Planes of satellite galaxies: when exceptions are the rule

Marius Cautun1*, Sownak Bose1, Carlos S. Frenk1, Qi Guo2, Jiaxin Han1, Wojciech A. Hellwing1,3, Till Sawala1 and Wenting Wang1

1 Institute for Computational Cosmology, Department of Physics, Durham University, South Road, Durham DH1 3LE, UK
2 Key Laboratory for Computational Astrophysics, The Partner Group of Max Planck Institute for Astrophysics, National Astronomical Observatories, Chinese Academy of Sciences, Beijing, 100012, China
3 Interdisciplinary Centre for Mathematical and Computational Modelling, University of Warsaw, ul. Pawińskiego 5a, Warsaw, Poland

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ABSTRACT
The detection of planar structures within the satellite systems of both the Milky Way (MW) and Andromeda (M31) has been reported as being in stark contradiction to the predictions of the standard cosmological model (ΛCDM). Given the ambiguity in defining a planar configuration, it is unclear how to interpret the low incidence of the MW and M31 planes in ΛCDM. We investigate the prevalence of satellite planes around galactic mass haloes identified in high resolution cosmological simulations. We find that planar structures are very common, and that ∼10% of ΛCDM haloes have even more prominent planes than those present in the Local Group. While ubiquitous, the planes of satellite galaxies show a large diversity in their properties. This precludes using one or two systems as small scale probes of cosmology, since a large sample of satellite systems is needed to obtain a good measure of the object-to-object variation. This very diversity has been misinterpreted as a discrepancy between the satellite planes observed in the Local Group and ΛCDM predictions. In fact, ∼10% of ΛCDM galactic haloes have planes of satellites that are as infrequent as the MW and M31 planes. The look-elsewhere effect plays an important role in assessing the detection significance of satellite planes and accounting for it leads to overestimating the significance level by a factor of 30 and 100 for the MW and M31 systems, respectively.

Key words: galaxies: haloes - galaxies: abundances - galaxies: statistics - dark matter

1 INTRODUCTION
While the Universe at large may be homogeneous and isotropic, on galactic scale the distribution of galaxies is highly anisotropic. This is most readily seen in the spatial and kinematical distribution of the Local Group (LG) satellites. In the MW, the 11 “classical” satellites define a thin plane (Lynden-Bell 1976) and some of the fainter satellites, tidal streams and young globular clusters have an anisotropic distribution reminiscent of this plane (Metz, Kroupa & Jerjen 2009; Pawłowski, Pfennig-Altenburg & Kroupa 2012). Many members of this “disk of satellites” have a common rotation direction and it has been claimed that the plane is a rotationally stabilized structure (Metz, Kroupa & Libeskind 2008; Pawłowski & Kroupa 2013). Similarly, the spatial distribution of satellites around M31 is anisotropic (Koch & Grebel 2006; McConnachie & Irwin 2006), with 15 out of 27 satellites observed by the Pan-Andromeda Archaeological Survey (PAndAS; McConnachie et al. 2009) located in a thin plane. Out of the 15 members of the plane, 13 of them share the same sense of rotation (Ibata et al. 2013, hereafter Ibata13).

Anisotropies in the distribution of satellites are a clear prediction of the Λ cold dark matter (ΛCDM) paradigm (Libeskind et al. 2005; Zentner et al. 2005; Libeskind et al. 2009, 2011; Deason et al. 2011; Wang, Frenk & Cooper 2013; Sawala et al. 2014). Such flattened satellite distributions, dubbed “great pancakes”, can arise from the infall of satellites along the spine of filaments (Libeskind et al. 2005; Buck, Macci’o & Dutton 2015), which in turn determine the preferential points at which satellites enter the virial radius of the host halo (Libeskind et al. 2011, 2014). The imprint of anisotropic accretion is retained in the dynamics of satellites, with a significant population co-rotating with the spin of the host halo (Libeskind et al. 2009; Lovell et al. 2011; Cautun et al. 2015).

Although flattened satellite distributions are common in ΛCDM, configurations similar to those of the MW and...
M31 are infrequent. Wang, Frenk & Cooper (2013) found that 5 – 10% of satellite systems are as flat as the MW’s 11 classical satellites. When it is required that the velocities of at least 8 of the 11 satellites should point within the narrow angle claimed by Pawlowski & Kroupa (2013) for the MW satellites, this fraction decreases to ~0.1% (Pawlowski et al. 2014). In the case of the M31 thin satellite plane, Bahl & Baumgardt (2014) found that, while similar spatial distributions of satellites are quite common in ΛCDM, there is only a 2% chance that 13 out of the 15 members in the plane would share the same sense of rotation. In similar studies, Ibata et al. (2014c) and Pawlowski et al. (2014) found an even lower occurrence for the M31 plane, with only ~0.1% of ΛCDM systems having a similar configuration.

Extending the above analysis to galaxies outside the LG is constrained by observational limitations, but some additional tests can be performed. Cautun et al. (2015) studied the flattening of the satellite distribution around isolated central galaxies in the SDSS, as viewed on the plane of the sky, finding good agreement between data and cosmological simulations. Using a similar approach, Ibata et al. (2014b) claimed a higher degree of flattening in their data, but their conclusions may be affected by systematics (see Cautun et al. 2015, for a discussion of this study). The expected signature of planar rotation has been investigated by considering the velocity correlation of satellite pairs observed on opposite sides of the host galaxy. Initially, using a sample of 23 systems, Ibata et al. (2014a) reported a significant excess, when compared to ΛCDM predictions. Cautun et al. (2015, see also Phillips et al. 2015) found that this excess decreases rapidly as the sample size is increased and also that the expected mirror image signal is absent for satellite pairs on the same side of the host galaxy, suggesting that the claimed excess is not robust.

In this study, we examine planar configurations of satellites identified in MW- and M31-like mock ΛCDM catalogues and compare them to the planar structures observed in the LG. Among others, we revisit the claims by Ibata et al. (2014c) and Pawlowski et al. (2014) that the two planes of satellites found in the LG are inconsistent with ΛCDM predictions. Along with similar studies, those works are based on two important axioms: that the majority of planar configurations are the same, and that the planes detected around the MW and M31 are representative of planar structures in general. We will examine these two conjectures within the context of ΛCDM predictions. Starting from high resolution cosmological simulations, we create mock catalogues that are used to identify planar satellite configurations similar to those found in the LG. We proceed to study the properties of the most prominent planes of satellites and to compare those ΛCDM predictions with the two planes of satellites observed in MW and M31.

This paper is organized as follows. In Sec. 2 we introduce the cosmological simulations as well as the selection criteria used to identify MW and M31 analogue systems; in Sec. 3 we present an objective method to identify spatially and kinematically rare planes; in Sec. 4 we identify planes of satellites for PAndAS-like mocks; in Sec. 5 we analyse MW-like systems; we conclude with a short discussion and summary in Sec. 6.

2 DATA AND SAMPLE SELECTION

This work makes use of two high-resolution ΛCDM DM-only cosmological simulations: the Millennium-II (MS-II; Boylan-Kolchin et al. 2009) and the Copernicus Complexio (COCO; Hellwing et al. 2015) simulations. Instead of using the original MS-II, which was run assuming a Wilkinson Microwave Anisotropy Probe (WMAP)-1 cosmogony, we employ a modified version of the simulation that has been rescaled to the WMAP-7 cosmology (Ω_m = 0.272, Ω_λ = 0.728, σ_8 = 0.81 and n_s = 0.968) using the scaling algorithm of Angulo & White (2010, see also Guo et al. 2013). The rescaled MS-II corresponds to a simulation in a periodic box of side-length 104.3 h^{-1}Mpc containing 2160^3 DM particles, each particle having a mass, m_p = 8.50 \times 10^6 h^{-1}M_\odot.

COCO simulates a smaller, roughly spherical, cosmological volume, V = 2.25 \times 10^4 (h^{-1}Mpc)^3, equivalent to a cubic box of side-length 28.2 h^{-1}Mpc, but at much higher resolution than MS-II, having 2374^3 particles each with mass, m_p = 1.14 \times 10^5 h^{-1}M_\odot, and a gravitational softening length, ε = 0.23 h^{-1}kpc. This volume is embedded within a larger periodic box, of side-length 70.4 h^{-1}Mpc, that is resolved at a significantly lower resolution (see Hellwing et al. 2015, for more details). COCO uses the same WMAP-7 cosmological parameters as the rescaled MS-II.

