Preliminary clustering properties of the DESI BGS bright targets using DR9 Legacy Imaging Surveys

Pauline Zarrouk,1,2,3 Omar Ruiz-Macias,4,5,6 Shaun Cole,2 Peder Norberg,2,4 Carlton Baugh,2,3 David Brooks,5 Enrique Gaztañaga,6 Ellie Kitanidis,7 Robert Kehoe,8 Martin Landriau,9 John Moustakas,10 Francisco Prada11 and Gregory Tarlé12

1 Sorbonne Université, Université Paris Diderot, CNRS/IN2P3, Laboratoire de Physique Nucléaire et de Hautes Energies, LPNHE, 4 place Jussieu, F-75252 Paris, France
2 Institute for Computational Cosmology, Department of Physics, Durham University, South Road, Durham DH1 3LE, UK
3 Institute for Data Science, Durham University, South Road, Durham DH1 3LE, UK
4 Centre for Extragalactic Astronomy, Department of Physics, Durham University, South Road, Durham DH1 3LE, UK
5 Department of Physics & Astronomy, University College London, Gower Street, London WC1E 6BT, UK
6 Institute of Space Sciences (ICE, CSIC), Campus UAB, Carrer de Can Magrans, sn, E-08193 Bellaterra, Barcelona, Spain
7 Department of Physics, University of California, 366 LeConte Hall, Berkeley, CA 94720, USA
8 Department of Physics, Southern Methodist University, 3215 Daniel Avenue, Dallas, TX 75205, USA
9 Lawrence Berkeley National Laboratory, 1 Cyclotron Road, Berkeley, CA 94720, USA
10 Department of Physics and Astronomy, Siena College, 515 Loudon Road, Loudonville, NY 12211, USA
11 Instituto de Astrofísica de Andalucía (CSIC), Glorieta de la Astronomía, s/n, E-18008 Granada, Spain
12 Department of Physics, University of Michigan, Ann Arbor, MI 48109, USA

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ABSTRACT

We characterize the selection cuts and clustering properties of a magnitude-limited sample of bright galaxies that is part of the Bright Galaxy Survey (BGS) of the Dark Energy Spectroscopic Instrument (DESI) using the ninth data release of the Legacy Imaging Surveys (DR9). We describe changes in the DR9 selection compared to the DR8 one and we also compare the DR9 selection in three distinct regions: BASS/MzLS in the north Galactic Cap (NGC), DECaLS in the NGC, and DECaLS in the south Galactic Cap (SGC). We investigate the systematics associated with the selection and assess its completeness by matching the BGS targets with the Galaxy and Mass Assembly (GAMA) survey. We measure the angular clustering for the overall bright sample (r_{mag} ≤ 19.5) and as function of apparent magnitude and colour. This enables to determine the clustering strength r_0 and slope γ by fitting a power-law model that can be used to generate accurate mock catalogues for this tracer. We use a counts-in-cells technique to explore higher order statistics and cross-correlations with external spectroscopic data sets in order to check the evolution of the clustering with redshift and the redshift distribution of the BGS targets using clustering redshifts. While this work validates the properties of the BGS bright targets, the final target selection pipeline and clustering properties of the entire DESI BGS will be fully characterized and validated with the spectroscopic data of Survey Validation.

1 INTRODUCTION

Building ever larger and more accurate 3D maps of the large-scale structure of the Universe is the key to understanding the formation and evolution of the Universe, and, in particular, the late-time acceleration of its expansion. However, the mechanism that is responsible for this acceleration remains a major unknown in modern cosmology. In order to account for this discovery, the cosmological constant Λ was re-introduced in Einstein’s equations of General Relativity, such that it represents the major component of the current energy content of the Universe. The observations could also be explained by more complex dark energy models with time-dependent properties or even a modification of General Relativity on cosmological scales (e.g. Linder & Cahn 2007; Guzzo et al. 2008; Alam et al. 2020).

The Dark Energy Spectroscopic Instrument (DESI; DESI Collaboration 2016b) has been designed to meet the goal of elucidating the cosmic acceleration and it is the first of the new generation of galaxy surveys, the so-called Stage IV dark energy experiments using the terminology of the Dark Energy Task Force report (Albrecht et al. 2006), that has already started taking data. DESI is a multifiber spectrograph installed on the Mayall 4-meter telescope at Kitt Peak, Arizona, in the United States, which can collect 5000 spectra simultaneously using robot fiber positioners.

In order to create these 3D maps, we first require imaging surveys that provide the targets we are going to put a fiber on with the DESI instrument so that we can extract accurate and reliable positional information in the radial direction. We use three different imaging surveys that provide optical data over 20 000 deg^2 for DESI targeting

* E-mail: pauline.zarrouk@lpnhe.in2p3.fr
and constitute the Legacy Imaging Surveys (Dey et al. 2019): (i) the Beijing–Arizona Sky Survey (BASS) imaged regions at Dec. $\geq +32$ deg in the Northern Galactic Cap (NGC), in the g and r bands, (ii) the Mayall z-band Legacy Survey (MzLS) imaged the Dec $\geq 32$ deg region and (iii) the DECaLS program that made use of other DECam data within the Legacy Surveys (LS) footprint.

DESI will conduct dark time and bright time programs in parallel. During the dark time program four main target classes will be observed: about 6 million luminous red galaxies (LRG) in the redshift interval $0.4 < z < 1.0$, 17 million [OII] emission-line galaxies (ELG) in $1.0 < z < 1.6$ and 2.5 million quasars (QSO), where quasars with $z > 2.1$ will serve as direct tracers of the matter density field and quasars with $2.1 < z < 3.5$ will provide Lyman α absorption features in order to probe the distribution of neutral hydrogen in the intergalactic medium. During bright time, when lunar conditions are too bright for good observations of these faint objects, DESI will obtain 10 million spectra of bright galaxies up to $z \sim 0.4$ (Bright Galaxy Survey; BGS) and a sample of local stars (Milky Way Survey; MWS). The DESI footprint will cover 14 000 deg$^2$.

Here, we focus on the DESI BGS program that will likely be comprised primarily of galaxies selected to a limiting magnitude. The dark energy science goals of the BGS are to produce a data comprised primarily of galaxies selected to a limiting magnitude. The core of this selection for the bright sample remains unchanged: a bright sample at $r < 19.5$ (BGS Bright) and a faint sample at $19.5 < r < 20$ (BGS Faint). From here out, in Section 2.1 we first describe the changes in the imaging pipeline for DR9 and the changes in the target selection cuts and then we present a comparison between BASS/MzLS and DECaLS in Section 2.2. We quantify the impacts of the changes between DR8 and DR9 for the masking around bright stars in Section 2.3 and for the large galaxies in Section 2.4.

2 BGS TARGET CATALOGUE USING DR9

A candidate BGS selection using DR8 DECaLS was first explored in Ruiz-Macias et al. (2021) and also in a research note (Ruiz-Macias et al. 2020). The core of this selection for the bright sample remains unchanged: a bright sample at $r < 19.5$ (BGS Bright) and a faint sample at $19.5 < r < 20$ (BGS Faint). From here out, in Section 2.1 we first describe the changes in the imaging pipeline for DR9 and the changes in the target selection cuts and then we present a comparison between BASS/MzLS and DECaLS in Section 2.2. We quantify the impacts of the changes between DR8 and DR9 for the masking around bright stars in Section 2.3 and for the large galaxies in Section 2.4.

