

Identification of FeLoBAL Quasars in SDSS DR7Q Using Convolutional Neural Networks (Postprint)

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Abstract

Iron low-ionization broad absorption line quasars (Fe Low-ionization Broad Absorption Line Quasar, FeLoBALQ) represent one of the rarest subclasses of quasars. The radiation from these objects violently expels surrounding material, forming powerful high-velocity outflows in which low-ionization species, notably iron, absorb the quasar radiation and produce characteristic low-ionization iron broad absorption line spectra. The energy carried by FeLoBALQ outflows is sufficient to explain the $M - \sigma$ relationship between supermassive black hole mass M and host galaxy bulge velocity dispersion σ , while some studies indicate that FeLoBALQs may be associated with starburst galaxies or major mergers. However, the limited number of FeLoBALQs discovered to date has hindered statistical verification of these theories. This research program conducts a large-scale search within existing large quasar samples to identify FeLoBALQs among previously discovered quasars, thereby providing a foundational sample for further investigation. Employing the convolutional neural network (Convolutional Neural Network, CNN) method from deep learning and utilizing previously discovered FeLoBALQ spectra as training samples, we have identified 160 new FeLoBALQ spectra from a total of 50,931 quasar spectra in the SDSS (Sloan Digital Sky Survey) DR7Q (Data Release 7 Quasar catalog) within the redshift range $0.8 < z < 2.125$. The study reveals that FeLoBALQs are redder than typical quasars, with previously discovered FeLoBALQs being slightly redder than the newly identified ones; these differences are more pronounced at the blue end and nearly vanish in the mid-infrared regime. Combining previously discovered FeLoBALQs with our new identifications, we estimate that the fraction of FeLoBALQs among the total quasar population within this redshift range of the sample is approximately 0.43%, a value slightly higher than previous estimates but potentially still underestimated. Future work aims to extend this method to larger samples such as SDSS DR16Q.

(Data Release 16 Quasar catalog) to discover additional FeLoBALQs and to employ large samples to investigate the relationship between FeLoBALQs and host galaxy star formation, major galaxy mergers, and the co-evolution of galaxies and their central supermassive black holes.

Full Text

Preamble

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Identifying FeLoBAL Quasars in SDSS DR7Q Using the Convolutional Neural Network Method

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Abstract

Iron low-ionization broad absorption line quasars (FeLoBALQs) represent one of the rarest subclasses of quasars. The radiation from these objects violently expels surrounding material, forming powerful high-velocity outflows. Low-ionization species, particularly iron, absorb the quasar radiation and produce characteristic broad absorption features from low-ionization iron. The outflows in FeLoBALQs carry sufficient energy to explain the M- relation between supermassive black hole mass and host galaxy bulge velocity dispersion, and studies suggest FeLoBALQs may be associated with starburst galaxies or major galaxy mergers. However, the limited number of FeLoBALQs discovered to date has hindered statistical verification of these theories. This research aims to conduct a large-scale search for FeLoBALQs within existing quasar samples to provide a foundation for further study. Using the convolutional neural network (CNN) method from deep learning, we identified 160 FeLoBALQ spectra from a total of 50,931 quasar spectra in the SDSS (Sloan Digital Sky Survey) DR7Q (Data Release 7 Quasar catalog) within the redshift range $M(\text{cid:27})(\text{cid:3})M(\text{cid:0})(\text{cid:27})(\text{cid:3})0.8 < z < 2.1250:43\%$. We find that FeLoBALQs are redder in color than typical quasars, with previously discovered FeLoBALQs appearing slightly redder than the newly identified ones.

These differences are more pronounced at bluer wavelengths and nearly disappear in the mid-infrared. Combining our new discoveries with previously known FeLoBALQs, we estimate that FeLoBALQs constitute approximately $M(\text{cid:27})(\text{cid:3})M(\text{cid:0})(\text{cid:27})(\text{cid:3})0.8 < z < 2.1250:43\%$ of all quasars in this redshift range, a fraction slightly higher than previous estimates but likely still underestimated. Future work will extend this method to larger samples such as SDSS DR16Q (Data Release 16 Quasar catalog) to discover more FeLoBALQs and enable large-sample studies of their relationship with host galaxy star formation, major mergers, and the co-evolution of galaxies and central supermassive black holes.