To construct mock catalogues, we use the semi-analytic galaxy formation model of Guo et al. (2011) that has been implemented for the rescaled version of MS-II in Guo et al. (2013) and for COCO in Guo et al. (2015). The semi-analytic model has been calibrated to reproduce the stellar mass, luminosity and autocorrelation functions of low redshift galaxies as inferred from SDSS. The abundance and radial distribution of satellite galaxies predicted by the model are in good agreement with SDSS data (Wang & White 2012; Wang et al. 2014), which makes the Guo et al. model a good test bed for studying planar structures of satellites.

Due to the relatively low resolution of MS-II, many of the satellite galaxies of interest for this work are found in haloes close to the resolution limit of \sim 10^9 h^{-1}M_\odot, which raises questions about the accuracy of the properties and orbits of these objects, especially after infall into the main halo. To test for any potential systematic effects arising from the limited resolution of MS-II, we compare with the results of COCO. Any such effects are significantly reduced or even absent in COCO, since it has 75 times higher mass resolution and four times better spatial resolution.

We select counterparts to the two massive members of the Local Group by identifying DM haloes with similar masses in the range (1 – 3) \times 10^{12} M_\odot, which is consistent with the mass of the MW and M31 halo (e.g. Fardal et al. 2013; Piff et al. 2014; Cautun et al. 2014b,a; Veljanoski et al. 2014; González, Kravtsov & Gnedin 2014, for a compilation of other measurements and discussions of systematic effects see Courteau et al. 2014; Wang et al. 2015). This results in 2849 MS-II haloes and in 63 COCO haloes in the required mass range. Compared to the previous studies that analysed the incidence of the M31 plane of satellites (e.g. Bahl & Baumgardt 2014; Pawlowski et al. 2014), we adopted a broader mass range to account for the large uncertainty in the mass measurements and also for possible systematic effects. We checked that the exact mass range used does not affect our final results.
3 IDENTIFYING PLANAR CONFIGURATIONS OF SATELLITES

In this section we introduce an objective method to identify, for each halo, the rarest plane of satellites, both spatially and kinematically. The method works by identifying the subsample of satellites that is the least likely to be obtained by chance. This is motivated by recent observations that have shown that only a subset of the satellite galaxies are potentially distributed along a plane. For example, out of the 27 M31 satellites in the PAndAS footprint, a significant plane is found for 15 of them (Ibata13), while the entire population is no more planar than would be expected for an isotropic distribution of equal size (Conn et al. 2013). Tully et al. (2015) found that the Centaurus A Group shows evidence for two planes of galaxies, which, between them, contain 27 out of the 29 members with known distances. Even for the MW, while the 11 classical satellites are found on a plane, only 8 have orbital poles in a narrow angle indicating a possibly long lived planar structure (Pawłowski & Kroupa 2013).

Previous methods of identifying planar configurations have been based on first examining the data and only in a second step defining an approach for selecting planes, resulting in a selection method that is both subjective and a posteriori. For example, the 15 members of the M31 plane have been found by noticing that, when increasing the number of satellites associated with a plane, the best-fitting plane hardly changes and that the thickness of the plane increases only slowly. If one considers more than 15 satellites, this then leads to a larger change in the thickness of the best-fitting plane and in its direction (Ibata13). Applying such a method to a large sample of systems is undesirable, since it implies choosing at least two threshold parameters: the maximum allowed changes in the thickness of a plane and in its direction when adding an extra valid plane member. There are no a priori well-motivated values for those threshold parameters. In addition, the two thresholds should likely depend on the radial distribution of satellites, since more radially centered distributions will likely have thinner planes. While other methods of identifying planar distributions have been proposed (e.g. Conn et al. 2013; Gillet et al. 2015), they all involve one or more subjective parameters. In contrast, the method we present here does not involve any such parameters and, in addition, it naturally takes into account the radial distribution of satellites in each system.

3.1 Spatial planes

We start by introducing a method for identifying the most prominent spatial plane. When comparing planes that contain the same number of satellites, the one that stands out the most is the thinnest plane. Difficulties arise when we have to compare planes with two different numbers of members, since it is not trivial to decide which one stands out more. In a nutshell, we identify all possible subsamples of satellites, out of a total sample of $N_{\text{max}}$ satellites, and, of those, we select the one configuration that is the least likely to be caused by statistical fluctuations. This is illustrated in Fig. 1.

In a first step, we identify the satellite subsets that are of interest for our study. We do so using the approach detailed in Appendix A1. For each such subsample, we find the best-fitting plane, which is the plane that minimizes the root-mean-square distance of the satellites from it. For this, we define the plane thickness, $r_\perp$, as

$$r_\perp = \left( \frac{\sum_{i=1}^{N_{\text{sat}}} (\mathbf{n}_{\text{plane}} \cdot \mathbf{x}_i)^2}{N_{\text{sat}}} \right)^{1/2},$$

where $N_{\text{sat}}$ is the number of satellites in the subsample and $\mathbf{n}_{\text{plane}}$ denotes the normal to the plane. With $\mathbf{x}_i$ we denote the position of each satellite in a coordinate system whose origin is the central host galaxy. The plane thickness, $r_\perp$, is in fact the dispersion in the distance of the satellites from a plane that goes through the central galaxy. The best-fitting plane is the one that minimizes $r_\perp$. The normal to this plane,
n_{plane}, is given by the eigenvector corresponding to the lowest eigenvalue of the inertia tensor of its members.

Each resulting plane is characterized in terms of its prominence, P, such that, the larger the prominence, the least likely it is that the plane is due to a chance alignment. For example, for plane \( i \) that has \( N_{sat; i} \) members and a thickness, \( r_{\perp; i} \), the spatial prominence is defined as:

\[
P_{\text{spatial}}^{plane} = \frac{1}{p(\leq r_{\perp; i} | N_{sat; i})}
\]

where the denominator gives the probability of obtaining by chance a configuration of \( N_{sat; i} \) satellites that is thinner than \( r_{\perp; i} \). This probability is computed using \( 10^9 \) isotropic distributions of satellites as outlined in Appendix A3. Since the radial distribution of satellites has a strong effect on the thickness of the resulting planes, we generate each isotropic realization to have the exact same radial distribution as the system under study.

Now it is only natural to characterize the most prominent, or rarest, plane as the one that is the least likely to be obtained by chance. Using our notations, this can be written formally as:

\[
P_{\text{rarest}}^{spatial} = \max_{\text{all planes } i} \left[ P_{\text{spatial}}^{plane; i} \right],
\]

which says that the rarest spatial plane is the one that has the largest prominence. It is important to note that, within this approach, every halo contains a rarest plane. Determining if this rarest plane is statistically significant is a separate question that we will address in Sec. 4.

### 3.2 Spatial and 2D-kinematic planes

The observational data for the MW and M31 satellites contains both positions and velocity information for these objects. It is natural to try to incorporate this additional velocity information into the detection of planar configurations of satellites. The M31 satellites have only radial velocity measurements, so the full 3D velocities are unknown. But since the M31 plane of satellites is almost parallel to the line of sight, the radial velocities can be used to estimate the sense of rotation of each satellite with respect to the best-fitting plane. In the following, we describe how to select spatial + 2D-kinematic planes, which are at the same time spatially thin and have a large number of members that share the same sense of rotation.

Before continuing, it is important to discuss some potentially misleading nomenclature used in previous studies. Satellites sharing the same sense of rotation have been referred to as corotating satellites (e.g. Ibata13, Bahl & Baumgardt 2014). This nomenclature is confusing since corotation is normally used to denote a rotation around a common axis. Thus, two satellites corotate if their orbital poles are very close together. In the absence of 3D velocities, we only know that, when projected on the best-fitting plane, 13 out of the 15 satellites rotate in the same sense, either clockwise or counter-clockwise.

In addition to the steps described in Sec. 3.1, for each satellite subset we also determine the number of members that share the same sense of rotation relative to the best-fitting plane. To determine the direction of rotation of each member, we take the scalar product between the plane normal and the orbital momentum of the satellite. A positive scalar product corresponds to clockwise rotation, while a negative one corresponds to counter-clockwise rotation.

