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# A spatial analysis of the XIII Italian Legislature * 

Massimiliano Landi<br>Singapore Management University<br>Riccardo Pelizzo<br>Singapore Management University

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#### Abstract

We present a spatial map of the Italian House during the XIII Legislature obtained by applying the Poole and Rosenthal methodology to roll call data. We obtain coordinates for almost all the 650 MPs that were on the House's floor at the time, and we aggregate them according to parties. We find that voting patters generate basically a two dimensional political space. The first dimension represents loyalty to either the ruling coalition or the opposing one. The second dimension may describe differences at the constitutional level. This finding is consistent with the exceptional case of the party Northern League, which at the time did not belong to either coalition, and presented itself as a northern and anti-system party. Last, we compute the average dispersion of party coordinates along each dimension and compare them with the Rice index of cohesion, the agreement index (which takes into account abstention), and one other index we construct to account for absence from voting. We find that absence is significantly correlated with the dispersion of parties along the second dimension. We use this to motivate the importance of further analysis on the massive absence in Italian Parliament from voting sessions.


[^0]
## 1 Introduction

This paper offers a spatial map of the Italian political space as it unravels from the final votes cast by the members of the House (the lower chamber of the Italian Parliament) in the course of the XIII legislature (1996-2001). We apply the methodology developed and applied to the U.S. Congress by Poole and Rosenthal [10] (PR henceforth) and we then address three basic questions: 1) what is the dimensionality of the Italian party system? 2) What do these dimensions mean? 3) How do our results relate to previous spatial studies? Further, we look at how party cohesion is related to dispersion in each dimension of the political space, and we find some suggestive evidence about the strategic role played by absence from and abstention in voting sessions.

The Italian political space has been investigated for the past three decades, under a variety of methodologies and in different time frameworks: Party positions have been estimated by using expert surveys data (Warwick [15]), mass survey data (Sani and Sartori [14], Corbetta, Parisi and Schadee [3]), ecological data (Ricolfi [11]), and party manifesto data (Budge et alii [1], Campus [4], and Pelizzo [8]).

Little agreement has been achieved about the number and the nature of the dimensions of the political space under investigation. Some have in fact suggested that the Italian political space is or can be properly represented as uni-dimensional, with parties ordered along the (ideological) left-right dimension. This is the interpretation given by Sani and Sartori [14] for the mid seventies, by Corbetta, Parisi and Schadee [3] for the 1966-1983 period, by Campus [4] for the 1996 elections, and by Pelizzo [8] for the 1948-1996 period. Others have argued that the Italian political space is multi-dimensional. This is the evidence found by Ricolfi [11] and [12] for the periods 1953-1992 and 1994-1996, by Loera and Testa [7] who have investigated the dimensionality of the perceptual space of the voters in the new millennium. It is however interesting to note that the scholars advocating the multi-dimensional interpretation of the Italian political space have not reached a consensus as to what are the structuring dimensions of the Italian political space. For several decades it was believed that the first dimension of the Italian political space was the left-right dimension and that the second dimension divided pro-system parties from anti-system parties (Sani [13]). Recent studies have instead suggested not only that the second dimension might have a different meaning (Ricolfi and Testa 2002) but also that the left-right dimension may no longer represent the main dimension of the Italian political space (Ricolfi [9] [page 31]).

By applying the PR methodology, we can estimate the political coordinates of each party and the dimensionality of the political space as they unravel from voting behaviors in the House. We find that the Italian political space is best understood as two-dimensional, and we discuss the nature of these two political dimensions. Specifically we argue that the first dimension structuring the parliamentary party system is the loyalty to the group each party belongs to, either Freedom Pole (Polo) or Olive tree (Ulivo), while the second dimension reflects differences at Constitutional level that are somehow related to the North-South cleavage. Moreover, we estimate parties' dispersion along each dimension, which constitutes another element of inter-party differences and similarities, and we find that dispersion on each dimension relates to different factors. In particular, dispersion on the first dimension is smaller and significantly correlated with the Rice agreement index. On the other hand, dispersion on the second dimension is significantly correlated with an agreement index we generate to account for absence in voting sessions. We take this as suggestive evidence of the strategic and yet not very well understood role of absenteeism in parliamentary dynamics.

The remainder of the paper is organized as follows: section 2 discusses the data we collected, section 3 describes in some detail the PR methodology; section 4 reports the spatial map so obtained and discusses the dimensionality of the political space and its interpretation, while section 5 discusses the relationship between the parties' cohesion and the nature of these political dimensions. Section 6 discusses some possible implications of our findings and suggests some possible venues for future research.

## 2 The data

Voting procedure in the Italian Parliament is very elaborate. In the standing orders of the House at least twelve articles (with several subarticles) that discipline parliamentary votes. In particular, any time the number of Yea (Nay) exceeds Nay (Yea) and at least half of the MPs are participating to the vote (i.e. if the quorum is met), a bill is passed (rejected). MPs that are absent because on duty, are considered as present in the computation of the quorum. Similarly, MPs that show up but abstain, i.e. do not express either Nay or Yea, are considered as present.