Keywords: galaxies: active, quasars: absorption lines, methods: deep learning, catalogs

PACS codes: P158; **Document code:** A

1 Introduction

Among broad absorption line (BAL) quasars, iron low-ionization broad absorption line quasars (FeLoBALQs) represent an extremely special subclass. The primary observational characteristic of these quasars is the presence of low-ionization iron absorption lines, dominated by Fe II with some objects showing Fe III, occasionally accompanied by strong absorption features from He I, Si II, and Ni II. The ionization potential for iron is lower than that for C IV (which characterizes high-ionization BALs or HiBALs) and Mg II (which characterizes low-ionization BALs or LoBALs). This indicates that the outflowing gas in FeLoBALQs originates at greater distances from the central region than in typical LoBAL quasars, suggesting more extreme outflow phenomena. Research indicates that establishing the famous M -relation in the co-evolution of galaxies and central black holes requires approximately 5% of the quasar's bolometric luminosity [?, ?], while Moe et al. [?] demonstrated that the kinematic luminosity of FeLoBALQ outflows reaches 1% of the bolometric luminosity, making it a significant component of the energy driving co-evolution. Faucher-Giguère et al. [?] showed that for the most luminous FeLoBALQs, this fraction can reach $M(\text{cid:27})(\text{cid:3})M(\text{cid:0})(\text{cid:27})(\text{cid:3})2\%(\text{cid:0})5\%18\%(\text{cid:12})$, potentially sufficient to entirely drive co-evolution. These studies suggest that quasars in the FeLoBALQ state may play a crucial role in establishing the relationship between central black holes and their host galaxies.

Since their discovery, the physical mechanisms and origins of FeLoBALQs have remained pressing questions. However, the rarity of these objects has limited systematic studies and prevented statistically robust conclusions. First, no consensus has been reached regarding the mechanism of FeLoBAL formation. Yuan et al. [?] and Runnoe et al. [?] found that BAL quasars exhibit weaker [O III] emission lines and stronger Fe II emission, implying higher accretion rates than non-BAL quasars. In contrast, Schulze et al. [?] studied LoBAL quasars and FeLoBALQs, and Rankine et al. [?] studied BAL quasars, concluding that the accretion rates of BALs and their subclasses (including LoBALs and FeLoB-

ALs) are similar to those of non-BAL quasars, contradicting earlier studies. Additionally, no unified model for FeLoBALQ formation has emerged. Farrah et al. [?] proposed in 2010 that FeLoBALQs may exist in environments satisfying three conditions: (1) the galaxy is in the late stage of a starburst driven by a major merger, (2) a high-luminosity quasar is in the late stage of accreting surrounding dust, and (3) we are observing from a specific viewing angle. This scenario has received support from several studies: Farrah et al. [?] discovered four quasars at the centers of ultraluminous infrared galaxies (ULIRGs) in 2005, one of which was a FeLoBALQ. The extreme infrared luminosity of ULIRGs indicates abundant warm gas and dust and extremely high star formation rates. Farrah et al. [?] conducted infrared observations of nine FeLoBALQs in 2007, finding their infrared luminosities consistent with ULIRGs and optical features indicating star formation rates of several hundred solar masses per year. This study suggested an inevitable connection between FeLoBALQs and the extreme star formation rates of their host galaxies, likely indicating co-evolution between the quasar and host. Lawther et al. [?] performed high-resolution observations of FeLoBALQs, confirming evidence of past major mergers in their environments. However, some studies have challenged this scenario: Violino et al. [?] conducted submillimeter observations of 17 FeLoBALQs and found no correlation between FeLoBALQs and star formation in host galaxies. Villforth et al. [?] performed high-resolution observations but found no evidence of major mergers in the environments of FeLoBALQs.