The number of satellites sharing the same sense of rotation, \( N_{s.s.r.; i} \), is the maximum of the number of objects rotating clockwise and those rotating counter-clockwise. Following this step, we assign to each plane a 2D-kinematic prominence, \( P_{2D-kin} \), which is defined as:

\[
P_{2D-kin}^{plane} = \frac{1}{p(\geq N_{s.s.r.; i} | N_{sat; i})},
\]

where the denominator gives the probability of obtaining by chance a configuration of \( N_{sat; i} \) satellites in which at least \( N_{s.s.r.; i} \) of them share the same sense of rotation. The procedure for computing this probability is detailed in Appendix A3.

We define the rarest spatial + 2D-kinematic planes as the one whose spatial and 2D kinematical distribution are the least consistent with a statistical fluctuation. Thus,

\[
P_{\text{rarest } spatial + 2D-kin} = \max_{\text{all planes } i} \left[ P_{\text{spatial}}^{plane; i} \times P_{\text{2D-kin}}^{plane; i} \right],
\]

that is the plane that maximizes the product of the spatial and the 2D-kinematic prominences.

### 3.3 Spatial and 3D-kinematic planes

In the case of the MW, the 3D velocities of the 11 classical satellites are known. This suggests that for the MW system one can identify planes that are both spatially thin and show a large degree of coherent 3D kinematics. For this, one needs to construct a cost function that rewards systems in which most satellites have orbital poles close together and penalizes systems in which the orbital poles are isotropically distributed. For example, to study long lived planar configurations, the cost function would preferentially reward systems in which the orbital momentum of its members is close to parallel or anti-parallel to the normal to the best-fitting plane. For this work, we employ a variant of the cost function suggested by Pawlowski & Kroupa (2013), since this one has been used in other studies that claim a tension between the MW satellite plane and ΛCDM predictions (e.g. Pawlowski et al. 2014). That function has been proposed after examining the orbital data of the MW satellites, and as such it is an a posteriori definition of the rotation characteristics of the Galactic satellite distribution. It may be that other satellite planes in the Universe have different orbital structures, in which case that cost function may not be optimally suited for characterizing their kinematical structure.

To compute the amount of kinematical information, we proceed as follows. For each of the satellite subsets used in Sec. 3.1, we determine the dispersion in the orbital poles, i.e. directions of the orbital momenta, of its members as

\[
\Delta_{std} = \sqrt{\frac{\sum_{i=1}^{N_{sat}} \arccos^2 (\mathbf{n}_{\text{orbit}} \cdot \mathbf{n}_{\text{orbit}; i})}{N_{sat}}}
\]

where \( \mathbf{n}_{\text{orbit}; i} \) denotes the orbital momentum direction of each member of the plane and \( \mathbf{n}_{\text{orbit}} \) the mean orbital pole of all the \( N_{sat} \) members found in the plane. Compared to our approach, Pawlowski et al. (2014) applied Eq. (6) to only 8 out of the 11 satellites found in the MW satellite plane. This
was motivated by the observation that only 8 out of the 11 members show close orbital poles\(^1\). In principle, we could follow a similar approach and take a subset of the plane members that shows the most concentrated orbital poles. This would amount to taking a subset of a subset, since our planes are already subsets of satellites from a larger sample. We prefer not to do so since it would add an additional layer of complexity to this method and also a significantly higher computational cost.

After applying Eq. (6) to each plane, we define the 3D-kinematic prominence, \(p_{\text{3D-kin}}\), of plane \(i\) as

\[
p_{\text{3D-kin}} = \frac{1}{p(\leq \Delta_{\text{std}, i} \mid N_{\text{sat}, i})},
\]

where \(\Delta_{\text{std}, i}\) denotes the orbital dispersion of the plane. The denominator gives the probability of obtaining by chance a configuration of \(N_{\text{sat}, i}\) that have an orbital pole dispersion less than \(\Delta_{\text{std}, i}\). The procedure for computing this probability is given in Appendix A3.

Now we can define the prominence of the rarest \(3\) spatial + 3D-kinematic plane as

\[
p_{\text{rarest}}^{\text{spatial + 3D-kin}} = \max_{\text{all planes } i} \left[ p_{\text{plane } i}^{3\textD-kin} \times p_{\text{plane } i}^{\text{spatial}} \right],
\]

which is the plane of satellites whose spatial and 3D-kinematic distribution is the least likely to be a statistical fluctuation.

\(^1\) Incorporating such a posteriori considerations incurs to designing tests that are specifically matched to one particular system and that may not be characteristic of the larger population.
we end up with 7350 mock satellite systems in MS-II and 180 in COCO.

4.1 The M31 system

In a first step, we apply our method to the actual PAndAS observations of M31. We do not use the same plane identification method as Ibata13, so it is important to check what it is that our approach identifies as the most prominent plane of the M31 system. To account for observational errors, we generate 1000 Monte Carlo realizations that sample the radial distance PDFs (Conn et al. 2012, Table 1) and radial velocities (Collins et al. 2013, Table 5) of the M31 satellites.

In the case of the M31 system, the observational data allow for the identification of spatial and spatial + 2D-kinematic planes. For each Monte Carlo realization of the M31 system, we identify the rarest plane. Due to the large radial distance errors, the rarest plane can vary between realizations. For example, the rarest spatial plane contains 14 members in 72% of the cases and 13 members in 22% of realizations. In the remaining 6% of realizations it contains even fewer satellites. For simplicity, we take the rarest plane as the one that is identified as such in the largest number of realizations. The rarest planar configurations of the M31 system and its characteristics are shown in Table 1.

We find that the rarest spatial plane consists of 14 satellites that are the same as the 15 members of the plane reported by Ibata13, except And III. This is in agreement with the results of Ibata13, who point out that choosing 13 or 14 satellites results in a higher spatial significance, i.e. lower probability of being obtained from an isotropic distribution, than for the full sample of 15. The spatial + 2D-kinematic plane found by our approach is the same as the one reported by Ibata13, even though our plane identification method is different. Ibata13 reported the significance of the M31 plane as compared to an isotropic distribution, so it is possible that they inadvertently choose the parameters of their method (see Sec. 3) such that it maximizes the plane significance. If that was the case, then both plane finding methods are basically the same.

4.2 The rarest M31-like planes

To better understand the M31 plane of satellites, we start by assessing the chance of obtaining more prominent plane configurations within ΛCDM. This is shown in Fig. 2, where we plot the cumulative distribution function (CDF) of the prominence, $\mathcal{P}_{\text{rarest}}^{\text{spatial}}$, of the rarest spatial planes. There is a very good match between the MS-II and COCO haloes, which suggests that satellite planes found in MS-II are not significantly affected by the limited resolution of the simulation. We find that most haloes have prominent plane configurations, for example 37, 12 and 4% of haloes have planes with $\mathcal{P}_{\text{rarest}}^{\text{spatial}} \geq 10^2$, $10^3$ and $10^4$, respectively.

The prominence is not simply the inverse of the probability for isotropic satellite distributions for reasons that will be discussed in Sec. 4.3. As such, the figure also shows the rarest planes found in an isotropic distribution of satellites. To obtain these, for each of the ΛCDM haloes, we generated an isotropic distribution by choosing random polar and azimuthal angles, while keeping the radial position of each satellite fixed. We then apply the same plane identification procedure to each isotropic satellite distribution. Unsurprisingly, we find a clear difference between the isotropic and ΛCDM results, with the isotropic CDF shifted towards the left of the ΛCDM CDF. This suggests that, compared to a uniform distribution, there is more structure in the distribution of ΛCDM satellites, in agreement with previous studies (e.g. Libeskind et al. 2005; Wang, Frenk & Cooper 2013; Pawłowski et al. 2014).