Voting is managed electronically or can be done by raising hand (the most common), by splitting the house in two groups, or by roll call. In some specific instances, such as for bills dealing with human and family rights,
constitutional bodies, and electoral rules, voting is secret. In other cases, bills may be passed within Committees, without involving the whole House. (See article 92.1 of the standing orders.)

During the XIII Legislature, 905 bills were published in the Gazzetta Ufficiale. ${ }^{1}$ Some may have been approved during the previous Legislature, some may have been passed with secret vote, while some others may have been approved within Commissions. As a result we a have track of voting records for 630 bills. The House hosts 630 MPs, but because of some turnover we have a record of as many as 651 of them. For each bill we know whether each MPs was absent, on a justified absence, present and voted Yea , or Nay or abstained. We just coded votes Yea and Nay while treated all the other cases as missing. The dataset so created has been used to estimate MP's policy positions according to the PR methodology, which we describe in the next section. ${ }^{2}$

## 3 The methodology

The methodology of estimating roll calls developed by PR is based on a spatial model of probabilistic voting in which MPs are assumed to have single peaked preferences, represented by a normal utility function, over an $n$ dimensional space (to be determined empirically). From now on we will refer to these peaks as MPs' bliss points or MPs' coordinates interchangeably. MPs vote for the alternative (Yea or Nay ) they prefer the most. The probability of making a mistake when voting is normally distributed and i.i.d across roll calls and alternatives.

As a consequence, the payoff associated to the, say, Yea outcome of roll call $j$, in an $s$ dimensional policy space is given by

$$
\begin{equation*}
u(i, j, y)=\beta e^{-\frac{1}{2} \sum_{k=1}^{s} w_{k} d_{i j k y}^{2}}, \tag{1}
\end{equation*}
$$

where $\beta$ is a scaling factor which is proportional to the variance of the error term, $w_{k}$ are the salience weights, and $d_{i j k y}^{2}$ is the squared distance of legislator $i$ from the Yea outcome along the dimension $k$. By letting $\epsilon_{i j k}$ denote the error term, it follows that the distribution of the difference between the two utilities is

$$
U(i, j, y)-U(i, j, n) \sim N\left(u(i, j, n)-u(i, j, y), \sigma^{2}\right)
$$

[^1]and the probability that legislator $i$ votes for Yea in roll call $j$ is given by
$$
P_{i j y}=\Phi\left(\frac{u(i, j, y)-u(i, j, n)}{\sigma}\right) .
$$

It is possible to estimate MPs' bliss points as well as each roll call coordinate (one for the Yea and one for the Nay outcome). The procedure is called NOMINATE and is based on a three steps algorithm: first, given some reasonable initial values of the MPs' bliss points, the roll call parameters are estimated; second, given those estimation of the roll call parameters, a new estimation for the MPs' bliss points is obtained. Last, given both MPs' bliss points and roll call points, the utility function parameters are estimated. ${ }^{3}$ The next section reports our findings and the spatial map of the Italian Parliament.

## 4 Spatial analysis

Table 1 shows the summary statistics of our estimations. The first row reports the number of bills that are in the dataset, the second indicates the cutoff criterium used for rejecting the bills; thus roll calls with a majority of at least $97.5 \%$ were not considered in the estimation. This is a traditional cutoff rule. We also tried several less stringent cutoff rules, to decrease the number of roll calls that were discharged, but the results, in terms of goodness of fit, do not improve. As a result, 388 roll calls were rejected (line three) and 242 were accepted (line four).

Similar numbers (rows five and seven) are reported for the number of MPs that were available (650) and those who were rejected (9). The cutoff criterium of 20 (which is still standard) means that only MPs for which there is a record of voting in at least 20 roll calls are considered (line six). Thus we could get coordinates for 642 MPs .

The remaining rows in table 1 show the goodness of fit statistics (percentage of correct classified - PCC, and aggregate proportional reduction in error - APRE) if we estimate a model with one, two or three dimensions. ${ }^{4}$

[^2]| Roll-calls read | 630 |
| :--- | :--- |
| Cutoff for bills | 0.025 |
| Number rejected | 388 |
| Number accepted | 242 |
| Legislators read | 651 |
| Cutoff for MPs | 20 |
| Number rejected | 9 |
| Number accepted | 642 |
| PCC(1) | 96.192 |
| APRE(1) | 0.827 |
| PCC(2) | 97.694 |
| APRE(2) | 0.895 |
| PCC(3) | 97.940 |
| APRE(3) | 0.906 |
| APRE(2) - APRE(1) | $6.8 \%$ |
| APRE(3) - APRE(2) | $1.1 \%$ |

Table 1: Summary statistics.


Figure 1: Normalized eigenvalues of the double centered agreement score matrix. The elbow is a good indicator of the dimensionality of the space.

Finally, the last three rows report the improvement in APRE if we increase by one the dimensionality of the model.