The primary reason for these inconsistent conclusions is the small number of FeLoBALQs discovered to date, making studies of limited samples susceptible to significant statistical uncertainties. Research on FeLoBALQs requires larger samples to yield statistically convincing results. The series of studies by Leighly and Choi et al. [?] on the spectral properties of FeLoBALQs, using a sample of 50 objects, represents the largest and most systematic recent work on FeLoBALQs, yielding consistent conclusions: using absorption features to study outflows, they found a positive correlation between outflow strength and velocity, with FeLoBALQ outflow strength sufficient to provide quasar feedback to host galaxies [?]. Their study of emission line strengths revealed that H lines in FeLoBALQs are generally broader, suggesting either high inclination angles or a lack of low-velocity emitting gas. After calculating accretion rates from emission strengths, they classified FeLoBALQs into low-luminosity, low-accretion-rate and high-luminosity, high-accretion-rate types [?]. Using models to study the locations of broad absorption features in both types, they found that for high-luminosity FeLoBALQs, the distance of broad absorption formation from the quasar center correlates negatively with accretion rate, while the opposite holds for low-luminosity types [?]. Comparing spectral energy distributions (SEDs) of FeLoBALQs with those of typical quasars, they found significant differences in emission lines and other parameters between FeLoBALQs, LoBAL quasars, and normal quasars, suggesting FeLoBALQs may be a particularly special class of quasars [?].

These series of studies, supported by a moderately sized FeLoBALQ sample,

have yielded results with some degree of universality. However, to verify these theories and support more systematic investigations of FeLoBALQ properties, larger samples are needed. The purpose of this study is to conduct a large-scale search and identification of FeLoBALQs, effectively expanding the FeLoBALQ sample set to provide crucial data support for systematic studies of FeLoBALQs and necessary statistical foundations for frontier research on galaxy-AGN co-evolution, quasar structure and physical properties, quasar classification, and astronomical statistics.

This paper employs the convolutional neural network (CNN) method from machine learning [?]. By using previously discovered FeLoBALQ spectra and normal quasar spectra as training sets, the program learns to identify FeLoBALQ spectra, which is then applied to the SDSS (Sloan Digital Sky Survey) DR7Q (Data Release 7 Quasar catalog) [?] to search for previously unidentified FeLoBALQ spectra, providing a solid sample foundation for future systematic studies. Since FeLoBALQs represent a small fraction of the population and their absorption features can be confused with other BALs, we restrict the redshift range of both training and target samples to $M(\text{cid:27})(\text{cid:3})M(\text{cid:0})(\text{cid:27})(\text{cid:3})0:8 < z < 2:1250:43\%$ to ensure that the absorption features of interest in the rest-frame wavelength range of 2000–3200 Å fall entirely within the effective spectral coverage.

Section 2 introduces the training and target samples, the CNN methodology, and important details of our research process. Section 3 presents the FeLoBALQ identification results and potential improvements. Section 4 summarizes the entire paper, clarifies its scientific significance, and outlines future research directions.

2.1 Redshift Selection and Target Sample

During preliminary testing using unprocessed spectra for CNN training and identification, we found that many spectra classified as FeLoBALQ showed only CIV or Mg II absorption components without clear Fe II absorption. Initial analysis revealed that different types of BALs have similar absorption profiles—all resembling Gaussian functions except for FeLoBALs, which produce continuous absorption troughs. Combined with redshift shifting absorption lines to arbitrary positions in the spectrum, the program struggled to distinguish whether absorption lines originated from Fe II rather than other elements.

Therefore, we truncated the spectra for both training and target samples, using only the rest-frame wavelength range of 2000–3200 Å that likely contains strong Fe absorption features. We retained Mg II as a reference while minimizing interference from CIV, Al III, and other contaminating features. The primary flux contributions in this band come from the quasar continuum, iron continuum emission, and the Mg II broad emission line, upon which iron absorption features are prominently displayed with minimal interference from other absorption components except for the Mg II broad absorption line. Based on

the SDSS BOSS (Baryon Oscillation Spectroscopic Survey)/eBOSS (extended BOSS) spectral wavelength range of 3600-10000 Å, we selected the redshift range $0.8 < z < 2.125$ to ensure the 2000-3200 Å rest-frame band is fully covered. For quasars in our SDSS DR7Q sample with redshifts around 2, the red end may not reach 3200 Å in the rest frame, but this does not significantly affect FeLoBALQ identification, and we include them to facilitate future comparative studies with SDSS DR16Q (Data Release 16 Quasar catalog) [?].