The corresponding M31 plane, entry (1) from Table 1, is shown as the dashed vertical line. It has a prominence, $\mathcal{P}_{\text{rarest}}^{\text{spatial}} = 1.0^{+1.1}_{-0.5} \times 10^3$, which means that for an isotropic distribution there is only a 1 in 1000 chance of obtaining
a thinner plane with 14 members. This result was found by computing the prominence of the rarest plane for each Monte Carlo realization of the M31 system. Following this, we quote the median value and the 1σ interval, corresponding to the 16th to 84th percentiles. We find that the spatial plane of the M31 system is consistent with ΛCDM expectations, since there is a 12.4% chance of finding an even more prominent plane in ΛCDM. In fact, the M31 spatial plane is consistent, at 2.5σ, even with an isotropic distribution, since there are 1.2+1.0−0.9% more extreme systems in this case (these results are summarized in Table 1).

Fig. 3 shows the prominence of the rarest spatial + 2D-kinematic planes. The general conclusions are the same as for the previous figure: we find a very good match between the MS-II and the COCO results; and ΛCDM satellite distributions have more prominent planes than isotropic distributions. The bumpy aspect of the CDF curves for the MS-II and the isotropic case reflects the discrete nature of the 2D-kinematical test, since the number of plane members sharing the same sense of rotation always takes integer values. For this test, the corresponding M31 plane, entry (2) from Table 1, which is the one identified by Ibata13, is characterized by a prominence, $P_{\text{spatial} + 2\text{D-kin}} = 3.4_{-1.6}^{+2.8} \times 10^4$. There are $8.8_{-1.6}^{+2.8}$ ΛCDM haloes with more prominent planes suggesting that the M31 plane of satellites is in agreement with ΛCDM predictions.

### 4.3 The detection significance of a plane

We now discuss the detection significance of a plane of satellites, i.e. the probability that a plane is due to a statistical fluctuation. For this, we need to take into account the “look-elsewhere” effect. This is a phenomenon in statistics where an apparently statistically significant observation may have actually arisen by chance because of the large size of the parameter space to be searched. It represents an important effect for cases where one does not have an a priori model or prediction to which the signal should appear, and, hence, when one needs to search for a signal in a large range. In such cases, the significance calculation must take into account that a high statistical fluctuation anywhere in that range could also be considered as a signal (e.g. see Gross & Vitells 2010, for a more rigorous discussion). The effect is particularly relevant in particle physics, and, in general, in any field in which one searches for uncommon events.

The look-elsewhere effect is important since we do not know a priori what is the number of satellites that we expect to find in a plane. In the case of PAndAS-like observations, there are 27 satellites in each system, which means that the most prominent plane can have anywhere between 3 to 27 members. To estimate the significance of a plane, we need to compute the probability that a statistical fluctuation generating such a prominent plane appeared for any combination of 3, 4, ..., 27 satellites — this is the marginalized probability. This is different from the conditional probability that has the number of satellites chosen as a prior and whose inverse gives the prominence of a plane (e.g. Eq. 2). The marginalized probability is the isotropic CDF shown in Figs. 2 and 3. In other words, the probability that a plane is a statistical fluctuation is given by the fraction of isotropic realizations that have a more prominent plane. Using this, we compute the significance of each plane, which is shown as the top x-axis of Figs. 2 and 3, with the tick marks spaced at 0.2σ intervals. The significance is expressed in multiples of the standard deviation, $\sigma$, of a normal distribution. Note that the one-to-one map between the prominence (bottom x-axis) and the significance level (top x-axis) differs between the two figures.

Some previous studies have incorrectly referred to the conditional probability as the detection significance of planes of satellites (e.g. Kroupa, Theis & Boily 2005, Ibata13). To better emphasize the difference between the two, we introduce the concept of trial factor (e.g. see Gross & Vitells 2010). This is the ratio between the marginalized and the conditional probability to obtain a statistical fluctuation with prominence, $P$. The former corresponds to the actual detection significance and is given by the CDF of an isotropic distribution. The latter corresponds to $P^{-1}$ since this is the definition of a plane’s prominence (e.g. Eq. 2). Thus,

$$\text{trial factor} = \left( \frac{\text{CDF isotropic}}{\text{CDF marginal}} \right)_{P > P_{\text{rarest}}} \cdot \left( \frac{1}{P_{\text{rarest}}} \right). \quad (9)$$

For example, the M31 spatial + 2D-kinematic plane has a trial factor of 115 (see entry (2) in Table 1 for numerical values). Thus, the chance of it being a statistical fluctuation is 115 times higher than naively expected if one considers only random planes with 15 members. Inevitably, this means that Ibata13 have overestimated the detection significance of the M31 plane by more than two orders of magnitude. For an isotropic distribution there is a 0% probability of obtaining a more prominent plane, and, hence, the M31 plane corresponds to a 2.9σ detection.

In Fig. 4 we show the trial factors for spatial and spatial + 2D-kinematic planes. The regions of interest are roughly $P_{\text{rarest}} = 10$ and $P_{\text{rarest}} = 100$, corresponding to the intervals in Figs. 2 and 3 where the isotropic CDF is different from unity. In those intervals, the trial factors increase only slowly with the plane prominence, so, to a first approx-
imation, the two plane types have a trial factor of $\sim 10$ and $\sim 100$, respectively. The \emph{spatial + 2D-kinematic} planes have a higher trial factor due to the larger range used to search for the most prominent plane, since, on top of the spatial distribution, also the 2D kinematics are considered.

### 4.4 The characteristics of rare planes

In Fig. 5 we show the characteristics of the rarest \emph{spatial + 2D-kinematic} planes that are at least as prominent as the M31 plane. We describe the planes in terms of the number of members sharing the same sense of rotation, the plane thickness and the radial extent of the plane, $r_\parallel$. This latter property characterizes the dispersion of the satellites within the plane and it is calculated as the mean sum of the squares of the distance projected on to the best-fitting plane. We choose these plane characteristics to be consistent with previous studies that investigated the incidence of the MW and M31 plane of satellites in term of these properties (e.g. Bahl & Baumgardt 2014; Ibata et al. 2014c).

Fig. 5 shows that there is considerable variation among the properties of the most prominent planes, suggesting that each plane is different. For example, the number of members sharing the same sense of rotation, $N_{\text{sat}_{s.s.r.}}$, can take values between $N_{\text{sat}}/2$ to $N_{\text{sat}}$. The top panel of the figure shows that the planes have $N_{\text{sat}_{s.s.r.}}$ values spanning the full allowed range, although there is a higher preference for $N_{\text{sat}_{s.s.r.}} \simeq N_{\text{sat}}$, since that will result in a higher prominence. The behaviour in the middle panel is governed by two requirements. Firstly, to be prominent, planes with a small number of members need to be very thin since such structures cannot have a high 2D-kinematic prominence, which explains the distribution seen in the left-half of the panel. Secondly, once the number of members is high enough, $\sim 16$ in this case, the 2D-kinematic prominence can be by itself very large, so that such planes do not necessarily need to be very thin. This explains the large scatter in the $r_\parallel$ values seen in the right-half of the middle panel. And lastly, the prominence of a plane does not depend on $r_\parallel$, which explains the large scatter in $r_\parallel$ values seen in the bottom panel of the figure.

Fig. 5 also shows the properties of the M31 plane of satellites whose position is marked with a large triangle. The M31 plane is within the scatter expected for $\Lambda$CDM planes, although it does stand out as having an unusually large radial extent, $r_\parallel$.

### 4.5 The incidence of rare planes

For each halo, we study the incidence of the rarest \emph{spatial + 2D-kinematic} plane among the distribution of satellites of all other $\Lambda$CDM haloes. This is motivated by the studies of Ibata et al. (2014c) and Pawlowski et al. (2014) that interpreted the low incidence of the M31 plane as evidence for an inconsistency between observed planes of satellites and $\Lambda$CDM predictions.