2.1 Changes between DR8 and DR9

2.1.1 Changes in the imaging pipeline

Compared to its predecessor, DR9 incorporates some major and minor changes that affect the photometry of the objects, and hence the target selection. These changes are incorporated in legacypipe, which is the DESI Imaging Legacy Surveys data reduction pipeline. Below we list the most important changes relevant for BGS target selection:

(i) Iterative detection: After the first round of fitting, TRACTOR$^2$ (Lang, Hogg & Mykytyn 2016) conducts a second round of detections over the data–model residuals with the aim of finding additional sources.

(ii) Extended PSF model: An extended PSF model is used to subtract the wings of bright stars from DECam images only.

(iii) Sersic fitting model: The composite (COMP) morphological model has been replaced by a Sersic profile (SER). A source is classified as SER if a Sersic profile provides a better fit than other profiles, PSF, EXP, and DEV, as quantified by a $\chi^2$ that takes account of additional free parameters of the Sersic fit.

(iv) Relaxed Gaia PSF criterion: TRACTOR forces Gaia objects to be fitted as PSF GAIA sources if they meet the condition $G \leq 18$ AE$^N < 10^{-3}$ OR $(G \leq 13)$ that was previously set to $(G < 19$ and $AEN < 10^{-5})$ OR $(G < 19$ and $AEN < 10^{-5} + 0.2G − 19)$. Here $G$ is GAIA DR2 G-band magnitude, and AEN is the Gaia astrometric excess noise parameter.

(v) Pre-fitting for large sources: Regions around large galaxies and globular clusters have their own local source extraction, which is performed separately from the normal TRACTOR run. The parent catalogues of these objects have improved since DR8 and this constitutes the Siena Galaxy Atlas 2020$^3$ (SGA 2020 Moustakas et al. 2021).

Footnotes:

1. https://github.com/legacysurvey/legacypipe
2. https://github.com/dsttndstr/tractor
3. https://sga.legacysurvey.org
were missing about 70 true galaxies deg$^{-2}$ (or is a fake source) in the old FRACS. In Section 2.1.4, we assess again the completeness as just new FRACS. The new FRACS is a subset of around 1/3 of the old FRACS. We will refer to DR8 QC as old FRACS, and for the QC in this work we will use the term ‘quality cuts’ (QC) defined in Ruiz-Macias et al. (2021), we assess the completeness of the previous selection using DR8 by matching the BGS targets with GAMA DR4 and found that we were missing about 25 true galaxies deg$^{-2}$. Compared with the BGS selection defined for DECaLS DR8, the current BGS selection with DR9 increases the overall target density by ~20 objects deg$^{-2}$. Table 1 shows the gain in target density for BGS BRIGHT and BGS FAINT and for the three regions, together with the resulting target density for this selection and the corresponding effective area after accounting for the spatial masking.

2.1.2 Changes in the BGS target selection

These improvements led to changes in the target selection pipeline that are summarized below:

(i) Bright stars: the size of the masking radius around bright stars is reduced by a factor of 2. In Section 2.3, we investigate the consequences of such a change in the BGS selection.

(ii) Large galaxies: the size of the masking radius around large galaxies in DR9 has increased from roughly 10 to 30 arcsec; however, the pre-fitting for large sources led to major improvements in the photometry extraction of these objects which allows us not to apply this mask in DR9. Moreover, we remind the reader that in Ruiz-Macias et al. (2021), we assess the completeness of the previous selection using DR8 by matching the BGS targets with GAMA DR4 and found that we were missing about 25 true galaxies deg$^{-2}$.

(iii) Quality cuts (QC): the ‘quality cuts’ (QC) defined in Ruiz-Macias et al. (2021) for DECaLS DR8 are cuts in FRACMASKED$_j < 0.4$, FRACIN$_j > 0.3$, FRACFLUX$_j < 5$, where $i \equiv g$, $r$, or $z$. FRACIN is used to select sources for which a large fraction of the model flux lies within the contiguous pixels to which the model was fitted, FRACFLUX is used to reject objects that are swamped by flux from adjacent sources, and FRACMASKED is used to veto objects with a high fraction of masked pixels. The overall improvement in the quality of the photometry assessed in the next Sections 2.1.3 and 2.1.4 allows us to adopt a less conservative definition of the QC.

2.1.3 Visual inspection on the imaging

The goal of performing an imaging VI is to assess our selection and reduce human uncertainties in the classification by involving as many people as possible. For this purpose, we use the LSVI (Legacy Surveys Visual Inspection), an interactive web framework that automates this process. LSVI essentially create postage stamps of the Legacy Surveys Sky Viewer web site viewer (Lang, in preparation) with their corresponding classification labels.

To assess the conservative selection of the QC we adopt for DR9, we took a sparsely selected sample of 2000 BGS targets in an ~420 deg$^2$ area in DECaLS DR9 flagged by the old FRACS. This sample accounts for ~35 per cent out of the total available in this area. The classification labels are, GAL: The object is a galaxy; STAR The object is a star; BT: The object is contaminated by (or is a fake source from) a bleed trail; DS/H: The object is contaminated by (or is a fake source from) a diffraction spike or a star halo; FRAG: The object is spurious from a fragmented large galaxy; JUNK: The object does not fit in any of the previous classifications but it is clearly spurious; UNK: The object does not fit in any of the previous classifications and it’s unclear whether it is a galaxy or not; NI: The object has not yet been inspected. A total of six people participated in the classification, and the results where classified in three categories: (i) confirmed galaxy (CG): two or more people say the object is a galaxy and less than two people say it is anything but a galaxy; (ii) confirmed non-galaxy (CNG): two or more people say the object is anything but a galaxy and less than two people say it is a galaxy; and (iii) inconsistent classification (IC): is neither of previous categories. Out of the 2000 objects, 35 per cent are CG, 43 per cent are CNG, and 22 per cent are IC. The CNG include stars and other artifacts such as bleed trails, diffraction spikes, stellar haloes, and fragmented images. There is no easy way to isolate the CG while getting rid of spurious sources. However, for the subset of these old FRACS, the new FRACS defined by equation (1) get rid of around 10 per cent of the CG, around 60 per cent of the CNG, and around 30 per cent of the IC.

2.1.4 Validation with GAMA

In DR8 DECaLS, we assess the completeness of our BGS catalogue with respect to GAMA, whose main target sample (Baldry et al. 2017) is highly complete for galaxies with $14 < r_{\text{SDSS}} < 19.8$, where $r_{\text{SDSS}}$ is the SDSS Petrosian $r$-band magnitude, with 98.85 per cent of the objects in the catalogue having good redshifts with a quality flag $\text{NO} \geq 3$, after applying a redshift cut at $z > 0.002$ to remove the remaining stars. We match the BGS targets with the GAMA Main Survey DR4 galaxy catalogue$^4$ using a maximum linking length of 1 arcsec and we focus on three of the five GAMA fields: G09, G12, and G15. First, we cross-matched the GAMA catalogue with the Legacy Imaging Surveys DR9. This results in a matched catalogue of 1007.5 objects deg$^{-2}$ in a 178 deg$^2$ area, which represents 99.8 per cent of all the GAMA galaxies. We define four different BGS selections to test

Table 1. Target density in objects deg$^{-2}(\eta)$, increase in target density in objects deg$^{-2}(\Delta \eta)$ of the current BGS target selection (DR9) compared with the BGS selection defined for DECaLS DR8 and the effective area ($A_{\text{eff}}$) in deg$^2$ after accounting for the spatial masking. We show results for BGS BRIGHT and BGS FAINT each divided into three regions: BASS/MzLS, DECaLS NGC, and DECaLS SGC.