The first question of interest is about the dimensionality of the political space. How many dimensions do we have? Figure 1 reports the plot of the normalized eigenvalues of the double centered agreement score matrix. A good way to understand the dimensionality of the political space, in fact, consists in locating the elbow in such a plot. As can been seen, it is very
following way:

$$
A P R E=\frac{\sum_{j}\{\text { minority vote }- \text { classification error }\}_{j}}{\sum_{j}\{\text { minority vote }\}_{j}}
$$

and it is equal to 0 when the model does not improve on the benchmark case, and 1 when the model achieves perfect classification. Note that APRE can also be negative, whenever the model generates more errors than the simple benchmark.

| Classification | H | Max spread | NH |
| :--- | :---: | :---: | :---: |
| TESEO | 0.088 | 0.0625 | 0.027 |
| OURS | 0.097 | 0.0625 | 0.037 |
| Clausen | 0.23 | 0.167 | 0.08 |
| Peltzman | 0.13 | 0.083 | 0.05 |

Table 2: Dispersion indexes for bills. H is Herfindahl index. Max spread is the minimum value for H . NH is the normalized Herfindahl index.
likely that the dimension is either two or three. In order to have a better understanding of the dimensionality, we look at the improvements in APRE when we increase by one the dimension. Table 1 shows that an increase from one to two dimensions improves APRE by 6.8 percentage points. Moving further to three dimensions only improves APRE by 1.1 percentage points. Thus we focus our analysis on a two-dimensional space.

Since the type of bills that are proposed and passed is, to a very large extent, endogenous to the legislative process, while we treat them as exogenous in our estimations, we decided to make sure that the low dimensionality is not just an artifact of this issue. To this end we coded the bills according to three main classification (Clausen, Peltzman, and TESEO) and one other classification we made on our own (OURS). TESEO is the classification of bills that is provided by the House itself. ${ }^{5}$ We then computed the Herfindahl concentration index for any of the above mentioned classifications, and then we normalized it in such a way that it ranges between 0 (maximum spread, that is even distribution, of the bills) and 1 (maximum concentration of bills in one category). ${ }^{6}$ Table 2 reports our findings for any classification of the bills we adopted. The worst possible case is if we look at the Clausen classification, which has the smallest number of categories. Even so, we have that NH is at $8 \%$, so that the variety of bills passed is just $8 \%$ far away from an even distribution. Therefore the low dimensionality in the Italian Parliament is not a mere artifact of the legislative process, and we can now move onto understanding the meaning of these two dimensions.

[^3]

Figure 2: Distribution of unconstrained bills by cutting line angles

Cutting line angles provide a useful way to understand which dimension is playing an important role in each roll call. In particular, any time the cutting line is (close to) vertical, we know that voting goes according to the first dimension. Similarly, when the cutting line is (close to) horizontal, we know that voting is mainly determined by the second dimension. Figure 2 reports the distribution of the unconstrained bills by cutting line angles. Unconstrained estimations are those where the cutting line angle is not forced to be on one of the edges of the space. As we can see, the mode of the distribution is represented by bills with vertical cutting lines. Despite the fact that the Italian political arena is characterized by a multiparty system, the XIII Legislatures offers us the possibility to compare the voting patterns within two main coalitions: Ulivo and Polo. Thus we define party-line votes whenever at least $90 \%$ of the Ulivo coalition voted against at least $90 \%$ of the Polo coalition. Figure 3 shows the distribution of the unconstrained bills with party-line votes by cutting line angles. As we can see, cutting line


Figure 3: Distribution of unconstrained bills with party-line vote by cutting line angles
angles are closer to vertical when party-line votes are present.
So, what do these dimensions mean? Figures 4 and 5 report the spatial maps obtained by, respectively, taking the mean and the median of each party member's coordinates. ${ }^{7}$ Like in the case for the U.S. Congress (see Poole and Rosenthal [10]) the first dimension is an expression of loyalty to the coalition the group belongs to. From left to right, we have groups more loyal to Ulivo, groups less loyal to either Ulivo or Polo, and groups loyal to the Polo. ${ }^{8}$ We tested this hypotheses by analyzing all the bills where party line vote was present. In particular, we computed the change in APRE if we move from one to two dimensions and we found that it is about $3.7 \%$, which confirms us that party line is the main determinant of the first dimension.

Under this perspective, it is not surprising that the Party of the Communist Reformation (PRC), a neo-communist party, is not at the far left of the first dimension. Their support for the Ulivo coalition has been pretty weak in the first years of the legislature, and has formally stopped when Prodi resigned from Prime Minister after losing a vote of no confidence from PRC themselves on October 9, 1998. In that period PRC also faced the departure of some of its members, who were more pro-Ulivo and decided to form a new neo-communist party, Party of the Italian Communists (PdCI). Their position in the first dimension is consistently to the left of PRC because, unlike the latter, they offered a support to the center left government that followed. Another interesting aspect of PRC comes by comparing the mean and the median position. In fact, the median member is more loyal to the Ulivo coalition than its average member, who nevertheless was more representative of the party stated position.

