation and/or larger flux ratios than the above limit. Therefore, we can reject such point sources by making a cut in image separation–flux ratio space. We do this fit on both the \( u \)- and \( i \)-band images. We use the \( i \)-band image because the flux ratio is defined by that of the \( i \)-band image. The \( u \) band is used because it is a powerful discriminator between lensed pairs and superpositions of quasars and stars or quasars and galaxies.

To determine the region of parameter space we should exclude, we perform fitting of simulated images (see § 4.1). We consider only the extreme image separation, \( \theta = 1'' \), for which fitting by two PSFs will be most difficult. The flux ratio varies over the range \( f_i = [10^{-0.5}, 1] \). We show the results for both \( u \)- and \( i \)-band simulated images in Figures 14 (distribution in image separation–flux ratio space for each band) and 15 (distribution in \( u \)- and \( i \)-band image separation space). It is seen that quadruple lenses show smaller image separations, on average, than double lenses; this is clearly because GALFIT tends to fit the two brightest images, which are not necessarily the two most separated from each other. From this result, we determine the selection criteria to be the following:

\[
\begin{align*}
\text{S1: } \theta(u) &\geq 0.4'', \\
&\theta(i) \geq 0.4'', \quad \Delta m(u) \leq 2.5, \\
&\Delta m(i) \leq 2.2, \quad |\theta(u) - \theta(i)| \leq 1'',
\end{align*}
\]

where \( \theta \) and \( \Delta m \) are the separation and magnitude difference between two fitted PSF components. We select candidates that meet all of the above criteria. In our simulation, more than 99% of double and quadruple lenses with \( \theta = 1'' \) pass these conditions; thus, we conclude that this additional selection hardly changes the high completeness of our statistical lens sample estimated in § 4. Condition S1 rejects \( \sim 85\% \) of the candidates selected by the morphological selection (see § 6).

6. Efficiency

A definitive estimate of the efficiency requires follow-up observations of all candidates, which is still in progress. Nevertheless, in this section we discuss the efficiency of our selection algorithm in light of the follow-up observations that we have performed thus far.

To do so, we again adopt the DR3 quasar sample (Schneider et al. 2005). As discussed in § 2, we use 22,682 quasars at \( 0.6 \leq z \leq 2.2 \) and \( 15.0 \leq i_{\text{cor}} \leq 19.1 \) as our source quasar sample. We apply our algorithms to the quasar sample and select 648...
candidates from the morphological selection and 228 candidates from the color selection. By applying the additional cut described in $\S$ 5 to the morphologically selected candidates, the number of candidates from the morphological selection reduces to 97, only $\sim$15% of the original. For color selection, we only select candidates with image separations smaller than 20$''$ and magnitude differences smaller than 1.3, since we concentrate our attention on the lens sample in the range 1$''$ $\leq \theta \leq$ 20$''$ and $I_l \geq 10^{-0.5}$. It is expected that the efficiency is a strong function of the image separation; thus, we consider three image separation ranges, $\theta < 2''$, 2$'' \leq \theta < 7''$, and 7$'' \leq \theta$, and derive the efficiency for each image separation range separately. The image separations of the morphologically selected candidates we used are those obtained by fitting i-band SDSS images with two PSFs using GALFIT (see $\S$ 5). The result is summarized in Table 4. We note that the algorithm successfully identified all known lenses included in the subset of the DR3 quasar catalog, thereby confirming the high completeness shown in $\S$ 4. Since we still have candidates that are to be observed, the efficiency shown in Table 4 should, in principle, be regarded as the lower limit. However, we believe that this is close to the true value because of the “observation bias,” whereby we preferentially observe better candidates. As expected, the efficiency is dependent on the image separation: the efficiencies are 8%, 3%, and 0.6% for $\theta < 2''$, 2$'' \leq \theta < 7''$, and 7$'' \leq \theta$, respectively. Most false positives, which have been revealed by follow-up observations, are quasar-star pairs or quasar pairs (see also Hennawi et al. 2006b). For the smallest image separation bin, the efficiency is comparable to that of the CLASS, which discovered 13 gravitational lenses out of 149 candidates (Browne et al. 2003). We note that the candidates can be restricted further by comparing with other existing observations (see Paper II), such as Faint Images of the Radio Sky at Twenty cm (Becker et al. 1995), which should increase the efficiency.