| BGS       | BASS/MzLS | DECaLS NGC | DECaLS SGC |
|-----------|-----------|------------|------------|
| $\eta_{\text{bright}}$ | 813.4     | 875.6      | 839.0      |
| $\eta_{\text{faint}}$  | 569.4     | 598.8      | 581.1      |
| $\Delta \eta_{\text{bright}}$  | +12.7     | +14.1      | +12.5      |
| $\Delta \eta_{\text{faint}}$ | +7.7      | +8.2       | +7.3       |
| $A_{\text{eff}}$      | 4493      | 5263       | 4326       |

4https://lsvi-webtool.herokuapp.com/  
5Legacy Surveys / D. Lang (Perimeter Institute) legacysurvey.org/viewer  
6This is an unreleased version that the GAMA collaboration made available to us. It is essentially the same as GAMA DR3, but with more redshifts.
In order to compare the photometry in DECaLS and BASS/MzLS, we find all target matches within a distance of 0.5 arcsec. To avoid incomplete regions, we require NOBS$_i > 0$ for $i = g, r, \text{and} \ z$ for the three surveys. After the match, we find agreement between 1328 BGS objects deg$^{-2}$, and disagreement for 66 objects deg$^{-2}$ that are in BGS in DECaLS but which are not BGS in BASS/MzLS; conversely there are 28 BGS objects deg$^{-2}$ in BASS/MzLS that are not in the DECaLS BGS.

We find that most of the disagreements are due to a shift in the $r$-band magnitude. Fig. 2 shows $r'_{\text{DECaLS}} - r'_{\text{BASS-MzLS}}$ as a function of $(g - r)'_{\text{BASS}}$ for two samples: one with only stars in both BASS/MzLS and in DECaLS, and the other with only BGS objects. We can see that the $r$-band magnitude in BASS/MzLS is typically fainter than in DECaLS. We can fit the magnitude difference as a linear function in $(g - r)'_{\text{BASS}}$. Equations (2) and (3) show these $r$-band photometry transformations of the BASS/MzLS system to the DECaLS system for BGS matches ($r'_{\text{gg}}$) and for star matches ($r'_{\text{as}}$), respectively.

\begin{align}
 r'_{\text{gg}} &= r_{\text{BASS}} - 0.039(g - r)_{\text{BASS}} + 0.011 \\
 r'_{\text{as}} &= r_{\text{BASS}} - 0.035(g - r)_{\text{BASS}} + 0.017.
\end{align}

The $r$-band magnitude offset is $r_{\text{BASS}} - r_{\text{DECaLS}} \approx 0.026$ for BGS objects. In Fig. 3, we show the number counts as a function of $r$-mag for BGS objects in both regions in the overlapping 76 deg$^2$ region and compare this with number counts of BASS/MzLS after correcting the $r$-mag by applying equations (2) and (3). These results show that both regions can achieve similar target densities if we apply a linear transformation in BASS/MzLS, increasing the overall target density to 1430 objects deg$^{-2}$. Compared to the 1383 objects deg$^{-2}$ without the colour-correction, this represents an increase of 3.4 per cent.

### 2.3 Bright stars

The masking radius around bright stars has been reduced by a factor of 2 in DR9 compared to DR8. Fig. 4 shows the stacked average density of BGS close to bright stars in BASS/MzLS. Distances have been rescaled to the masking radius $R_{\text{m}}(m)$ where the DR9 masking radius is represented by the smaller black circle while the radius applied in DR8 is shown by the larger red circle. Contamination around bright stars seems to be higher for the brightest stars ($8 < m < 12$) but the density profile of BGS objects (solid red line in Fig. 4) shows that the contamination outside the DR9 masking radius is negligible. In the left-hand panel, the horizontal blue spikes correspond to bleed trails, further masking around those features may be required depending on the science application of the DESI BGS data set. Results for DECaLS look similar to Fig. 4 and we believe is not necessary to include the corresponding figures. Overall, in DECaLS and BASS/MzLS, the reduction in the masking radius increases targetable area, and the BGS sample increases by around 28 objects deg$^{-2}$.

In order to investigate the potential systematic effect associated with bright stars, we measure the angular cross-correlation between the BGS targets after masking and the stellar catalogue for BASS/MzLS, DECaLS-NGC, and DECaLS=SGC. We tested several configurations for the target selection cuts: (i) with and without applying the masking around large galaxies (LG), (ii) considering the three options for the ‘FRACS cut’: (i) not applying the FRACS (no FRACS), (ii) applying the conservative definition of DR8 (old FRACS), (iii) applying a less conservative definition (new FRACS). The consequences of these different choices for the measured angular cross-correlation function of the BGS targets with Gaia stars are shown in Fig. 5. As expected, the large galaxy mask has no effect on the stellar contamination and the other configurations show a negligible impact given the error bar. Therefore, we conclude...
stars was shown in Kitanidis et al. (2019). Moreover, we checked source objects. The correlation of these dark-time DESI targets with (ELG), which are fainter, or with Quasars (QSO), that are point-like seen for other DESI targets, such as the Emission Line Galaxies selection, we note that the effect remains small compared with that of stars, indeed, although stars represent the main systematic in the BGS autocorrelation functions. The BGS targets seem uncorrelated with the median number of objects within the hexagonal cell. The red solid line shows the median and the orange shaded region shows the 3 and 97 percentiles. The black dashed lines show the best straight line fits to the median relation as given in the green boxes above.

![Figure 2](image-url)  
**Figure 2.** DECaLS and BASS/MzLS matched Stars (left) and BGS targets (right) showing \( \frac{DECaLS}{BASS/MzLS} \) as a function of \((g-r)_{BASS}\) colour. Matched stars are shown in the left-hand side while BGS in the right-hand side. Stars are as defined by our Gaia star–galaxy classification. The colour bar indicates the number of objects within the hexagonal cell. The red solid line shows the median and the orange shaded region shows the 3 and 97 percentiles. The black dashed lines show the best straight line fits to the median relation as given in the green boxes above.

![Figure 3](image-url)  
**Figure 3.** The BGS target density \((n)\) divided by the target density of a fiducial linear fit, \(\log(n_0) = 0.46 \times r_{\text{mag}} + 6.10\). The solid black and red lines show the DECaLS and BASS/MzLS BGS target densities within their full footprints. The red dashed and dotted lines show BASS/MzLS BGS target density after applying the \(r\)-mag transformation equations (2) and (3), respectively.

| stars | BGS |
|---|---|
| \( \frac{DECaLS}{BASS/MzLS} \) | \( \frac{DECaLS}{BASS/MzLS} \) |
| \((g-r)_{BASS}\) | \((g-r)_{BASS}\) |
| \(N\) | \(N\) |

that none of these options should affect significantly the angular clustering of the BGS targets, as confirmed later in Section 3.1.

Fig. 6 shows the ratio between the cross-correlation function of BGS targets and stars, and the autocorrelation function of the stars. At large angular scales, this ratio gives an estimate of the fraction of stellar contamination in the BGS sample. The error bars are estimated using 100 jackknife regions for both the cross- and autocorrelation functions. The BGS targets seem uncorrelated with stars, indeed, although stars represent the main systematic in the BGS selection, we note that the effect remains small compared with that seen for other DESI targets, such as the Emission Line Galaxies (ELG), which are fainter, or with Quasars (QSO), that are point-source objects. The correlation of these dark-time DESI targets with stars was shown in Kitanidis et al. (2019). Moreover, we checked that at large angular scales, the cross-correlation signal tends to zero, which confirms the normalization is correct. This small scale-independent offset could be due to a gradient in stellar density. Fig. 6 shows that there is no significant stellar contamination in the BGS sample, which is also consistent with the results from the star–galaxy separation when considering Gaia objects. Furthermore, the DESI Survey Validation (SV) selection will test the possibility to relax the star–galaxy criterion.