At the other extreme of the first dimension we have LN. A member of the center-right coalition during the XIII Legislature, LN caused the crisis of Mr Berlusconi's government which eventually led to early elections in which the this party ran alone against Roma-ladrona (Rome big thief) of both Roma-Polo and Roma-Ulivo. Thus in the XIII Legislature, LN represents the case of opposition to both coalitions, and its location to the right indicates its proximity to Polo more than Ulivo. In fact, during the current (XIV) Legislature, the LN belongs to the Polo which constitutes the majority coalition.

As for the second dimension, the first hypothesis that comes to mind

[^4]

Figure 4: Average coordinates by groups


Figure 5: Median coordinates by groups
by looking at the scatter plot is that it represents the North - South spectrum. Parties whose electoral support was greater in the North than in the other parts of the peninsula are located near the upper end of the second dimension. Parties whose success was mostly due to the Southern votes are located near the lower end of the second dimension, while the parties which had the best electoral performance in Center Italy, are located in between the other two clusters of parties (Cartocci [2][page 195]). We computed the correlation of each party's percentage total vote at Northern Italian regions with the party's second dimension, and we found that sign is positive (r $=.402$ ). Similarly, we computed the correlation of each party's percentage total vote at the Southern Italian regions with the party's second dimension and we found almost identical results but with the (expected) opposite sign ( $\mathrm{r}=-.395$ ). Such a territorial component could emerge from bills focused on Southern areas (which belong to category 7 of TESEO). Unfortunately there are not enough observations to look at changes in APRE from one to two dimensions because of regional factors. On the other hand, the territorial factor could as well emerged from bills related to constitutional matters. In fact, the territorial cleavage was a heated issue for most of the XIII Legislature not only because LN was advocating a secession of the Northern regions, had created a Parliament for Padania and had held mock elections for the Padanian parliament, but also because the reform of the form of state was one of the main issues of the reformist agenda of the Bicameral Committee for the constitutional reforms presided over by D'Alema. The debates as to how to reform the Italian form of state opposed the parties in favor of preserving the unitary form of state and were generally stronger in central and southern Italy to the parties advocating a reform of the form of state and were stronger in the Northern Regions. The fact that the Bicameral Commission basically ended up as a big failure, since no agreement was found, led us think as the possible correlation between territorial and constitutional interests. Our analysis reveal actually that constitutionalism provides the best explanation for the change in the dimensionality of the political space in the Italian parliament and, indirectly, for our interpretation of the second dimension as territorial. In fact, to test this hypothesis we once again looked at the change in APRE if we move from one to two dimensions by restring our attention to bills within the same category in each of the bills classifications. In particular, we restricted our attention to only those categories with at least 10 roll calls and changes in APRE above $20 \%$. Not surprisingly we find that neither the Clausen nor the Peltzman categories fit well the Italian case. With the TESEO classification, we find that the most significative changes in APRE are related to the following
categories: 1 (Constitutional matters), 4 (Health), 9 (Labor), and 11 (Productive activities). With OURS classification, we have that only category 1 (Constitutional matters) is relevant. (Tables 8 through 10 in the appendix show our findings.)

Similarly, we looked at the changes in APRE by moving to a two dimensional model in the cases were cutting line angles were close to either 0 (namely less than 20) or 180 (namely greater than 160) degrees. Those cases are rolls calls were the second dimension plays a major role. In fact the changes in APRE are huge. We then looked at the distribution of categories of bills in this set. We found that Constitutional matters is one of them, but not the only one. In OURS, all the bills who represent ratifications, delegations, extensions of deadline are even more numerous. Figures 11 and 12 in the appendix report the distributions of bills according to TESEO and OURS classification whose cutting line angle is almost flat.

## 5 Party cohesion within the House.

A second measure of interest we can obtain from individual MPs' coordinates is the groups' standard deviations, which we take as a measure of group's cohesion. We think this is an interesting alternative measure of party cohesion because it comes directly from a structural model of congressional voting. In fact, as Desposato [5] points out, standard cohesion scores do not always provide accurate estimates of parties cohesion, just because they lack an underlying model of congressional voting. As an example, he shows that in a random utility model of legislative behavior, those standard scores tend to overestimate the cohesion of small parties. Another reason of interest of our measure of cohesion lies in the fact that dispersion scores account for the dimensionality of politics and political behavior. Hence they do not only allow the analyst to measure parties' cohesion, but they also allow the analyst to assess party cohesion on each of the dimensions of a given political space and, therefore, to investigate whether and to what extent cohesion is affected by the nature of the political issues that are relevant on each dimension.

By performing our analysis of party's cohesion, we found that the Italian parties differed from each other not only because they occupy different positions in the political space (average of the positions occupied by the various MPs belonging to a party) but also because they display different levels of concentration/dispersion (standard deviation). While some parties like the

|  | $\Delta X 1$ | $\Delta X 2$ | Size |
| :---: | :---: | :---: | :---: |
| $\Delta X 1$ | 1.000 |  |  |
| $\Delta X 2$ | $0.635^{*}$ | 1.000 |  |
|  | $(0.005)$ |  |  |
| Size | -0.196 | -0.121 | 1.000 |
|  | $(0.437)$ | $(0.633)$ |  |

Table 3: Correlations between dispersions on each dimension and size. Asterisk indicates value that is significant at 0.05 level or better. Values in brackets are the significance levels.