We selected the SDSS DR7Q catalog, applying redshift constraints to obtain 50,931 spectra as our target sample. This sample represents an important milestone of SDSS, with well-established scientific validity and a moderate size that facilitates methodological iteration and research development. It also allows convenient extension to the SDSS DR16Q sample.

Since many quasar spectra have low signal-to-noise ratios, using SDSS spectral intersection masks (and mask) or union masks (or mask) to flag bad pixels would eliminate substantial useful information. Therefore, we employed standard deviation clipping (-clipping) to mask interference such as sky lines. We then transformed the spectra to the rest frame, extracted the 2000-3200 Å wavelength range, and resampled to one data point per Ångström (to satisfy CNN requirements for uniform data length). We subtracted the mean flux of this band from each data point to reduce the impact of the continuum and highlight absorption and emission features. Finally, we applied smoothing to the spectra before CNN input to reduce noise interference and emphasize macroscopic features such as broad absorption lines.

2.2 Training Sample Selection

To train the CNN program, we compiled FeLoBALQs identified in previous studies [?], totaling 220 FeLoBALQ sources. We used their SDSS spectra as training samples. Among these 220 sources, 186 were located in SDSS. Counting multiple observations of the same source, we obtained 354 FeLoBALQ spectra. To ensure training quality, we retained only high-quality spectra with prominent iron absorption features, leaving 319 FeLoBALQ spectra. After applying the redshift restriction, the sample contained 167 spectra. During trial runs to validate CNN performance, we identified 33 additional FeLoBALQ spectra, which were added to expand the training sample. The final training sample included 200 FeLoBALQ spectra. We also randomly selected 379 non-FeLoBAL quasar spectra of varying quality to help the program learn to distinguish unique FeLoBALQ features. We used these 579 spectra as our total CNN training sample, randomly dividing them in a ratio of approximately 2% (FeLoBALQ) to 98% (non-FeLoBALQ). The training set contained 153 FeLoBALQ and 282 non-FeLoBALQ spectra for CNN training, while the test set contained 47 FeLoBALQ and 97 non-FeLoBALQ spectra to determine when the CNN achieved stable identification accuracy (using the model at that iteration stage for target sample classification). We applied the

same data processing pipeline to training sample spectra as used for the target sample.

2.3 CNN Structure and Training

We chose the CNN method for spectral identification primarily because CNNs excel at image recognition, and spectra can be treated as one-dimensional images. We constructed the CNN with five alternating convolutional and pooling layers to extract spectral features. The convolutional layers had kernel sizes of 3, 3, 5, 5, and 5, respectively, while all pooling layers used kernel size 2. The model then fed into two fully connected layers with 360 and 50 hidden units for FeLoBALQ classification. The neural network employed Rectified Linear Units (ReLU) [?, ?] as activation functions to improve noise robustness and avoid gradient vanishing during training. The neural network structure is illustrated in Figure 1 [Figure 1: see original paper].

Figure 1 Convolutional neural network structure used in this research.

We input the training sample into the CNN network, using the training set to train the model and the test set to validate results and provide feedback. After 20 iterations, the classification accuracy on the test set stabilized at M(cid:27)(cid:3)M(cid:0)(cid:27)(cid:3)2%(cid:0)5%18%(cid:12). Therefore, we selected the model after 20 iterations to classify the target sample. The confusion matrix at this stage is shown in Figure 2 [Figure 2: see original paper].

Figure 2 Confusion matrix for the test set when the CNN stabilized at the 20th iteration.

3 Results

We input the target sample into the trained CNN model, which identified approximately 1,000 FeLoBALQ candidate spectra. After manual inspection, we excluded spectra with weak iron absorption features that were difficult to confirm as FeLoBALQs and removed spectra already present in the training sample, yielding 160 newly discovered FeLoBALQ spectra (including 33 identified during the trial phase). Example spectra are shown in Figure 3 [Figure 3: see original paper]. Appendix 1 provides detailed information for the newly discovered FeLoBALQs, including SDSS name, right ascension (RA) and declination (Dec), SDSS spectral identification (plate number, Modified Julian Date MJD, and fiber number), redshift (z), SDSS u, g, r, i, z band magnitudes, and WISE (Wide-field Infrared Survey Explorer) W1 and W2 band magnitudes.