### 2.4 Large galaxies

Further changes in the BGS selection include not masking around the Large Galaxies of the Siena Galaxy Atlas (SGA; Moustakas, Lang, et al. in preparation), and a revisiting of the quality cuts (QCs). The SGA galaxy catalogue has gone through a series of improvements and TRACTOR was run separately in regions within the confines of these galaxies, which has led to a reduction in the number of spurious objects around these galaxies in DR9 compared to DR8. These spurious objects were due to large galaxies not being appropriately fitted by TRACTOR and as a result, these galaxies were fragmented in many fake sources. In addition to that, in DR8, TRACTOR forced PSF fits to all the objects around large galaxies, compromising the photometry of the potential BGS targets in these regions and forcing us to mask them. In DR9, we have visually inspected around 1 per cent of the BGS within the SGA mask and found \( \sim 50 \) per cent of them are galaxies and the remaining \( \sim 50 \) are either stars or fragmented galaxies. We have decided to target all these to ensure completeness for clustering studies. We can reject spurious objects at a later stage. Using GAMA DR4, we match BGS with GAMA and for the matched objects within the LG mask, we were able to identify that around 40 per cent are spectroscopically confirmed galaxies that would otherwise be rejected by the LG mask. The comparison with GAMA shows that the QCs applied in DR8 erroneously rejected some sources that are galaxies: 80 per cent of the sources removed by the FRACIN, QC turn out to be galaxies, along with 50 per cent of the objects removed by the FRACMASKED, QC, and 20 per cent of objects in the case of FRACFLUX.

Large galaxies correspond to the brightest BGS galaxies in our sample. By comparing the angular cross-correlation between the large galaxies and BGS faint targets (with \(18 < r < 19\) for instance) and the angular cross-correlation between the BGS bright targets (\(15 < r < 16\) for instance) with the same BGS faint galaxies, we can estimate whether we have an excess or deficit of BGS targets in the vicinity of the large galaxies due to spurious or misclassified sources. A similar test for DECaLS DR8 was performed in Ruiz-Macias et al. (2021) and those results are shown in black on the top panel of Fig. 7. This panel also shows the result of this test when masking around the large galaxies in DR9, which also includes BASS/MzLS. In DR9, the median large galaxy masking radius is about 30 arcsec (shown by the dotted vertical line in Fig. 7) which is twice the size of that used.
Figure 4. 2D histograms of the positions of BGS objects from BASS/MzLS relative to their nearest Bright Star (BS) taken from the Gaia and Tycho catalogues down to $G$-mag and visual magnitude $\text{MAG}_{VT}$ of 13, respectively. These stacks are performed in magnitude bins in the bright stars catalogue from magnitude 8–12 (left) and 12–13 (right). The stacks are made using angular separations rescaled to the masking radius function of DR9 (inner black circle), which means that objects within a scaled radius of $0–1$ will be masked out by the BS veto while objects with $r/R_{BS} > 1$ will not (here $r^2 = (\Delta \text{RA}^2 \cos(\text{Dec})^2 + \Delta \text{Dec}^2)$). The colour scale shows the ratio of the density per pixel ($\eta$) to the mean density ($\bar{\eta}$) within the shell $1.1 < r/R_{BS} < 4$. The density ratio is shown on a log2 scale where red shows overdensities, blue corresponds to underdensities and white shows the mean density. The outer red circle shows the masking radius of the LS DR8 data. The red solid line shows the radial density profile on the same scale as the colour distribution $\log_2(\eta(R)/\bar{\eta})$ where $\eta(R)$ is the target density within the annulus at radius $R$ of width $\Delta R \sim 0.06$.

Figure 5. Angular cross-correlation function between the BGS Bright targets in each region and Gaia stars for different configurations of the BGS target selection. The Gaia stars have $12 < G < 17$. The shaded region shows the $1\sigma$ error estimated using a jackknife resampling of the data for ‘LG old fracs’ configuration only.

Figure 6. Ratio between the angular cross-correlation function between the BGS targets and Gaia stars and the autocorrelation function of Gaia stars. The dashed line shows this ratio after removing regions of high stellar density (stellar density $<1000$ objects deg$^{-2}$). The shaded region shows the $1\sigma$ error estimated using a jackknife resampling of the data. The Gaia stars have $12 < G < 17$.

in DR8 as one can see from the figure where the black curve drops at smaller scale. As expected, and as we found in DECaLS DR8, when masking is applied the dashed curves that correspond to the cross-correlation function between the large galaxies and the BGS faint targets drop dramatically on scales below the masking radius, meaning that we are missing BGS targets on these scales. The bottom panel of Fig. 7 shows a similar study but without applying the large galaxy mask. The solid and dashed curves now agree much better on scales below 30 arcsec, meaning that we recover the BGS targets in the vicinity of the large galaxies. However, the overall amplitude seems larger than what is obtained from the MXXL light-cone for BGS (Smith et al. 2017) when measuring the cross-correlation.
Figure 7. The angular cross-correlation function measured between faint BGS targets in $18 < r < 19$ and large galaxies from the LSLGA (dash–dotted) and between the same faint BGS targets and brighter BGS targets in $15 < r < 16$ (solid), the magnitude range in which most of the large galaxies reside. We also compare with the angular cross-correlation between these two bins in apparent magnitude measured in the MXXL light-cone (dashed). The vertical dotted line shows the mean LG mask radius which is about 20 arcsec. Top: with applying the large galaxy mask when selecting the BGS targets. Bottom: without applying the large galaxy mask.

Figure 8. Same as in Fig. 7 but here we focus on scales below the masking radius around large galaxies and we show the impact of different quality cuts (fracs). These quality cuts aim at removing spurious objects in the vicinity of the large galaxies without removing true BGS targets.

BGS targets in the vicinity of the large galaxies using the LSVI web tool (see Section 2.1.3) with the goal of determining the exact fraction of spurious objects and identifying common properties that could define more suitable cuts.

3 ANGULAR CLUSTERING MEASUREMENTS

While many cosmological studies require knowledge of the 3D distribution of galaxies, the angular clustering also provides valuable information about both cosmology and the galaxy–halo connection. In Section 3.1, we present a detailed study of the angular correlation function that includes a comparison with theory and the MXXL light-cone of Smith et al. (2017) for BGS. In particular, we study the consistency between the BASS/MzLS and DECaLS BGS catalogues in terms of the angular correlation function (Section 3.1.2), then we analyse the clustering as a function of magnitude (Section 3.1.3) and as a function of colour (Section 3.1.4). Finally, in Section 3.2, we investigate the higher-order statistics of the galaxy density field using counts-in-cells.

3.1 Angular correlation function

3.1.1 Methodology

We measure the angular correlation function, $w(\theta)$, of the BGS targets using the estimator of Landy & Szalay (1993):

$$w_{LS}(\theta) = \frac{D_1 D_2(\theta) - D_1 R_2(\theta) - D_2 R_1(\theta) + R_1 R_2(\theta)}{R_1 R_2(\theta)},$$

(4)

where $DD$, $DR$, and $RR$ are, respectively, the data–data, data–random, and random–random pair counts at average separation $\theta$. The random catalogue is provided by the Legacy Imaging Surveys.\footnote{https://www.legacyarchive.org/dr9/files/##random-catalogs-randoms} The form given in equation (4) is for the cross-correlation of two samples. For the autocorrelation function, the labels 1 and 2 are indistinguishable and this simplifies to $w_{LS}(\theta) = (DD - 2DR + RR)/RR$. We used the cross-correlation version of the estimator in Section 2.3 (BGS targets and stars) and in Section 2.4 (BGS targets and Large Galaxies). In Section 4, we will cross-correlate the BGS targets with external spectroscopic data sets. We use the publicly available code...
TWOPCF\(^8\) to compute the angular correlation function together with jackknife errors. These jackknife errors are obtained by dividing the footprint into 100 independent regions of similar area such that each region contains the same number of points in the random catalogue.