Democrats of the Left (DS) or PdCI display low levels of dispersion, others like the Segni Pact or the Union of the Democrats for Europe (UDEUR) display much higher levels. So parties have different degrees of homogeneity. The first factor affecting parties' homogeneity is represented by the nature of political issues or the nature of the political dimension. Let's go Italy (FI),UDEUR, Mixed Group, Italian renewal (RI) and the Segni Pact are more homogeneous on the first dimension, while the all the other parties are more homogeneous on the second dimension. In general we find that dispersion along the first dimension is positively correlated with dispersion along the second dimension, but that dispersion along either dimension is not significantly correlated with the size of the party. (See table 3) We can push the analysis a step further and ask: what factors are related to parties' dispersion on the two dimensions? In order to answer this question, we correlate the estimates concerning parties' dispersion on each axis with three measures of cohesion, namely the Rice index, the Agreement Index (AI) proposed by Hix, Noury and Roland [6], and the Modified Agreement Index (MAI), proposed by us. The basic difference between these three measures of cohesion concerns the number of voting options that are available to MPs. There are two voting options (yes, no) considered by the Rice index, there are three voting options (yes, no, abstain) considered by AI, and there are four voting options (yes, no, abstain, absence) considered by MAI. ${ }^{9}$ The

[^5]measures of party cohesion and on parties' dispersion on both axes are presented in table 4. The need to develop this new index of cohesion resulted from the fact that the number of MPs abstaining if fairly small, the number of MPs being absent is fairly high. This form of legislative behavior cannot be simply explained by the fact that individual MPs have something else or more important to do. The analysis of the votes on the final bills in the Italian chamber of deputies reveals that for some bills the MPs of all the major opposition parties (LN, FI, and National Alliance, AN) either stayed home or abstained from voting, thus lowering the quorum needed to approve the bills. One may therefore ask oneself whether this form of behavior is an explicit support from the opposition members to a divided, undisciplined government coalition, to pass some crucial bills in the Legislature. Other hypothesis may be of interest as well. Mass absence can, for example, be a sign of protest against the bill being voted on, or even a (failed) attempt to avoid meeting the quorum. Given the possible strategic significance of absence and abstention, these two phenomena should be considered as one of the voting options, and should hence be taken into consideration to properly measure parties' cohesion. Having measured parties' cohesion on the basis of these there indexes, we correlate them with parties' dispersion on each dimension.

By performing these two sets of analysis, we find that the correlation coefficients are all properly signed and suggest that parties' cohesion declines when parties become more dispersed. Furthermore, we find that the correlation between dispersion on each dimension is not significantly (from a statistical point of view) related to two measures of cohesion. Dispersion on the first dimension relates significantly only to the Index of Agreement, while the dispersion on the second dimension relates significantly only to the MAI index. This finding is relevant for three reasons. First because it shows that the party dispersion on the two axes relates to different factors. On the first dimension, which basically captures the government-opposition conflict, parties' dispersion relates to how MPs vote when they are present. On the

The Agreement index is:

$$
A I_{j}=\frac{1}{m} \sum_{i=1}^{m} \frac{\max \left\{s_{i j}, n_{i j}, a_{i j}\right\}-1 / 2\left(s_{i j}+n_{i j}+a_{i} j-\max \left\{s_{i j}, n_{i j}, a_{i j}\right\}\right)}{s_{i j}+n_{i j}+a_{i j} j} ;
$$

The Modified Agreement index is:
$M A I_{j}=\frac{1}{m} \sum_{i=1}^{m} \frac{\max \left\{s_{i j}, n_{i j}, a_{i j, h_{i j}}\right\}-1 / 3\left(s_{i j}+n_{i j}+a_{i j}+h_{i j}-\max \left\{s_{i j}, n_{i j}, a_{i j}, h_{i j}\right\}\right)}{s_{i j}+n_{i j}+a_{i} j+h_{i j}}$.

Table 4: Agreement indexes for each party and dispersions along each dimension. FI $=$ Let's go Italy; LN $=$ Northern League; DS $=$ Democrats of the Left; $\mathrm{DE}=$ Democrats; $\mathrm{PP}=$ Popular Party; UDEUR $=$ Union of the Democrats for Europe; MISTO = mixed; SDI = Italian Socialist Democrats; $\mathrm{CCD}=$ Christian Democratic Center; RI $=$ Italian Renewal; PRC $=$ Party of the Communist Reformation; SEGNI $=$ Segni Pact; ML $=$ Linguistic Minorities; VERDI $=$ Greens; FLDR $=$ Federation of Liberals, Democratic and Republicans; PDCI = Party of the Italian Communists; AN $=$ National Alliance; UDR/CDU $=$ Christian Democratic United.