We define the incidence or frequency of a plane using the approach of Ibata et al. (2014c). Each plane of satellites is characterized by: its number of members, $N_{\text{sat}}$; how many of them share the same sense of rotation, $N_{\text{sat}_{s.s.r.}}$; the thickness, $r_\perp$; and radial extent, $r_\parallel$; of the plane. Then, the frequency or incidence, $f_{\Lambda\text{CDM}}(N_{\text{sat}}, r_\perp, r_\parallel, N_{\text{sat}_{s.s.r.}})$, of this plane is given by the fraction of $\Lambda$CDM systems that have a similar plane. To describe the procedure, we exemplify it for the case of two systems A and B. We are interested in the frequency of the rarest plane of satellites of halo A and we wish to find out if halo B has a similar plane. We take all possible satellite configurations of system B that have $N_{\text{sat}}$ members, which
we find using the procedure described in Appendix A1. If any of those configurations is similar to plane A, i.e. \( r^B_\parallel \leq r^A_\parallel \) and \( r^B_\perp \geq r^A_\perp \), then halo B has at least one planar configuration similar to that of system A. We compute the frequency for each MS-II and COCO halo, by taking the characteristics of the rarest plane identified around each halo. Each such rarest plane is compared to the satellites distribution of all the other PAndAS-like mocks. This equates to comparing the rarest plane found in one halo with all possible planes, not only the rarest ones, around all other systems.

The frequency of the rarest \( r^{\text{spatial}} + 2D^{\text{kinematic}} \) planes of satellites is shown in Fig. 6, where we plot the CDF of the frequency for MS-II and COCO planes. We find that over half of ΛCDM haloes have a plane with a frequency of 0.02 or lower, and one tenth have a frequency as low as 0.001. It is important to note that the low frequency is not a consequence of our plane identification method. For each halo, our method selects the plane that is the least likely to be a statistical fluctuations, which is fully independent of the planes found around other ΛCDM haloes. Thus, there is a large diversity of planes of satellites. In other words, if we find a planar configuration around one system, it does not tell us anything about the properties of planes around other haloes.

We performed the same calculation for the M31 system. For each Monte Carlo realization, we compute the incidence of the rarest \( r^{\text{spatial}} + 2D^{\text{kinematic}} \) plane of that realization by comparing with the PAndAS-like mocks. We found that the M31 plane has a frequency of \( 6.8^{+7.2}_{-2.7} \times 10^{-4} \) (1σ confidence interval; see Table 2), which is in good agreement with the results of previous studies (Ibata et al. 2014c; Pawlowski et al. 2014). This low incidence of the M31 plane has been claimed by Ibata et al. (2014c) and Pawlowski et al. (2014) to be a source of discrepancy with ΛCDM. From Fig. 6, which shows the M31 frequency as a vertical dashed line, we find that, within ΛCDM, \( 5.1^{+4.5}_{-0.9} \)% of systems have planes with even lower frequencies. Thus, the low incidence of the M31 plane does not pose a challenge to the current paradigm, in fact, it is consistent with ΛCDM predictions.

Fig. 7 investigates which planes are the ones with the lowest incidence. For this, we plot the incidence of each plane as a function of its prominence, and find a strong anti-correlation between the two, albeit with a large scatter.

The incidence of the M31 and the MW plane of satellite galaxies. Table 2.

| ID | Host | \( f_{\Lambda CDM}(N_{\text{sat}}, \leq r_\perp, \geq r_\parallel, \geq N_{\text{s.s.r.}}) \) | \( f_{\Lambda CDM}(N_{\text{sat}}, \leq r_\perp, \geq r_\parallel, \leq \Delta_{\text{sat}}) \) | fraction of ΛCDM systems with lower frequencies (%) |
|---|---|---|---|---|
| (2) | M31 | \( 6.8^{+7.2}_{-2.7} \times 10^{-4} \) | - | \( 5.1^{+4.5}_{-0.9} \) |
| (4) | MW | \( 4.6^{+2.8}_{-2.5} \times 10^{-3} \) | - | \( 18^{+6}_{-3} \) |
| (5) | MW | - | \( 1.5^{+1.9}_{-1.2} \times 10^{-3} \) | \( 11^{+6}_{-7} \) |

The table columns give: the plane ID from Table 1, the central galaxy (M31 or MW), the incidence, \( f_{\Lambda CDM}(N_{\text{sat}}, \leq r_\perp, \geq r_\parallel, \geq N_{\text{s.s.r.}}) \), and \( f_{\Lambda CDM}(N_{\text{sat}}, \leq r_\perp, \geq r_\parallel, \leq \Delta_{\text{sat}}) \), of similar planes among the MS-II haloes, and the fraction of ΛCDM haloes that have planes of satellites with even lower frequencies.

**Figure 6.** The CDF of the incidence, \( f_{\Lambda CDM}(N_{\text{sat}}, \leq r_\perp, \geq r_\parallel, \geq N_{\text{s.s.r.}}) \), of similar planes to the rarest one in ΛCDM. For each ΛCDM halo, we take the most prominent plane and find its frequency among all other ΛCDM haloes (see the text for more details). Most planes have a low incidence (half of systems have a frequency of 0.02 or lower) indicating that each ΛCDM halo has a different planar configuration. The vertical dashed line and the grey area show the incidence and the 1σ error for the M31 plane. We find that \( 5.1^{+4.5}_{-0.9} \)% of ΛCDM systems have a lower frequency than the M31 plane. Thus, the low incidence of the M31 plane is not in tension with ΛCDM, as claimed by Pawlowski et al. (2014), but instead is consistent with ΛCDM expectations.

**Figure 7.** The correlation between a plane’s prominence, \( p_{\text{rarest}}^{r^{\text{spatial}} + 2D^{\text{kinematic}}} \), and its incidence, \( f_{\Lambda CDM}(N_{\text{sat}}, \leq r_\perp, \geq r_\parallel, \geq N_{\text{s.s.r.}}) \), among other ΛCDM haloes. The small dots correspond to the rarest plane found for each MS-II halo. The solid line shows the median trend. The large symbol with error bars shows the position of the M31 plane. The small down-pointing arrows found at the bottom of the graph show upper limits corresponding to planes that do not have another counterpart in MS-II.
Table 3. The positions and velocities of the 11 classical MW satellites with respect to the Galactic Center.

| Name       | $x$ (kpc) | $y$ (kpc) | $z$ (kpc) | $V_x$ (km/s) | $V_y$ (km/s) | $V_z$ (km/s) |
|------------|-----------|-----------|-----------|--------------|--------------|--------------|
| Sagittarius| 17.1 ± 1.9| 2.5 ± 0.2 | −6.5 ± 0.5| 234 ± 7      | 19 ± 21      | 224 ± 21     |
| LMC        | −0.5 ± 0.3| −41.8 ± 1.6| −27.5 ± 1.1| −42 ± 12     | −226 ± 13    | 234 ± 16     |
| SMC        | 16.5 ± 1.6| −38.5 ± 2.4| −44.7 ± 2.8| 2 ± 18       | −161 ± 26    | 149 ± 21     |
| Draco      | −4.4 ± 0.3| 62.3 ± 4.9 | 43.2 ± 3.4 | 74 ± 24      | 43 ± 14      | −210 ± 19    |
| Ursa Minor | −22.2 ± 0.6| 52.0 ± 2.1| 53.5 ± 2.1 | 7 ± 28       | 89 ± 20      | −186 ± 20    |
| Sculptor   | −5.2 ± 0.3| −9.7 ± 0.7 | −85.3 ± 5.9| −33 ± 44     | 188 ± 45     | −99 ± 6      |
| Sextans    | −36.6 ± 1.3| −56.8 ± 2.6| 57.8 ± 2.7 | −168 ± 160   | 114 ± 133    | 117 ± 127    |
| Carina     | −25.0 ± 1.0| −95.8 ± 5.5| −39.7 ± 2.3| −74 ± 44     | 8 ± 19       | 40 ± 41      |
| Fornax     | −41.1 ± 2.7| −50.8 ± 4.1| −134 ± 11 | −38 ± 27     | −156 ± 42    | 113 ± 18     |
| Leo II     | −77.3 ± 4.1| −58.3 ± 3.5| 214 ± 13  | 102 ± 127    | 237 ± 156    | 117 ± 50     |
| Leo I      | −124 ± 7   | −119 ± 7  | 192 ± 12  | −167 ± 31    | −35 ± 33     | 96 ± 24      |

The x-axis points from the Sun towards the Galactic Centre, the y-axis points in the direction of Galactic rotation at Sun’s position and the z-axis points towards the North Galactic Pole. Since this is a rotated coordinate system, the uncertainties are correlated and are very anisotropic in the plane of the sky.