Figure 3 Example spectra of four newly identified FeLoBALQs in this research. Each panel shows the locally observed full spectrum in the inset (top left), with local observed wavelength on the x-axis and local observed flux on the y-axis. Red dashed lines mark 2000 Å and 3200 Å in the rest frame, representing our wavelength range of interest. The main plot shows the spectral

detail in the rest-frame wavelength range used by our CNN, transformed using each FeLoBALQ's redshift value. Green and cyan dashed lines indicate absorption lines from Hall et al. [?] Table 1 (cyan marks Mg II absorption, a hallmark of BALs), blueshifted from the tabulated wavelengths. The blueshift velocity (outflow velocity) is given in each title. The tabulated absorption lines cannot fully match the observed features, and varying absorption strengths and widths among different FeLoBALQ spectra produce different absorption profiles, making conventional identification methods difficult and necessitating machine learning approaches.

We constructed color-color diagrams comparing the target sample (SDSS DR7Q quasars) with the training sample (previously known FeLoBALQs only, excluding trial-run identifications) and our 160 newly discovered FeLoBALQ candidates, shown in Figure 4 [Figure 4: see original paper]. Since all three samples lie primarily off the Galactic plane with minimal Milky Way extinction, and share similar redshift ranges, we calculated colors directly from observed magnitudes for statistical analysis. Relative to the target sample, FeLoBALQs are generally redder, with the difference more pronounced at bluer wavelengths and nearly vanishing in the mid-infrared W1-W2 color. Physically, the bluer portion of quasar spectra is increasingly dominated by continuum emission, and FeLoBALQ outflows produce stronger extinction of this continuum, resulting in redder colors. Figure 4 supports this interpretation. The newly discovered FeLoBALQs show similar color distributions to previously known ones, though the previously known objects are slightly redder overall. This can be explained by the following: stronger absorption in FeLoBALQ outflows produces redder colors; objects with stronger absorption exhibit more prominent absorption complexes and troughs in their spectra, making them easier to discover in earlier studies and causing them to appear systematically redder. Conversely, FeLoBALQs with weaker absorption are generally discovered later, comprising a larger fraction of our new candidates.

Figure 4 Color-color diagram for FeLoBALQs and SDSS DR7Q quasars for statistical analysis, using four colors calculated from SDSS g, r, i, z band magnitudes and WISE W1, W2 band magnitudes. Red dots represent FeLoBALQs newly identified in this research, green dots represent previously identified FeLoBALQs, and black dots represent DR7Q quasars. FeLoBALQs are redder than other quasars, with previously identified ones slightly redder than newly identified ones. This effect is more obvious at the bluer end than the redder end, and color differences nearly disappear in mid-infrared W1-W2.

The training sample contains 61 spectra from SDSS DR7Q. Currently, 221 FeLoBALQ spectra have been discovered within the SDSS DR7Q range, representing $M(\text{cid:27})(\text{cid:3})M(\text{cid:0})(\text{cid:27})(\text{cid:3})0.43\%$ of the total 50,931 spectra—slightly higher than the 0.33% fraction reported by Trump et al. [?] in 2006. We believe the true fraction of FeLoBALQs among all quasars should be larger than this value. First, FeLoBALQs are redder than other quasars, and outflow absorption makes them fainter in the optical and near-infrared bands as

observed from Earth, increasing confusion with brown dwarfs and causing them to be missed during optical target selection. Second, the dramatic absorption features in FeLoBALQ spectra make classification and redshift determination difficult (for example, some FeLoBALQs in our training sample were classified as galaxies by SDSS), causing them to escape target sample selection.