| PARTY | RICE | AGREEMENT | MAI | DISPERSION |  |
| :--- | ---: | ---: | ---: | ---: | ---: |
| ON X1 |  |  |  |  |  | DISPERSION ON X2


|  | Rice | Agreement | MAI |
| :--- | :---: | :---: | :---: |
| $\Delta X 1$ | -.432 | -.606 | -.128 |
|  | $(.073)$ | $(.008)$ | $(.612)$ |
| $\Delta X 2$ | -.453 | -.422 | -.519 |
|  | $(.059)$ | $(.081)$ | $(.027)$ |

Table 5: Correlations between dispersion of coordinates and agreement indexes.
second dimension instead, which captures the Constitutional conflict, parties' dispersion relates to a metric that accounts not only for how MPs vote when they're present but also for whether MPs are actually present. Second this finding is relevant because it further corroborates, however indirectly, our understanding of the second dimension as a constitutional dimension. In fact, the extent to which parties are concentrated or dispersed is affected also by parties' absence rate, which provides a good indication of the importance parties attach to the work of parliament and, more broadly, to the functioning of the existing political institutions and constitutional order. Finally, this finding is important because it highlights the importance of absence as a mode of legislative behavior, a topic that probably deserves more attention than what has received so far.

## 6 Conclusion

This paper offers a spatial of the Italian House during the XIII legislature. We found that the Italian political space is virtually two-dimensional. The first dimension is related to the loyalty to either coalition, with Ulivo to the left of Polo. Thus, this dimension looses its typical ideological connotation it may have had in the first XII Legislatures, (where the electoral system was purely proportional), and it becomes more a measure of the degree with which the dialectic between government and opposition operates. Government parties would be located the extreme left of this dimension (as Ulivo was in power). The opposition parties would be to the right of the Ulivo's parties. The further away, the harsher the opposition to the government. Thus the position of LN reflects the least compromising one within the opposing parties.

The second dimension order parties according to the Constitutional is-
sues. From top to bottom we have parties whose position towards a more centralized form of government is increasingly stronger. In addition, we find that the distribution of parties along the second dimension is correlated with the geographic nature of parties: from top to bottom we have parties whose support is mainly in Northern Italy, in the Red Belt and in Southern Italy.

Last, the overall eccentric position of LN is consistent with the antisystem role played by this party during the whole legislature.

We also found that intra-party homogeneity varies from party to party and that this variance is due to different factors. Interestingly enough, party cohesion may not be positively correlated with size. If it is true that, ceteris paribus, it is easier to enforce party discipline in smaller parties, it is also true that the benefit from doing that is bigger in larger parties. In other words, MPs in smaller parties may be a better bargaining in the process of votes acquisition, and this may result in higher heterogeneity for smaller parties. Further research might show the extent to which our findings are related to the groups' compositions and the House procedural rules. Yet we believe that intra-party homogeneity is worth studying and may shed some additional light on the life of parliamentary parties.

Finally, the results of our analysis underline the importance, at least in the Italian context, of the absence from voting sessions as mode of legislative behavior. During the XIII Legislatures, absences are not only caused by opportunity cost considerations. Absence is strategic. Some of the bills passed because or despite all the members of the opposition parties stayed home or abstained. So the questions become: are we facing a sort of agreement between the two main forces (Polo and Ulivo) so that the opposition stays home to lower the quorum over a divided government? Are we facing a harsh opposition between the two main forces, so that one stays home to take as much distance as possible (and hopefully avoiding the quorum to be met) towards some bills? And in the first case, how should this cross-coalitional courtesy be understood? Is it a sign or responsibility or collusion? And even assuming that it is a sign of responsibility, how much should the responsible party government framework be modified to account for this type of party politics? This paper cannot provide an adequate answer for all these questions, but it provides a strong suggestion for where future research should be going.

## A Tables and charts

In this section we report tables and charts of our findings. Table 6 report the classification in TESEO. The code number in the second column indicates how we aggregate those bills into 16 categories. Table 7 reports OURS classification.

Tables 8 through 10 report the change in APRE from one to two and from two to three dimensions by bill according to the four classifications we have analyzed. Last, tables 11 and 12 report the distribution of bills whose cutting line angle is almost horizontal.

## B Codebook and estimated coordinates

Tables 13 through 6 report individual MPs coordinates.

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Table 6: TESEO classification of bills.
TESEO classification Code
Constitutional matters ..... 1
Foreign affairs ..... 2
European affairs ..... 15
Agriculture ..... 3
Environment ..... 8
Social security and welfare ..... 9
Banks, credit and money ..... 5
Budget ..... 5
Stock exchange and financial activities ..... 5
Foreign trade ..... 2
Consumers protection ..... 11
Culture, entertainment, sport and tourism ..... 12
National defense and army ..... 6
Civil liberties and civil rights ..... 1
Commercial and corporate law ..... 11
Law and justice ..... 16
Energy ..... 11
Family and childhood ..... 7
Local and regional public finance ..... 10
Public finance and taxes ..... 5
Industry and craftsmanship ..... 11
Information and communication ..... 13
Public works and housing ..... 14
Southern Italy and depressed areas ..... 7
Labor and employment ..... 9
Public order and police forces ..... 16
Public economy and privatizations ..... 5
Public sector ..... 9
Regions and local autonomies ..... 10
Health sector ..... 4
Education and research ..... 12
Transportations ..... 13
Unions and workers rights ..... 9