In order to characterize the clustering of the BGS targets, we compare it to theoretical predictions based on the halo model (e.g. Peacock & Smith 2000; Seljak 2000; Cooray & Sheth 2002). In the current paradigm of galaxy formation, galaxies form within dark matter haloes and the overall galaxy clustering can be modelled by two contributions: one contribution due to galaxy pairs within dark matter haloes (the 1-halo term) and another contribution due to galaxy pairs in separate haloes (the 2-halo term; see e.g. Benson et al. 2000; Zheng et al. 2005). When combined, these two terms result in an approximate power law, with a feature corresponding to the 1-halo to 2-halo transition occurring around few h\(^{-1}\) Mpc, the typical virial radius of a halo, as first measured in the SDSS Main Galaxy Sample (Zehavi et al. 2004). Then, to obtain a prediction for the observed angular clustering, \(w(\theta)\), based on a model for the full 3D clustering, \(\xi(r)\), we can use Limber’s approximation (Limber 1953) to project the real-space clustering into angular space, assuming a flat sky and small angular separations (for a discussion of the validity of Limber’s approximation, see Simon 2007):

\[
w(\theta) = \frac{2}{c} \left( \int_0^{\infty} dz H(z) \frac{dN}{dz} \right)^2 \int_0^{\infty} d\xi \left( r = \sqrt{u^2 + x^2(z)\theta^2} \right),
\]

where \(dN/dz\) is the normalized redshift distribution, \(x(z)\) is the comoving distance to redshift \(z\) and the integral takes account of the reduction or dilution of clustering due to the chance alignments of uncorrelated galaxies at significantly different redshifts along the line of sight. This dilution effect is larger when the sample covers a larger footprint into 100 independent regions of similar area such that each region contains the same number of points in the random catalogue.

Previous studies showed that the observed correlation function can be modelled as a single power law in \(r\) and \(z\) up to separations of about \(\sim 10 h^{-1}\) Mpc (e.g. Davis & Peebles 1983; Gaztanaga 1995; Maddox, Efstathiou & Sutherland 1996):

\[
\xi(r) = \left( \frac{r_0}{r} \right)^\gamma (1 + z)^{-(3+\epsilon)},
\]

where \(r_0\) is the clustering length, the scale at which \(\xi = 1\), and \(\gamma\) is the power-law slope. When the clustering properties do not evolve with proper coordinates, we have \(\epsilon = 0\) (Gaztanaga 1995). Assuming this power-law form for the correlation function, equation (6) becomes:

\[
w(\theta) = \theta^{\gamma - 1} r_0^2 \left( \frac{\Gamma(y/2 - 1/2)}{\Gamma(y/2)} \right)^2 \int_0^{\infty} d\bar{r} f(\bar{r})^2 (1 + z)^{\gamma - 3} r^{\gamma - 1}.
\]

This final equation can be considered as \(w(\theta) = A_{r_0,\gamma} \theta^{\gamma - 1}\), where the integral and \(\Gamma\) functions have been absorbed into a constant, \(A_{r_0,\gamma}\), whose value is set by the choices for \(\gamma\) and \(r_0\). Plotting \(w(\theta) \times \theta^{-(1 - \gamma)}\) will result in a constant if the power-law model is a good description of the measured angular clustering. As one can see, there is a degeneracy between the inherent clustering amplitude and the redshift distribution of the galaxies in the sample. In what follows, we will fit the observed angular clustering with this theoretical prediction in order to extract the clustering length \(r_0\) and slope \(\gamma\), using the \(dN/dz\)

\(^{8}\)https://github.com/lstothert/twopcf
with the errors when using 100 jackknife regions. We notice a slight underestimation when increasing the size of the jackknife region as expected, but otherwise the effect remains small which validates our interpretation of Fig. 10: the difference in clustering amplitude in this regime is consistent with being due to a statistical fluctuation.

### 3.1.3 Clustering as a function of magnitude

As an additional check for systematics, we compute the angular correlation function for different apparent magnitude bins and compare the results of the BGS targets with the MXXL simulation as shown in Fig. 11. The quantity plotted on the y-axis, \(w(\theta) \times \theta^{-(1-\gamma)}\), was chosen such that one can see the domain of validity of the power-law form, as for Fig. 9. Table 2 presents the results of the power-law fitting on both DR9 and MXXL for the five apparent magnitude bins we consider.

In order to help interpret these results we first quantify some properties of the matching MXXL mock catalogue. In Fig. 12, we see that the distribution of absolute magnitude has very little dependence on the apparent magnitude range of the sample. Hence, we would expect each of our apparent magnitude samples to be dominated by galaxies of the same absolute magnitude and hence have similar 3D clustering, \(\xi(r)\). The main way in which the samples differ in their normalized dN/dz shown in the bottom panel of Fig. 12.

#### Table 2. Best-fitting values for the clustering length, \(r_0\), and power-law slope, \(\gamma\), of the BGS targets in BASS/MzLS, DECaLS-NGC, and DECaLS-SGC, compared to the results from the MXXL light-cone simulation, when using a power-law approximation over the fitting range 0.001 < \(\theta\) < 1 deg.

| Data set       | \(r_0\) (h^{-1} Mpc) | \(\gamma\) |
|----------------|-----------------------|------------|
| BASS/MzLS      | 5.477 ± 0.117         | 1.792 ± 0.007 |
| DECaLS-NGC     | 5.653 ± 0.118         | 1.781 ± 0.007 |
| DECaLS-SGC     | 5.010 ± 0.079         | 1.818 ± 0.005 |
| MXXL           | 4.817 ± 0.106         | 1.789 ± 0.006 |
| \(< r_{mag} < 16\) |                      |            |
| BASS/MzLS      | 6.173 ± 0.703         | 1.642 ± 0.033 |
| DECaLS-NGC     | 4.413 ± 0.498         | 1.761 ± 0.036 |
| DECaLS-SGC     | 5.446 ± 0.558         | 1.698 ± 0.034 |
| MXXL           | 5.731 ± 0.628         | 1.736 ± 0.039 |
| \(< r_{mag} < 17\) |                      |            |
| BASS/MzLS      | 5.889 ± 0.359         | 1.744 ± 0.021 |
| DECaLS-NGC     | 5.309 ± 0.448         | 1.761 ± 0.027 |
| DECaLS-SGC     | 5.962 ± 0.368         | 1.715 ± 0.022 |
| MXXL           | 6.189 ± 0.181         | 1.753 ± 0.029 |
| \(< r_{mag} < 18\) |                      |            |
| BASS/MzLS      | 5.844 ± 0.198         | 1.776 ± 0.012 |
| DECaLS-NGC     | 6.226 ± 0.275         | 1.746 ± 0.015 |
| DECaLS-SGC     | 5.514 ± 0.225         | 1.793 ± 0.015 |
| MXXL           | 5.909 ± 0.206         | 1.788 ± 0.012 |
| \(< r_{mag} < 19\) |                      |            |
| BASS/MzLS      | 5.360 ± 0.146         | 1.750 ± 0.008 |
| DECaLS-NGC     | 5.444 ± 0.237         | 1.742 ± 0.013 |
| DECaLS-SGC     | 5.393 ± 0.122         | 1.745 ± 0.007 |
| MXXL           | 4.590 ± 0.140         | 1.803 ± 0.007 |
| \(< r_{mag} < 20\) |                      |            |
| BASS/MzLS      | 5.286 ± 0.098         | 1.725 ± 0.006 |
| DECaLS-NGC     | 5.336 ± 0.122         | 1.720 ± 0.007 |
| DECaLS-SGC     | 5.032 ± 0.100         | 1.740 ± 0.006 |
| MXXL           | 4.382 ± 0.107         | 1.774 ± 0.006 |
Therefore, massive galaxies with red colours typically have older stellar populations while galaxies with intermediate masses are bluer and younger with higher star formation rates.