The vertical concentrations of points correspond to planes that have a very high 2D-kinematic prominence but only a very low spatial one, with the discrete nature of the 2D-kinematic test leading to many planes having very similar $\alpha_{\text{spatial}}$ + 2D-kin values. So, on average, the more prominent a plane is, the lower is its incidence among ΛCDM systems. This explains why the M31 plane, which has a high prominence, also has a low incidence. The M31 plane, shown in Fig. 7 as a square symbol with error bars, is consistent with the object-to-object scatter in the prominence-incidence relation. Compared to the median trend, the M31 plane has an ~5 times lower incidence for its prominence, which could be due to its unusually large radial extent (see bottom panel of Fig. 5).

5 MW-LIKE PLANES OF SATELLITES

In this section we investigate the MW system of satellites and how it compares with other ΛCDM planar configurations. Compared to the M31 analysis, there are three main differences: we use systems with 11 instead of 27 satellites, the survey geometry is different and, most importantly, we can perform additional tests since we have proper motion data for the MW satellites. In this analysis, we consider only the brightest 11 classical Galactic satellites since only these objects have measured proper motions. A twelfth satellite, Canes Venatici, has a similar absolute magnitude as the faintest of the classical satellites (McConnachie 2012), but it does not have a measured proper motion, so we do not include it in this study.

To create MW-like mocks, we consider the 11 satellites with the largest stellar masses that are within a distance of 260 kpc from the central galaxy (corresponding to Leo I which is the furthest at 254 kpc) and that, at the same time, are outside an obscuration region consisting of 33% of the sky. The obscuration region accounts for the Galactic zone of avoidance, where, due to large extinction and confusion by foreground stars, it is possible to have yet undiscovered bright satellites. For this, we use the estimate of Willman et al. (2004) according to which the census of Galactic dwarf galaxies may be 33% incomplete. Yíñiguez et al. (2014) estimates an even higher incompleteness, with most of those undetected systems further than 100 kpc from the Galactic Centre. A satellite is lying inside the obscuration region, and hence undetected, if its latitude, $\theta$, is in the range $-\theta_{\text{crit}} \leq \theta \leq \theta_{\text{crit}}$, with $\theta_{\text{crit}} = 19.5^\circ$ (due to an error, Wang, Frenk & Cooper (2013) misquoted the critical angle as having a value of 9.5°, which would correspond to an obscuration region of 17%). To generate mock observations, we take three viewing angles for each simulated halo such that the mock north Galactic pole corresponds to the x, y and z axes of the simulation. This procedure yields 8547 mock satellite systems for MS-II and 189 for COCO.

5.1 The MW system

In a first step, we analyse the planar configurations found around the MW. For each of the 11 Galactic satellites, we take their radial distance and velocity, as well as the errors associated with these quantities, from McConnachie (2012). To compute the mean proper motion, we follow the approach of Pawlowski & Kroupa (2013) and weigh the different measurements according to their errors. We obtain the same mean proper motions as they do, except for Draco, where the latest measurement, $(\mu_\alpha, \mu_\delta) = (0.177 \pm 0.063, -0.221 \pm 0.063)$ mas yr$^{-1}$ (Pryor, Piatek & Olszewski 2015), is significantly different from the value given by Pawlowski & Kroupa (2013). Using this updated measurement results in a weighted mean value $(\bar{\mu}_\alpha, \bar{\mu}_\delta) = (0.187 \pm 0.063, -0.201 \pm 0.063)$ mas yr$^{-1}$. We also included an additional proper motion measurement for Sagittarius of $(\mu_\alpha, \mu_\delta) = (-2.95 \pm 0.21, -1.19 \pm 0.16)$ mas yr$^{-1}$ (Massari et al. 2013) that resulted in a mean value of $(\bar{\mu}_\alpha, \bar{\mu}_\delta) = (-2.711 \pm 0.066, -1.043 \pm 0.065)$ mas yr$^{-1}$, nearly the same as the mean value used by Pawlowski & Kroupa (2013).

We transform the satellite positions and velocities to a Cartesian coordinate system with the origin at the Galactic Centre. The x-axis points from the Sun towards the Galactic Centre, the y-axis points in the direction of Galactic rotation at the Sun’s position and the z-axis points towards the North Galactic Pole. For this transformation we adopt: the distance of the Sun from the Galactic Centre $d_0 = 8.29 \pm 0.16$ kpc, the circular velocity at the Sun’s position, $V_{\text{circ}} = 239 \pm 5$ km/s (McMillan 2011), and the Sun’s motion with respect to the Local Standard of Rest, $(U,V,W) = (11.1 \pm 0.8, 12.2 \pm 0.5, 7.3 \pm 0.4)$ km/s (Schönrich,
Binney & Dehnen 2010). To account for observational errors, we generate 1000 Monte Carlo realizations of the MW system of satellites. We sample the satellite positions and proper motions from Gaussian distributions centred on the most likely values of each quantity and with dispersion equal to the uncertainties. Similarly, we also account for the errors in $d_0$, $V_{circ}$ and the Local Standard of Rest. Following this, we transform from heliocentric coordinates to Galactic ones, with the result used as input for our plane detection method. We summarize in Table 3 the mean positions, velocities and $1\sigma$ errors associated with each Galactic satellite.

As for the M31 case, for each Monte Carlo realization of the MW system we compute the rarest spatial, spatial + 2D-kinematic and spatial + 3D-kinematic planes. The results are summarized in Table 1. Independently of the plane type, we find that the rarest plane is the one that contains all the 11 Galactic satellites. This is in agreement with previous studies that found that all the classical satellites are members of the MW satellite plane (e.g. Kroupa, Theis & Boily 2005).

5.2 The rarest MW-like planes

To put the MW plane of satellites into context, we proceed by identifying the rarest planes around mock MW-like observations. The outcome is shown in Figs. 8-10 that give the CDF of the prominence of the rarest spatial, spatial + 2D-kinematic and spatial + 3D-kinematic planes, respectively. The main conclusions are the same as for the PAndAS mocks analysed in Sec. 4.2.

The look-elsewhere effect again plays an important role for the MW-like mocks, with spatial planes having a trial factor ~8, while spatial + 2D-kinematic and spatial + 3D-kinematic planes have a trial factor ~30. The trial factors are roughly constant in the region where the isotropic CDF is below unity, reminiscent of the results in Fig. 4. For brevity, we do not show these results. The trial factors are lower in the case of the MW mocks than in the case of M31, reflecting the narrower range used to search for planar configurations in the MW (11 satellites compared to 27 for the M31).

Figs. 8-10 also indicate the prominence of the MW plane of satellites as a dotted vertical line. The MW plane stands out the most in terms of its spatial + 3D-kinematic prominence since in this case it corresponds to a $2.8\sigma$ statistical fluctuation. Not accounting for the look-elsewhere effect, would lead one to estimate the MW plane as a $3.8\sigma$ detection. While the MW spatial + 3D-kinematic plane is quite conspicuous, it is consistent with ΛCDM since $5.0^{+4.1}_{-2.7}\%$ of galactic-mass haloes have even more prominent planes.

As we emphasized in Sec. 3.3, one needs to be careful when interpreting the results of the 3D-kinematic analysis since this test has been designed a posteriori. In fact, one could easily come up with other 3D kinematic tests that are physically better motivated, as we discussed in Sec. 3.3. Given that observationally we have only one satellite system with 3D velocities, it is impossible at present to assess if the 3D-kinematic test that we have applied is generic and thus appropriate to the whole population of satellite systems or if it is matched to the particular details of the MW satellites. If the latter is true, then the fraction of ΛCDM haloes with more prominent planes does not convey any physically meaningful information. Thus, there is currently

Figure 8. The CDF of the prominence, $\mathcal{P}_{\text{rarest}}$, of the rarest spatial plane of satellites for mock MW observations. The vertical dashed line and shaded region show the prominence and $1\sigma$ error for the MW plane of satellites, with $(12 \pm 1)\%$ of ΛCDM haloes having a more prominent plane.

Figure 9. As Fig. 8, but for the prominence, $\mathcal{P}_{\text{rarest}}$, of the rarest spatial + 2D-kinematic plane of satellites. In this case, $(10 \pm 5)\%$ of ΛCDM hosts have a more prominent plane than the MW.