Table 7: OURS Bill classifications.

| OURS | Code |
| :--- | ---: |
| Constitutional matters | 1 |
| Law and justice | 2 |
| Foreign affairs | 3 |
| National defense and army | 4 |
| Budget, public finance and taxes | 5 |
| Public expenditure | 6 |
| Education and research | 7 |
| Environment | 8 |
| Transportation | 9 |
| Health | 10 |
| Labor | 11 |
| Welfare ans social policy | 12 |
| Agricolture | 13 |
| Delegation, ratification, extension of terms | 14 |
| Information and communication | 15 |
| Internal affairs | 16 |
| Others | 0 |

Table 8: Changes in APRE in OUR classification.

| TESEO | APRE(2) - APRE(1) | APRE(3) - APRE(2) | SIZE |
| ---: | ---: | ---: | ---: |
| 2 | -0.098 | 0.040 | 49 |
| 5 | 0.200 | 0.018 | 40 |
| 9 | 0.286 | 0.033 | 37 |
| 16 | 0.019 | 0.048 | 30 |
| 12 | 0.198 | 0.021 | 29 |
| 6 | 0.199 | -0.436 | 20 |
| 13 | 0.160 | 0.003 | 20 |
| 1 | 0.317 | 0.039 | 18 |
| 4 | 0.214 | 0.075 | 15 |
| 11 | 0.238 | 0.082 | 12 |
| 14 | 0.190 | 0.008 | 12 |
| 3 | 0.187 | 0.098 | 11 |
| 15 | -0.235 | -0.090 | 10 |
| 8 | -0.151 | 0.185 | 7 |
| 10 | 0.297 | 0.042 | 6 |
| 7 | -0.042 | 0.009 | 4 |

Table 9: Changes in APRE in TESEO classification.

| MASSI | APRE(2) - APRE(1) | APRE(3) - APRE(2) | SIZE |
| ---: | ---: | ---: | ---: |
| 14 | 0.165 | 0.067 | 45 |
| 5 | 0.070 | 0.024 | 33 |
| 1 | 0.284 | 0.041 | 20 |
| 3 | 0.189 | 0.042 | 20 |
| 2 | 0.014 | 0.019 | 15 |
| 16 | 0.026 | 0.011 | 15 |
| 7 | 0.056 | 0.016 | 12 |
| 9 | 0.087 | 0.036 | 9 |
| 10 | -0.005 | 0.126 | 9 |
| 12 | 0.010 | 0.039 | 9 |
| 13 | 0.041 | 0.000 | 9 |
| 6 | 0.059 | 0.016 | 8 |
| 4 | 0.190 | -1.383 | 7 |
| 11 | 0.295 | 0.110 | 7 |
| 15 | 0.354 | -0.193 | 5 |
| 8 | 0.097 | 0.050 | 4 |

Table 10: Changes in APRE in Peltzman and in Clausen classifications.

| PELTZMAN APRE(2) - APRE(1) | APRE(3) - APRE(2) | SIZE |  |
| ---: | ---: | ---: | ---: |
| 1 | 0.110 | -0.017 | 25 |
| 2 | 0.103 | 0.018 | 32 |
| 3 | 0.184 | 0.039 | 32 |
| 4 | 0.162 | 0.042 | 25 |
| 5 | 0.085 | 0.006 | 15 |
| 8 | 0.130 | 0.000 | 58 |
| 9 | 0.000 | 0.357 | 1 |
| 10 | -0.035 | -0.028 | 1 |
| 61 | 0.246 | 0.026 | 3 |
| 62 | 0.152 | 0.029 | 13 |
| 71 | 0.018 | -0.221 | 9 |
| 72 | 0.099 | 0.000 | 35 |
|  |  |  |  |
| CLAUSEN | APRE(2) - APRE(1) | APRE(3) - APRE(2) | SIZE |
| 1 | 0.103 | 0.022 | 85 |
| 5 | 0.154 | -0.174 | 58 |
| 6 | 0.162 | 0.050 | 41 |
| 2 | 0.098 | 0.033 | 36 |
| 3 | 0.001 | 0.089 | 12 |
| 4 | 0.023 | 0.036 | 10 |

Table 11: Distribution of bills according to TESEO for bills whose cutting line angle is almost flat.


Table 12: Distribution of bills according to OURS whose cutting line angle is almost flat.