4 Summary and Outlook

In this paper, we used previously identified FeLoBALQ spectra as training samples to establish a CNN model for FeLoBALQ identification and classification. Based on the absorption characteristics of FeLoBALQs, we determined a redshift range of $M(\text{cid:27})(\text{cid:3})M(\text{cid:0})(\text{cid:27})(\text{cid:3})0.8 < z < 2.1250.43\%$ for our target quasars. We classified all 50,931 quasar spectra from SDSS DR7Q in this range, discovering 160 new FeLoBALQ candidates and estimating that FeLoBALQs constitute $M(\text{cid:27})(\text{cid:3})M(\text{cid:0})(\text{cid:27})(\text{cid:3})0.8 < z < 2.1250.43\%$ of quasars in this redshift interval. We briefly analyzed why FeLoBALQs appear redder than other quasars and discussed potential statistical biases in our study.

Future work will include continuous improvement of the program and CNN architecture, refinement of training samples, and extension of FeLoBALQ identification to the SDSS DR16Q sample (330,386 spectra, excluding those from SDSS DR7Q) in the same redshift range. We expect to discover approximately 1,400 FeLoBALQ spectra in this sample, with about 1,100 being newly identified, thereby substantially expanding the existing FeLoBALQ catalog and providing data support for related research.

Additionally, we plan to conduct high-resolution imaging, spectroscopic, and multi-band photometric observations of FeLoBALQ candidates in our sample. High-resolution imaging will search for signs of major mergers in host galaxies, addressing whether FeLoBALQs are primarily triggered by major mergers. High-resolution spectroscopy and multi-band photometry will provide more precise confirmations and determine physical parameters including black hole mass, bolometric luminosity, outflow velocity and strength, and host galaxy star formation rates. Analysis of these collected physical properties will help answer whether FeLoBALQs are associated with ULIRGs and whether they represent an evolutionary stage from ULIRGs to quasars. Furthermore, by comparing physical properties of FeLoBALQs at different redshifts, we will investigate whether the FeLoBALQ occurrence rate correlates with redshift and cosmic age, which is crucial for understanding galaxy-supermassive black hole co-evolution.

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and specutils packages. We express our gratitude to these projects, databases, and software packages. We thank the referee for valuable comments and Li Ruancun for helpful suggestions.

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Appendix 1 Table of Newly Discovered FeLoBALQ Sources

Table 1 FeLoBALQs newly identified in this study

SDSS Name	RA	Dec	Plate	MJD	Fiber	z	u	g	r	i	z	W1	W2
J1050-0105	163.8653
		1.0988											

SDSS														
Name	RA	Dec	Plate	MJD	Fiber	z	u	g	r	i	z	W1	W2	
J1055+00263.8653	3387
J1240+00190.1102	2953
J1310+01097.5286	6735
J1316-	199.0434
0036	0.6100													
J1339+00094.9391	1589
J1447-	221.9385
0034	0.5677													
J1209+00182.3804	4544
J2336-	354.0709
0107	1.1260													
J0338+0056.5410	59389
J0156+1332.1237	38708	
J0207+1429.8638	44980	
J0805+45321.3855	5955	
J0242-	40.6830
0722	7.3682													
J0918+58339.7258	5587	
J0948+62147.1962	2407	
J1239+02359.8925	4939
J1010+57232.5952	23899	
J1020+03365.0533	4086
J1056+05164.0331	4308
J1457+04264.3734	4419
J1420+60275.0409	7950	
J1628+47317.0227	89252	
J1649+40242.4245	2408	
J2347-	356.7977
1037	10.6285													
J2357-	359.2702
0903	9.0540													
J0102-	15.5064
0853	8.8957													
J0121-	20.3620
0929	9.4847													
J2137+10324.3110	6864	
J2204+13331.0913	4561	
J0026+13556667	13.9258	
J1120+62070.0062	90832	
J1202+63180.5088	2998	
J1448+56322.1610	6553	
J1534+51233.6058	7924	
J1530+51232.6150	2540	