7. CONCLUSION

We have presented the SQLS lens candidate selection algorithm from the SDSS spectroscopic quasar catalog. Our algorithm consists of two selection methods, a morphological selection that identifies “extended” quasars and a color selection that picks up adjacent objects with similar colors. Both selection methods rely only on imaging parameters that the SDSS reduction pipeline generates, thereby allowing easy and fast selection of lens candidates. To reduce the number of candidates further, we fit the images of morphologically selected candidates with two PSFs, deriving the image separations and magnitude differences and applying cuts in the image separation–magnitude difference space. This allows us to remove most single quasars that are otherwise selected by the morphological selection.

We have performed simulations of lensed quasar images in the SDSS to determine the effectiveness of the selection algorithm. In the simulations, distributions of field parameters (i.e., seeings and sky levels), as well as those of source quasars, are taken from the real SDSS data. We have also taken differential reddening by lens galaxies into account. The simulated SDSS images are then passed through the imaging data processing pipeline used in the SDSS. This allows a reliable estimate of the completeness of the algorithm to be made. We have found that our selection algorithm is almost complete at image separations larger than 1$''$ and flux ratios larger than $10^{-0.5}$. We have also quantified the magnification factor of lensed images as a function of the image separation, which is important for accurate computation of magnification bias. The algorithm successfully identifies both double and quadruple lenses, although there is a systematic difference of completenesses between double and quadruple lenses. The efficiency depends strongly on the image separation; at separations less than 2$''$ the preliminary efficiency of our selection algorithm is about 8%, comparable to that of the CLASS.

An important effect we have neglected in our simulation is fluxes from lens galaxies. Although for quasars with $r \lesssim 19.1$ and redshifts above 0.6 lens galaxies tend to be fainter than lensed quasars, they could affect quantitative results. An important point is that lensed quasars are not selected by the quasar spectroscopic target selection if the fluxes of lens galaxies dominate. A good example is the high-redshift lens SDSS J0903+5028 (Johnston et al. 2003), which was targeted for SDSS spectroscopy as a luminous red galaxy, rather than a quasar, because the lensing galaxy is brighter than the lensed quasar images. Therefore, such systems will not be included in SDSS spectroscopic quasar catalogs. This means that we need to simulate quasar spectroscopic target selection for lensed systems, as well as lens selection algorithms, to address the effect of lens galaxy fluxes. This is beyond the scope of this paper.

The well-understood selection function, together with the knowledge of redshift and magnitude distributions of source quasars, is very important to control possible systematic effects. We plan to construct a first SQLS complete lens sample from the DR3 quasar catalog, which will be presented in Paper II.

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The SDSS is managed by the Astrophysical Research Consortium for the Participating Institutions. The Participating Institutions are the American Museum of Natural History, the Astrophysical Institute Potsdam, the University of Basel, Cambridge University, Case Western Reserve University, the University of Chicago, Drexel University, Fermilab, the Institute for Advanced Study, the Japan Participation Group, Johns Hopkins University, the Joint Institute for Nuclear Astrophysics, the Kavli Institute for...
Gravitationally lensed quasars are sometimes classified as extended objects (objc_type = 3) by PHOTO. This is very important for lensing of high-redshift (z > 3) quasars because extended high-redshift quasars are not targeted by the quasar target selection pipeline (see Fig. 1). Therefore, we need to take this bias into account when we constrain the shape of the high-redshift quasar luminosity function from the absence of strong gravitational lensing (e.g., Richards et al. 2006a). In addition, this bias strongly affects the search for lenses using a photometrically selected quasar sample (Richards et al. 2004), because extended quasars are not included in the photometric sample, again in order to enhance the efficiency. In this appendix we derive the fraction of lens systems that are classified as extended objects from our simulations of lensed quasar images (x4). Although we focused our attention on bright (i < 19:1) low-redshift quasars, and therefore the result may not apply to high-redshift quasars that are fainter and have quite different colors, the result is still useful to understand how PHOTO classifies lensed quasar images.

Figure 16 shows the fraction of double lenses that are classified as extended objects in the $\theta$ vs $\theta$ plane. As expected, the fraction is a strong function of the image separation: the lenses with $\theta > 1''$ are most likely judged as extended objects. The fraction depends on the flux ratio; lenses with larger $f_i$ are more likely to be classified as extended objects. In Figure 17 we plot the same fraction for quadruple lenses. Although the system-by-system difference is large, at $\theta > 1''$ quadruple lenses are more likely to be classified as extended objects, on average, than double lenses, which is clearly because of the existence of close fold-cusp image pairs.

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