Figure 10. As Fig. 8, but for the prominence, $\mathcal{P}_{\text{rarest}}$, of the rarest spatial + 3D-kinematic plane of satellites. For this test, $5.0^{+4.1}_{-2.7}\%$ of ΛCDM haloes have a more prominent plane than the MW.
The characteristics of the rarest planes of satellites that are at least as prominent as the MW plane of satellites, i.e., $P_{\text{rarest}}^{\text{spatial + 3D-kin}} \geq 2.3 \times 10^4$. There are 778 such systems. The grid cells are coloured according to the number of systems with those properties. The three panels show the orbital pole dispersion, $\Delta_{\text{std}}$ (top), the thickness, $r_\perp$ (centre), and the radial extent of the plane, $r_\parallel$ (bottom), as a function of the number of satellites in the plane, $N_{\text{sat}}$. The large triangle shows the corresponding characteristics of the MW plane of satellites.

Figure 12. The CDF of the incidence, $f_{\Lambda\text{CDM}}(N_{\text{sat}}, \leq r_\perp, \geq r_\parallel, \geq N_{\text{s.s.r.}})$, of finding similar planes in $\Lambda\text{CDM}$. As in the M31 case, this illustrates the diversity of planar configurations found in MW-like mocks. The MS-II and COCO results are consistent with the scatter expected due to the low number of COCO haloes. The vertical dashed line and the grey area show the incidence and the 1σ error for the MW plane of satellites. We find that $18^{+6}_{-8}\%$ of $\Lambda\text{CDM}$ systems have an even lower frequency than the MW plane.

Figure 13. As Fig. 12, but for the incidence, $f_{\Lambda\text{CDM}}(N_{\text{sat}}, \leq r_\perp, \geq r_\parallel, \leq \Delta_{\text{std}})$. In this case, we compare the orbital pole dispersion, $\Delta_{\text{std}}$, between different planes and not the number of satellites sharing the same sense of rotation, $N_{\text{s.s.r.}}$, as in Fig. 12. We find that $11^{+9}_{-5}\%$ of $\Lambda\text{CDM}$ systems have an even lower incidence than the MW plane. Thus, the low incidence of the MW plane is not a symptom of discrepancy with $\Lambda\text{CDM}$, as claimed by Pawłowski et al. (2014).

not enough data to decide if one should be concerned that only $\sim 5\%$ of $\Lambda\text{CDM}$ haloes have a more prominent spatial + 3D-kinematic plane than the MW one.

In Fig. 11 we plot the properties of the rarest spatial + 3D-kinematic planes that are at least as prominent as the MW satellite plane. In analogy to Sec. 4.5, we find that the planes are characterized by a large diversity in orbital pole dispersion, thickness and radial extent. A plane can be very prominent by being very thin, by having a small orbital pole dispersion or by a combination of the two, which explains the large scatter seen in the $\Delta_{\text{std}}$ and $r_\perp$ properties. Interestingly, we find that most of such planes ($43\%$) have $N_{\text{sat}} = 11$, the same as the number of members in the MW plane of satellites whose characteristics are shown as a large triangle.
5.3 The incidence of MW-like planes

We follow the same approach as in Sec. 4.5 and compute the incidence of the rarest plane around each ΛCDM halo. The outcome is presented in Fig. 12 which shows that most systems have planar configurations that are very infrequent, with over half of the haloes having a plane with an incidence of 0.03 or lower. The slight disagreement between the MS-II and the COCO results is consistent with the scatter expected for the much smaller sample of COCO systems. This appears as a systematic shift due to the correlations between points in the CDF. For this test, the MW plane of satellites has a frequency of \(4.6^{+2.5}_{-4.6} \times 10^{-3}\) (vertical dashed line in Fig. 12) that is consistent with ΛCDM expectations, since 18.3\(^\%\) of haloes have an even lower incidence.

Since for the Galactic satellites we have full kinematical data, we can define a new incidence, \(f_{\text{ACDM}}(N_{\text{sat}}, \leq r_\perp, \geq r_\parallel, \leq \Delta_{\text{std}})\), that includes the 3D kinematics. This is similar to the incidence introduced in Sec. 4.5, except that now we compare \(\Delta_{\text{std}}\) between different planes instead of \(N_{\text{sat}}\). The CDF of this new frequency is shown in Fig. 13. Compared to Fig. 12, the outcome is very similar except for a slight shift in the CDF towards the left, i.e. towards lower frequencies.

The new statistics indicates that the MW plane of satellites has an incidence of \(1.5^{+1.2}_{-1.3} \times 10^{-3}\), as shown by the vertical dashed line in Fig. 13. This value agrees with the result of Pawlowski et al. (2014, entry 12 of their Table 5), whose test is very similar to ours, except that those authors considered the orbital pole dispersion of only 8 out of 11 satellites. Pawlowski et al claimed that this low incidence of the MW plane of satellites is indicative of a shortcoming of the ΛCDM paradigm. Instead, we find that the low incidence indicates that planes of satellites are very diverse.

The distribution of satellites around the MW agrees with ΛCDM predictions since 11.4\(^\%\) of similar mass haloes have an even lower frequency.

6 DISCUSSION AND CONCLUSIONS

In this paper, we have investigated the prevalence and properties of planar configurations of satellites around galactic mass haloes. Using two very high resolution cosmological simulations, MS-II and COCO, we have built mock galaxy catalogues corresponding to the satellite distributions around the MW and M31. MW-like mocks consist of the most massive 11 satellites found outside an assumed zone of avoidance, mimicking the observations of the 11 classical Galactic satellites. M31-like mocks are modelled according to the PAndAS footprint and consist of the most massive 27 satellites found within that region, corresponding to the M31 satellites observed by PAndAS.

We identify the most prominent planar configuration as the subsample of satellites whose spatial and kinematical distribution is the least likely to be a statistical fluctuation. Applying our approach to the MW and M31 observations results in the same planar distributions as determined by Kroupa, Theis & Bovy (2005) and Ibata13, respectively, even though those studies used different identification methods. The good agreement is possibly due to a posteriori selection bias, since those authors may have inadvertently tuned their methods to maximize the significance of the detection. This would result in all the methods converging to the same planes.

We have found that planar configurations of satellites are very common around ΛCDM haloes, and, moreover, approximately 5 and 9% of haloes have even more prominent planes than those found in the MW and M31, respectively (see Table 1, and Figs. 3 and 10). The look-elsewhere effect is crucial in assessing the detection significance of a planar distribution, i.e. in estimating the probability of obtaining such a structure in an isotropic distribution (Sec. 4.3). By neglecting this effect, one can easily overestimate the significance level by factors of \(~30\) and \(~100\) for the MW and M31 planes respectively. For example, while the M31 plane was originally reported to have a 99.998\(^\%\) significance (4.3\(\sigma\) detection; Ibata13), accounting for the look-elsewhere effect results in a more modest 99.7\(^\%\) significance (2.9\(\sigma\) detection).

While ubiquitous, the planes of satellites show a large diversity in characteristics, e.g. in the number of members, the plane thickness and radial extent, as well as the kinematical structure (see Figs. 5 and 11). Most planar configurations are distinct, which has two major implications. Firstly, the notion of a representative plane of satellites does not exist since one cannot find a majority of ΛCDM haloes that have the same planar configuration. Secondly, the large diversity of planes precludes using one or two observed systems for testing the cosmological paradigm on small scales. For such a test, a large sample of satellite systems would be needed to obtain a statistical measure of the system-to-system variation.

The diversity of the planes of satellites is also the root cause behind previous claims that planes found in observations are inconsistent with ΛCDM (Ibata et al. 2014c; Pawlowski et al. 2014). These authors computed the incidence of the MW and M31 plane of satellites to find out that roughly only 1 out of 1000 ΛCDM systems have such planes. To understand this result, we have computed the incidence of the rarest plane of satellites identified around each ΛCDM halo. We have found that the majority of planar configurations have a very low incidence and that 11 and 5% of ΛCDM planes have even lower incidence than that of the planes of satellite in the MW and M31, respectively (see Table 2, and Figs. 6 and 13). The low incidence is a manifestation of the diversity of satellite planes and, thus, contrary to the claim by Pawlowski et al. (2014), it does not rule out the ΛCDM paradigm.