SDSS														
Name	RA	Dec	Plate	MJD	Fiber	z	u	g	r	i	z	W1	W2	
J1233+06088.36521357..			
J1319+05109.96523381..			
J1337+05204.47111592..			
J1344- 206.2090		
0158		1.9786												
J1348- 207.1283		
0153		1.8994												
J1527- 231.8533		
0210		2.1820												
J0743+22095.8371.1625		
J1010+45182.7249.3047		
J1011+57122.8657.2049		
J1006+05131.5230.2208..		
J0801+31130.4131.18916		
J2140- 325.0573		
0653		6.8969												
J0834+06138.6812.2214..		
J1145+11076.4841.0051.		
J1041+09250.4977.1236..		
J0811+24122.7521.18753		
J0912+35138.1255.2607		
J1302+48195.7349.9979		
J1333+49263.4245.2731.		
J0912+09138.2302.2542..		
J1239+57159.8757.9288		
J1359+55309.7715.5383		
J1054+44283.5441.4790		
J1440+37200.0037.1829		
J1155+46178.7947.9844		
J0003+00008580.1750..		
J1646+19231.7298.7169		
J1612+25243.1728.2691.		
J1617+25094.4285.1595		
J1539+32034.9962.0863		
J1547+32306.9237.5133		
J0823+24405.9626.7814		
J1539+07374.9619.0222..		
J1610+07212.5280.2208..		
J0743+43175.9142.9516		
J0949+12377.4282.6292		
J1022+13015.7053.0236		
J1043+13270.8329.4635		
J1102+14385.5591.6479		

SDSS													
Name	RA	Dec	Plate	MJD	Fiber	z	u	g	r	i	z	W1	W2
J0832+07188.1919661..
J0825+06126.2604681..
J1222+14135.65693975
J1336+08304.1352086..
J1335+10183.85633108
J1432+05278.09547897..
J1507+29006.78201020
J1543+26205.76256812
J1553+23288.40219767
J1548+24327.152185477
J1551+25237.752717400
J0800+19430.06287332
J1022+34195.56379916
J1029+37017.432730242
J1326+28391.512816660
J1232+41068.07511105
J1234+38388.64326542
J1147+37376.913715775
J1306+32196.673299676
J1245+34191.37302026
J0206+22271.684224567
J0133+2333.408235583
J1409+27292.27204947
J1635+14218.89572657
J1215+29343.94235694
J1230+27187.74229901
J1248+28402.14207691
J0813+13423.38747543
J0033+068249236.5403..
J1017+23034.342300562
J1000+20060.21251102
J1022+21155.60229758
J0249+3320.34033.3347
J1042- 160.6780
0008 0.1392
J1045+22161.25221779
J1050+20042.622710728
J1114+22208.502213699
J1133+22373.28225186
J1632+09218.0827069..
J0849+11232.301154273
J1209+16312.29168590
J1214+17493.625718187

SDSS														
Name	RA	Dec	Plate	MJD	Fiber	z	u	g	r	i	z	W1	W2	
J1227+17186.90110.8497
J1259+20174.79201.9606
J1305+19326.4866.5444
J1334+20003.5629.0771
J1247+23291.9028.3933
J1316+25099.0522.1236
J1338+22264.6025.7751
J1439+15319.9281.5876
J1524+11231.0966.8867
J1417+17234.2783.7187
J1533+15003.4215.0165
J1500+16285.0874.8152
J1533+56203.3126.8458
J1059+09364.8402.8444
J0753+08228.3282.7190
J0925+32111.4032.8502
J0921+32440.4531.7330
J0914+32138.6932.8908
J1035- 158.9373
0029	0.4902													
J1053- 163.4703
0058	0.9813													
J1056+00124.0902.2034
J1104- 166.1701
0004	0.0782													
J1107- 166.8686
0051	0.8563													
J1133- 173.4220
0057	0.9611													
J1158- 179.7203
0043	0.7172													
J1607+00381.9950.9711
J1716+64349.0168.5730
J1723+65260.8089.7962
J1733+55203.3785.3419
J1729+55262.2896.7857
J1725+52261.2522.9139
J1736+58264.2428.9172
J2339- 354.8934
0029	0.4924													
J0321- 50.3167
0013	0.2230													
J0925+01361.2972.2036

SDSS													
Name	RA	Dec	Plate	MJD	Fiber	z	u	g	r	i	z	W1	W2
J1235+0138.9585536
J0857+5034.46735234
J1040+6120.20323792
J0849+02132.4732881
J1045+05401.3088675
J1432+03218.2218814
J1355+61208.77618694
J1507+58026.91380703

Note: The complete table with all data columns is available in the electronic version of this paper.

Note: Figure translations are in progress. See original paper for figures.

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