In order to disentangle the colour, luminosity, and redshift dependence of the galaxy clustering, we compute the colour-dependent clustering in two apparent magnitude bins for both BGS DR9 and MXXL. For each apparent magnitude bin, we split the sample into the 50 per cent bluest galaxies and 50 per cent reddest galaxies using \( g - r \) colour. We found that considering a fixed fraction of blue/red galaxies instead of fixed colour cuts results in a fairer comparison between BGS DR9 and MXXL, as the colour distribution in the MXXL simulation does not match perfectly that of the observations, particularly at fainter magnitudes. The results of this exercise are shown in Fig. 13, where the top panel corresponds to galaxies with \( 17 < r_{\text{mag}} < 18 \) and the bottom panel to galaxies with \( 19 < r_{\text{mag}} < 20 \). For each magnitude bin the angular clustering of blue and red galaxies is shown for DR9 BGS (solid) and the MXXL simulation (dotted). As expected, we can see that red galaxies are more strongly clustered than blue ones at intermediate to small angular separations. The overall agreement with the light-cone is good over a large range of angular scales, which thus validates the colour-assignment procedure in MXXL.

3.2 Higher order statistics using counts-in-cells

If the density field is a purely Gaussian random field, then its probability distribution function can be described by just two numbers: the mean and the variance. A Gaussian primordial density field is well supported by observations of the CMB (Planck Collaboration VI 2018). However, one can easily show that, in the gravitational instability scenario, this primordial distribution of density fluctuations will evolve into a distinctly asymmetric density field. Thus, the observed higher order moments of the local density field contain information besides the two-point statistics, such as the departure from Gaussianity, which can inform us about the growth of cosmic structures, and more specifically on the bias between galaxies and the underlying matter distribution (see the review by Bernardeau et al. 2002). Moreover, in order to produce more realistic mock catalogues, it is essential to reproduce the higher order clustering statistics of the BGS sample, especially for regions of high density where spectroscopic incompleteness due to the finite size of the fibre allocation has a significant impact on clustering (Burden et al. 2017; Hahn et al. 2017; Bianchi et al. 2018; Smith et al. 2019; Bianchi & Verde 2020). These higher order statistics can be explored using...
corresponds to a cell size of roughly 0.06 deg which is above the DESI the total number of pixels in the map. In what follows, we consider thus to eliminate the measurement errors. In order to reproduce having only one set of sampling cells to be beaten down and infinite oversampling that enables the noise that is introduced by the HEALPIX pixels distribution are removed while decreasing the HEALPIX resolution partition is comprised of 12 base pixels and the resolution variance, skewness, and kurtosis by making use of the CIC method. We use the HEALPIX\textsuperscript{9} package (Górski et al. 2005) which divides the sky such that each pixel covers the same surface area. This method works for the entire DESI footprint, unlike the one used in Kitamidis et al. (2019), which is based on a transformation of the angular coordinates into Cartesian coordinates. This is a good approximation for regions close to the Galactic plane, such as the rectangles defined in their paper for DECaLS, but is no longer valid when considering BASS/MzLS for instance. In the HEALPIX pixelation, the lowest resolution partition is comprised of 12 base pixels and the resolution increases by dividing each pixel into four new ones such that nside = 2\textsuperscript{resolution} is the number of pixels per side and npix = 12 × nside\textsuperscript{2} is the total number of pixels in the map. In what follows, we consider resolutions above 4 to the maximum 10. The maximum resolution corresponds to a cell size of roughly 0.06 deg which is above the DESI patrol radius. For each resolution of the HEALPIX maps, we remove pixels that are not fully within the survey boundaries by determining a threshold based on the expected number density using the random catalogue. The threshold is determined such that these outliers in the HEALPIX pixels distribution are removed while decreasing the effective area by less than 10 per cent, as confirmed in Fig. 14 which shows the difference in effective area after and before removing the outliers for BASS/MzLS (red), DECaLS-NGC (blue), and DECaLS-SGC (red) based on the random catalogue.

For each resolution of the HEALPIX map, we compute the effective mean density per square degree, the standard deviation, the skewness, and the kurtosis. Szapudi & Colombi (1996) showed that the CIC statistics are sensitive to sample variance (shot noise, edge effects, and finite volume) and to measurements errors due to the finite number of sampling cells. Szapudi (1998) proposed a method of infinite oversampling that enables the noise that is introduced by having only one set of sampling cells to be beaten down and thus to eliminate the measurement errors. In order to reproduce this oversampling effect, we dither by a fraction of cell size each HEALPIX map, compute the CIC statistics for each rotation and take the average. In practise, first we convert the RA, Dec. into x, y, and z coordinates and then rotate the coordinates by an angle $\phi$ (in deg) around an arbitrary rotation axis vector. The angle $\phi$ is randomly chosen in a Gaussian distribution of width the HEALPIX cell size (we also tried twice and five times the HEALPIX cell size). Eventually, we convert back the shifted x, y, z into new RA, Dec.. We do 5 rotations and compute the mean and standard deviation of each quantity above. We did not find any shift in the mean value of each CIC statistics associated with this shifting of pixels, which confirms that we are carrying out a robust sampling and that the tails of the counts distribution are well measured and not unduly affected by the sampling of extreme voids or overdensities. In order to estimate errors, we define a set of 100 jackknife regions, the same set for every pixel size, and we compute the effective density, standard deviation, skewness, and kurtosis in each region, take the mean and the standard deviation. First, we test the procedure using the MXXL light-cone that we split into BASS/MzLS, DECaLS-NGC, and DECaLS-SGC regions. Fig. 15 shows the results of the MXXL light-cone in dashed for each statistic as a function of HEALPIX cell size for BASS/MzLS (blue), DECaLS-NGC (red), and DECaLS-SGC (green) with the coloured regions representing the 1σ errors from the 100 jackknife regions. As expected, the measurements for the different MXXL regions all agree to within the errors. The solid curves show the same results for the BGS DR9 targets with their jackknife errors. The values of the target density for the three imaging surveys are consistent with the ones given in Section 2.1 with 7 per cent difference at maximum when correcting for the magnitude and colour shift between BASS/MzLS and DECaLS. Regarding MXXL, we remind the light-cone was not tuned to fit the BGS number density but it is instead built on the SDSS and GAMA luminosity functions, see Smith et al. (2017). The number density in the observed sample is dominated by photometric variations as explained in Section 2.2. Regarding MXXL, it was not tuned to fit the BGS number density and it is built on SDSS LF.