While the planes of satellites around the MW and M31 are consistent with ΛCDM, both systems fall in the 10\(^\%\) tail of the distribution. If both planes were independent of each other, one might argue that there is only a \(~1\%\) chance that both systems are randomly drawn from a ΛCDM distribution. This interpretation is problematic for at least two reasons. Firstly, both the MW and the M31 are located in the same large-scale environment, which in turn determines the preferential directions of satellite accretion (Libeskind et al. 2014, 2015). Thus, if the environment is especially conducive to the formation of prominent satellite planes, then it may not be surprising that both systems host prominent planes. Secondly, the tests used to assess the prominence of these planes were designed a posteriori, after investigating the observational data. This is especially true for the definition of the orbital pole dispersion that has been motivated by examining the MW data. Such an a posteriori approach incurs
the danger of designing tests that are specifically matched to the peculiarities of a particular system and are not characteristic of the population as a whole.

Our analysis has shown that the planes of satellites identified in the MW and M31 are consistent with ΛCDM predictions based on high resolution cosmological simulations. This agrees with the results of Cautun et al. (2015), which compared the spatial and kinematical distributions of satellites around a large sample of isolated galaxies in SDSS to find agreement between observations and theoretical predictions. Previous claims of an inconsistency with ΛCDM are based on misinterpreting the low incidence of satellite planes (e.g., Ibata et al. 2014c; Pawlowski et al. 2014) and on non-robust detections (Ibata et al. 2014a, see Sec. 1 and Cautun et al. 2015 for details). Thus, there is no convincing evidence for a discrepancy between observed planes of satellites and the ΛCDM predictions.

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In the case of the M31 system, we have to generate \( N_{\text{max}} = 27 \) satellites for many thousands of systems, and for each we would need to consider. This analysis would have to be done in a way that results in a very large number of combinations that need to be taken into account every possible combination of \( N_{\text{sat}} \) objects, with \( 3 \leq N_{\text{sat}} \leq N_{\text{max}} \). Planes with 2 or fewer objects are not physically interesting since any two satellites will determine a plane of \( \perp = 0 \). This naive approach, however, would result in a plane. In those cases we end up identifying the second or the third rarest planes as the most prominent planar configurations. Using a small sample of around 200 \( \Lambda \)CDM haloes, we have checked that using \( N = 10^3 \) instead of \( N = 10^4 \) brings only minor changes to the CDF of the prominence, \( \mathcal{P}_{\text{max}} \), of the rarest plane (e.g. Fig. 2) and to the CDF of the frequency, \( f_{\Lambda \text{CDM}} \), of those planes (e.g. Fig. 6). Thus, any missing subsets of satellites will not change our overall conclusion.

The subsets of satellites used to compute the frequency of the rarest planes (Secs. 4.5 and 5.3) were identified employing the same procedure except that we used \( N = 10^5 \). That is, we used \( 10^5 \) random planes uniformly distributed on a hemisphere, which is the same as the approach used by Bahl & Baumgardt (2014) and subsequent studies.

## A2 Generating isotropic distributions

Each isotropic realization is generated by picking random polar and azimuthal angles\(^2\) for each satellite, while keeping constant the radial distance from the host galaxy\(^3\). Thus, each isotropic realization has the same radial distribution of satellites as the original system. This point is crucial, since the radial distribution of satellites has a strong effect on the thickness of the resulting planes. Radially concentrated satellite distributions result in thinner planes than more radially extended ones. Thus, we need to generate new isotropic realizations for each system of satellites.

When constructing the isotropic distributions, we also generate random 3D velocities, which are used for computing the distribution of orbital pole dispersions, \( \Delta_{\text{stat}} \). Since we are only interested in the direction of the orbital momentum, the magnitude of the velocity is not important. Thus, the velocities are generated by picking random polar and azimuthal angles.

\(^2\) The cosine of the polar angle and the azimuthal angle are selected from a uniform distribution spanning the interval \([-1, 1]\) and \([0, 2\pi]\), respectively.

\(^3\) Each random realization lies within the survey mask, which is PAndAS for the M31 and a 19.5\(^\circ\) obscuration angle for the MW. If a random point falls outside the mask, we generate new random angles till the point is located within the survey mask.
azimuthal angles for each satellite, with the two angles fully independent from the random polar and azimuthal angles used to obtain the position of each satellite.

A3 The probability distribution of statistical fluctuations

We now describe how to estimate the probability that the spatial or kinematical distribution of a set of $N_{\text{sat}}$ satellites is the result of a statistical fluctuation. This probability is computed using isotropic distributions, which characterize the degree of planarity expected from chance alignments and from the discreteness of the satellite distribution.

The probability of obtaining by chance a plane of $N_{\text{sat}}$ that is thinner than $r_{\perp}$ is given by

$$p(\leq r_{\perp} \mid N_{\text{sat}}) = \int_{0}^{r_{\perp}} PDF_{\text{isotropic}}(r_{\perp}^{'}) \, dr_{\perp}^{'} \, ,$$  \hspace{1cm} (A2)

where the integrand is the PDF of obtaining in an isotropic distribution planes with $N_{\text{sat}}$ members and thickness $r_{\perp}^{'}$. To compute the PDF, for each halo we generate $10^5$ isotropic realizations using the procedure described in Appendix A2. For each such realization we find the thinnest plane with $N_{\text{sat}}$ members. The corresponding histogram over all realizations gives the PDF of the orbital pole dispersion, $\Delta_{\text{std}}$, for an isotropic distribution. We show planar configurations that contain $N_{\text{sat}} = 5, 8$ and $27$ satellites out of a maximum of $N_{\text{max}} = 27$ satellites.

Figure A1. The PDF of the plane thickness, $r_{\perp}$, for isotropic satellite distributions inside the PAndAS survey footprint. We show results for planar configurations that contain $N_{\text{sat}} = 8, 15$ and $27$ satellites out of a maximum of $N_{\text{max}} = 27$ satellites. The vertical arrow and shaded region illustrate the probability that a fictitious plane A, which has $N_{\text{sat}} = 15$ and $r_{\perp} = 17.5$ kpc, is due to a statistical fluctuation (see the text for details).

The probability of obtaining by chance a plane of $N_{\text{sat}}$ members that has an orbital pole dispersion less than $\Delta_{\text{std}}$ is calculated as

$$p(\leq \Delta_{\text{std}} \mid N_{\text{sat}}) = \frac{N_{\text{sat}}!}{N_{\text{sat}}!} [\langle N_{\text{sat}} - N_{\text{sat}} \rangle!]^{2N_{\text{sat}}} \, ,$$  \hspace{1cm} (A3)

where the first factor of 2 comes from the fact that we do not fix a preferential sense of rotation, allowing both clockwise and counter-clockwise rotations.

The probability of obtaining by chance a plane with $N_{\text{sat}}$ members sharing the same sense of rotation is given by the binomial distribution with a success probability of 0.5. Thus,

$$p(\geq N_{\text{s.s.r.}} \mid N_{\text{sat}}) = \frac{N_{\text{sat}}!}{N_{\text{s.s.r.}}!} [\langle N_{\text{sat}} - N_{\text{s.s.r.}} \rangle!]^{2N_{\text{sat}}} \, ,$$

where $N_{\text{s.s.r.}}$ is the number of satellites that have a shared sense of rotation.

The integrand gives the PDF of the orbital pole dispersion, $\Delta_{\text{std}}$, for an isotropic distribution, which is estimated using $10^5$ random realizations, as for Eq. (A2). For each such isotropic realization, we find the subsample of $N_{\text{sat}}$ satellites that has the lowest orbital pole dispersion. The histogram of the lowest $\Delta_{\text{std}}$ values over all realizations gives the PDF used in Eq. (A4). In Fig. A2 we illustrate the outcome of such a calculation for the case of $N_{\text{sat}} = 5, 8$ and $11$ out of a maximum satellite count, $N_{\text{max}} = 11$.

Figure A2. The PDF of the angular dispersion of orbital poles, $\Delta_{\text{std}}$, for an isotropic satellite distribution. We show planar configurations that contain $N_{\text{sat}} = 5, 8$ and $11$ satellites out of a maximum of $N_{\text{max}} = 11$ satellites. To obtain these results, we used the same survey footprint and number of satellites as employed for the analysis of the MW satellite system in Sec. 5.