The other statistics show a better agreement between the three imaging regions of the BGS data and with the MXXL light-cone, even for the third (skewness) and fourth (kurtosis) moments of the galaxy density field although no direct information about these higher order statistics was included in the creation of the MXXL light-cone for BGS. We note that both the skewness and kurtosis are non-zero at small scales and are consistent with zero at large scales as expected from a primordial Gaussian random density field which evolved under gravitational instability and led to the hierarchy of gravitationally bound structures that form the cosmic web with filaments, sheets, knots, and voids.

\footnote{https://healpix.sourceforge.io}

\begin{figure}
\centering
\includegraphics[width=\textwidth]{image.png}
\caption{Ratio of the effective area after and before removing the outliers in the HEALPIX distribution based on the random catalogue for the three imaging surveys: BASS/MzLS (blue), DECaLS-NGC (red), and DECaLS-SGC (green). The shaded grey region shows 10 per cent variation of this fractional area.}
\end{figure}
4 CROSS-CORRELATIONS WITH EXTERNAL SPECTROSCOPIC DATA

In this section, we study the cross-correlations between the BGS targets and external spectroscopic data sets that we present in Section 4.1 in order to determine the real-space projected cross-correlation function $w_p(r_p)$ in Section 4.2 and the redshift distribution $dN/dz$ of the BGS targets in Section 4.3.

4.1 External spectroscopic data sets

In order to study the clustering properties of the BGS targets as a function of redshift, we make use of two external spectroscopic data sets that overlap both in area and redshift with BGS. At $z < 0.2$, in the NGC, we use the SDSS DR7 MGS clustering catalogue (Ross et al. 2015) that covers 6813 deg$^2$ and contains 63 163 SDSS DR7 spectroscopically identified galaxies. This data set overlaps with about 915 deg$^2$ of BASS/MzLS and with about 1410 deg$^2$ of DECaLS-NGC. In both the NGC and SGC, up to $z < 0.6$, we use the LOWZ and CMASS clustering catalogues (Reid et al. 2016) from DR12 of SDSS-III BOSS (Eisenstein et al. 2011; Dawson et al. 2013). The LOWZ sample contains galaxies at $z \leq 0.4$ covering 8337 deg$^2$ and the CMASS sample contains galaxies in $0.4 \leq z \leq 0.7$ covering 9376 deg$^2$. In total, the overlap with BASS/MzLS corresponds to roughly 2965 deg$^2$, the one with DECaLS-NGC to 4335 deg$^2$ and the one with DECaLS-SGC to 2585 deg$^2$. Fig. 16 shows the footprint of the DR9 Legacy Imaging Surveys restricted to the DESI footprint, together with the footprint of SDSS MGS and BOSS LOWZ-CMASS.

4.2 Clustering as a function of redshift

By making use of cross-correlation with samples of known redshift, we can infer the projected real-space cross-correlation function, $w_p(r_p)$, which corresponds to the integral of the 2D spatial correlation function $\xi(r_1, r_2)$ over the line-of-sight direction $r_1$ (Peebles 1980). This projected real-space correlation function can be related to the angular correlation function $w(\theta)$. Following Padmanabhan et al. (2009), if we consider that the spectroscopic sample lies at $\chi = \chi_0$, then according to the flat-sky approximation and by applying the Limber approximation, the relation simplifies to:

$$w(\theta) = f(\chi_0)w_p(r_p),$$

where $f(\chi_0)$ is the normalized radial distribution of the photometric sample. We can generalize this to a narrow spectroscopic redshift slice where we consider the clustering as constant over the redshift slice and we can assume that the redshift of the photometric galaxy is the same as the one of the spectroscopic galaxy it is correlated with. It allows us to rebin the pair counts of the angular correlation function in transverse separation, $w_\theta(r_p)$. We can thus rewrite equation 9 as:

$$w_\theta(r_p) = \langle f(\chi) \rangle w_p(r_p),$$

where $f(\chi)$ is the normalized radial distribution of the photometric sample over the spectroscopic bin considered. In what follows, we use the MXXL light-cone simulation to estimate the contribution from the redshift distribution of the BGS targets. Recently, this observable, the projected cross-correlation function binned into transverse comoving radius, was also developed for improving the measurement of the Baryon Acoustic Oscillations feature for a sparse spectroscopic sample (Patej & Eisenstein 2018). The first detailed application to data was proposed in Zarrouk et al. (2021) where a...
sample of sparse quasars from eBOSS was cross-correlated with a dense photometric sample of galaxies using the DR8 DESI Legacy Imaging Surveys.

We use the Landy–Szalay estimator as defined by equation (4) where we count pairs in 16 logarithmically spaced angular bins between \( \theta = 0.001 \) and \( \theta = 1 \) deg. We divide the external spectroscopic data sets into redshift slices of width \( \Delta z = 0.1 \), and each redshift slice is cross-correlated with the BGS catalogue. We measure the angular cross-correlation function binned into transverse distances, as expected the amplitude of the signal is higher when cross-correlating with the redshift slices closest to the peak redshift distribution of the BGS which is for \( 0.2 < z < 0.3 \). Then we divide by the normalized radial distribution of the BGS sample in each redshift slice using the MXXL light-cone. Errors bars are estimated using the bootstrap technique presented in Kitanidis et al. (2019), using 500 bootstrap realizations. We checked that our results are robust to the choice of the number of bootstrap realizations. We can also compare the projected real-space cross-correlation function with theory by assuming a power-law model for the correlation function, which gives:

\[
 w_p(r_p) = r_p^{1-\gamma} \sqrt{\pi} \frac{\Gamma(\gamma/2 - 1/2)}{\Gamma(\gamma/2)}.
\]  

Fig. 17 shows the real-space projected cross-correlation function of the BGS targets in BASS/MzLS (top), DECaLS-NGC (middle), and DECaLS-SGC (bottom). The colours represent the cross-correlation functions with different redshift slices of the external spectroscopic data sets (SDSS MGS DR7 for BASS/MzLS and DECaLS-NGC, BOSS LOWZ-CMASS for all three DECaLS regions). The dashed black line shows the theoretical prediction when assuming a power law for the correlation function as given by equation (11) with the clustering length \( n_0 \) and slope \( \gamma \) corresponding to the best-fitting parameters from Table 2 for each imaging region. We thus obtain a consistent behaviour between the angular correlation function, \( w(\theta) \), and the projected real-space cross-correlation function, \( w_p(r_p) \). Fig. 17 also shows a slight decrease of the clustering amplitude with increasing redshift which is in competition between the redshift evolution of the absolute magnitude – as redshift increases, flux-limited samples of galaxies, such as BGS, are intrinsically brighter – and the redshift evolution of the clustering amplitude – as redshift increases, the clustering amplitude decreases. Using the formula for the luminosity dependence of the clustering of Norberg et al. (2001) for the extreme slices where the peak luminosity shifts from \(-18\) in the lowest redshift bin and to \(-22\) in the highest redshift bin, we found a 0.73 reduction in clustering amplitude moving from high to low redshift. At the same time, the growth factor at these two redshifts changes the clustering amplitude by 1.51, which gives an overall 10 per cent effect on the clustering amplitude, i.e. \( 1.51 \times 0.73 = 1.10 \).

4.3 Clustering-redshift \( dN/dz \)

The cross-correlation signal between a photometric sample and spectroscopic data sets can also be used to infer the redshift distribution of the photometric sample. Using cross-correlations to get information about the properties of the photometric galaxies is not new. For instance Seldner & Peebles (1979) used cross-correlations of quasars and galaxy counts to test how galaxies are distributed around quasars and to see if this depends on the redshift and apparent magnitude of the quasar. Schneider et al. (2006) first discussed the idea of using a spectroscopic sample to determine the redshift distribution of photometric galaxies, then Newman (2008) expanded on this and described the clustering-redshift technique. We follow the approach proposed by Ménard et al. (2013) and Rahman et al. (2015) which uses smaller scales and accounts for the redshift evolution of the galaxy bias. We measure the angular cross-correlation function between the BGS targets and narrow redshift slices of the spectroscopic data sets (\( \Delta z = 0.05 \)). For each redshift slice \( \Delta z \), we compute the angular cross-correlation function \( w_{\theta}(\theta, \Delta z) \), then we integrate over an annulus centered on the spectroscopic object from \( \theta_{\min} \) to \( \theta_{\max} \). Indeed, Ménard et al. (2013) showed that the sensitivity of the cross-correlation signal is improved when including information from many clustering scales. In order to maximize the signal-to-noise ratio, we can weight each measurement by \( \theta^{-1} \). Applying this weighting gives the following...
cross-correlation function using a function of redshift. We use CMASS, we use the large-scale bias measurements of Ross et al. spectroscopic sample. For SDSS MGS DR7 and BOSS LOWZ-CMASS galaxies (<z< 0.07) spectroscopic pipeline. we note that the Survey Validation is currently providing accurate additional check of the clustering evolution with redshift. Moreover, technique but rather to show an illustration of it and provide an is not to provide the most accurate redshift distribution using this BGS targets, we use the DESI Final Design Report values (DESI of the bias is negligible over the redshift slice respectively. For both spectroscopic samples the redshift evolution where we set θ_{min}, θ_{max} to match a fixed range of projected radii r_{p,min}, r_{p,max} so that we can probe the same range of physical scales as a function of redshift. We use r_{p,min} = 0.05 and r_{p,max} = 5 h^{-1} Mpc.

Finally, the redshift distribution can be inferred from the integrated cross-correlation function using:

\[ \frac{dN}{dz} \propto \frac{\theta_w(z)}{b_g(z) b_b(z)}, \]  

where \( b_g \) is the bias of the BGS sample and \( b_b \) is the bias of the spectroscopic sample. For SDSS MGS DR7 and BOSS LOWZ-CMASS, we use the large-scale bias measurements of Ross et al. (2015), Howlett et al. (2015), and of Rodríguez-Torres et al. (2016), respectively. For both spectroscopic samples the redshift evolution of the bias is negligible over the redshift slice \( \Delta z = 0.05 \). For the BGS targets, we use the DESI Final Design Report values (DESI Collaboration 2016a) and we find consistency when comparing with the MXXL BGS light-cone. However, we highlight that the aim is not to provide the most accurate redshift distribution using this technique but rather to show an illustration of it and provide an additional check of the clustering evolution with redshift. Moreover, we note that the Survey Validation is currently providing accurate redshifts for building truth tables and testing the performance of the spectroscopic pipeline.

Fig. 18 shows the redshift distribution \( dN/dz \) of the BGS targets in BASS/MzLS, DECaLS-NGC, and DECaLS-SGC using this clustering-redshift technique with SDSS MGS DR7 galaxies in 0.07 < z < 0.2 and BOSS LOWZ-CMASS galaxies 0.2 < z < 0.6. We confirm that the inferred redshift distributions are all in agreement between the three imaging regions, and also with the redshift distribution of the BGS targets that match GAMA galaxies using their spectroscopic redshifts (in black).

5 CONCLUSIONS

We have presented the target selection pipeline that selects the BGS bright targets \( r_{mag} \leq 19.5 \) from the latest release of the Legacy Imaging Surveys (DR9) that uses BASS, MzLS, and DECaLS over the DESI footprint of 14000 deg². This includes several changes with respect to what was first presented using DECaLS DR8 in Ruiz-Macias et al. (2021) that we summarize here:

(i) Thanks to major improvements in the iterative fitting around bright stars, the radius of the bright star mask is half the size in DR9 than it was in DR8. We have checked that this change does not introduce any spurious effects. We found that our star–galaxy separation based on GAIA yields less than 2 per cent stellar contamination which is also confirmed by the cross-correlation signal of BGS targets and bright stars at large scales.

(ii) Thanks to major improvements in the photometry, we no longer need to apply a spatial masking around large galaxies: we were missing true BGS targets in the vicinity of the large galaxies that we are now able to recover. We checked that this change improves the completeness with respect to GAMA which is highly complete in 14 < r < 19. We also developed a visual inspection webtool which confirms that not applying the large galaxy mask does not introduce a significant fraction of spurious objects.

(iii) Using this visual inspection webtool, we also found that we could make a less conservative choice on some of the quality cuts that involve FRACIN, FRACMASKED, and FRACFLUX which increases the completeness slightly while ensuring a negligible fraction of spurious objects in the vicinity of the large galaxies.

(iv) Finally, the DR9 selection cuts yield a completeness that is above 99 per cent with respect to GAMA DR4, which is complete in 14 < r < 19. DR9 also results in a BGS bright sample \( r \leq 19.5 \) that meets the requirement for target density which is above 800 deg⁻² for the three imaging surveys.

After presenting the target selection cuts and studying the main systematics, we characterized the clustering properties of the BGS bright sample by looking at several statistics:

(i) The angular clustering shows very good consistency between the three imaging surveys and with the MXXL BGS light-cone (Smith et al. 2017). It is also consistent with a power-law model on angular scales below 1 deg and it gives comparable clustering strength \( r_0 \) and slope \( \gamma \), both between the three imaging surveys, with the mock and with previous measurements in the literature.

(ii) The colour-dependent clustering confirms that brighter and redder galaxies are more strongly clustered as red galaxies tend to be more luminous than blue ones. We also found a very good agreement with MXXL when splitting both samples by magnitude and fixed fraction of blue/red galaxies.

(iii) We also investigated higher order clustering properties by making use of the counts-in-cells technique using the HEALPIX package. We compared the mean density, standard deviation, skewness, and kurtosis of the galaxy density field in BASS/MzLS, DECaLS-NGC, DECaLS-SGC for the data and MXXL.

(iv) Finally, we looked at the redshift evolution of the clustering by cross-correlating the BGS targets with external spectroscopic data sets that overlap in area and redshift. We used SDSS MGS DR7 and BOSS LOWZ-CMASS to compute the projected real-space cross-correlation function in redshift slices of width \( \Delta z = 0.1 \). We found very good agreement and a consistent clustering with the prediction from a power-law model using the best-fitting values for \( r_0 \) and \( \gamma \) that we measured by fitting the angular clustering. We also inferred the redshift distribution of the BGS targets using the clustering-redshift technique. We were able to recover a redshift distribution that is consistent between the three imaging surveys and with the one we
estimated by matching the BGS targets with GAMA galaxies and by using their spectroscopic redshifts.

As a summary, this work validates the selection for the BGS bright sample which will be very close to the final one. We have characterized and studied several clustering statistics that provide the first large-scale angular clustering analysis of the DESI BGS. Modelling the sample is an important first step for doing cosmology with DESI, and our clustering results will also aid in the creation and validation of accurate mock catalogues just as the ones we are currently developing with the AbacusSummit simulations (Maksimova et al., in preparation) which uses the ABACUS code (Metchnik 2009; Garrison et al. 2018). However, additional work is required using the spectroscopic data of the ongoing Survey Validation period in order to investigate (1) the spectroscopic completeness, (2) the fiber assignment efficiency and impact on the clustering, and (3) the 3D clustering. Moreover, the final target selection will also include a fainter sample that will be fully characterized in a future work.

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DATA AVAILABILITY

The DESI Legacy Imaging Surveys used for this work is public at https://www.legacysurvey.org/dr9/description/.

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