Table 13: Individual MPs coordinates.

| Fullname | Code | X1 | X2 | Fullname | Code | X1 | X2 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| ABATERUSSO ERNESTO | 1 | -0.758 | -0.085 | BERLINGUER LUIGI | 51 | -0.776 | -0.102 |
| ABBATE MICHELE | 2 | -0.794 | -0.041 | BERRUTI MASSIMO | 53 | 0.193 | -0.107 |
| ABBONDANZIERI MARISA | 3 | -0.779 | -0.321 | BERSELLI FILIPPO | 54 | 0.3 | -0.329 |
| ACCIARINI MARIACHIARA | 4 | -0.786 | -0.079 | BERTINOTTI FAUSTO | 55 | -0.569 | 0.822 |
| ACIERNO ALBERTO | 5 | -0.087 | -0.648 | BERTUCCI MAURIZIO | 56 | 0.218 | -0.161 |
| ACQUARONE LORENZO | 6 | -0.741 | -0.12 | BIANCHI GIOVANNA | 57 | 0.788 | 0.616 |
| AGOSTINI MAURO | 7 | -0.757 | -0.048 | BIANCHI GIOVANNI | 58 | -0.748 | -0.091 |
| ALBANESE VALERIA | 8 | -0.798 | -0.085 | BIANCHI VINCENZO | 59 | 0.226 | -0.202 |
| ALBERTINI GIUSEPPE | 9 | -0.809 | -0.194 | BIASCO SALVATORE | 60 | -0.74 | -0.052 |
| ALBONI ROBERTO | 10 | 0.344 | -0.585 | BICOCCHI GIUSEPPE | 61 | -0.557 | -0.063 |
| ALBORGHETTI DIEGO | 11 | 0.857 | 0.516 | BIELLI VALTER | 62 | -0.752 | -0.073 |
| ALEFFI GIUSEPPE | 12 | 0.248 | -0.044 | BINDI ROSY | 63 | -0.821 | -0.199 |
| ALEMANNO GIOVANNI | 13 | 0.303 | -0.34 | BIONDI ALFREDO | 64 | 0.234 | -0.239 |
| ALOI FORTUNATO | 14 | 0.316 | -0.532 | BIRICOTTI ANNAMARIA | 65 | -0.782 | -0.077 |
| ALOISIO FRANCESCO | 15 | -0.767 | -0.069 | BOATO MARCO | 67 | -0.898 | 0.223 |
| ALTEA ANGELO | 16 | -0.747 | -0.219 | BOCCHINO ITALO | 68 | 0.303 | -0.405 |
| ALVETI GIUSEPPE | 17 | -0.756 | -0.085 | BOCCIA ANTONIO | 69 | -0.685 | -0.09 |
| AMATO GIUSEPPE | 18 | 0.278 | -0.236 | BOGHETTA UGO | 70 | -0.468 | 0.884 |
| AMORUSO FRANCESCO | 19 | 0.227 | -0.739 | BOGI GIORGIO | 71 | -0.751 | 0.122 |
| ANDREATTA BENIAMINO | 20 | -0.81 | -0.227 | BOLOGNESI MARIDA | 72 | -0.765 | -0.095 |
| ANEDDA GIANFRANCO | 21 | 0.345 | -0.323 | BONAIUTI PAOLO | 73 | 0.223 | -0.113 |
| ANGELICI VITTORIO | 22 | -0.769 | 0.086 | BONATO FRANCESCO | 74 | -0.524 | 0.852 |
| ANGELINI GIORDANO | 23 | -0.715 | -0.097 | BONITO FRANCESCO | 75 | -0.71 | -0.116 |
| ANGELONI BERARDINO | 24 | 0.11 | -0.866 | BONO NICOLA | 76 | 0.3 | -0.52 |
| ANGHINONI UBER | 25 | 0.843 | 0.538 | BORDON WILLER | 77 | -0.785 | -0.212 |
| APOLLONI DANIELE | 26 | 0.509 | 0.352 | BORGHEZIO MARIO | 78 | 0.814 | 0.568 |
| APREA VALENTINA | 27 | 0.212 | -0.205 | BORROMETI ANTONIO | 79 | -0.741 | -0.084 |
| ARACU SABATINO | 28 | 0.155 | -0.178 | BOSCO RINALDO | 80 | 0.711 | 0.703 |
| ARMANI PIETRO | 29 | 0.382 | -0.43 | BOSELLI ENRICO | 81 | -0.791 | -0.023 |
| ARMAROLI PAOLO | 30 | 0.35 | -0.606 | bova domenico | 83 | -0.768 | -0.097 |
| ARMOSINO MARIATERESA | 31 | 0.268 | -0.2 | BRACCO FELICE | 84 | -0.769 | -0.063 |
| ASCIERTO FILIPPO | 32 | 0.307 | -0.362 | BRANCATI ALDO | 85 | -0.761 | -0.084 |
| ATTILI ANTONIO | 33 | -0.759 | -0.084 | BRESSA GIANCLAUDIO | 86 | -0.762 | -0.098 |
| BACCINI MARIO | 34 | 0.175 | -0.195 | BRUGGER SIEGFRIED | 87 | -0.8 | 0.166 |
| BAGLIANI LUCA | 35 | 0.625 | 0.439 | BRUNALE GIOVANNI | 88 | -0.782 | -0.085 |
| BAIAMONTE GIACOMO | 36 | 0.193 | -0.112 | BRUNETTI MARIO | 89 | -0.859 | 0.511 |
| BALLAMAN EDOUARD | 37 | 0.773 | 0.635 | BRUNO DONATO | 90 | 0.235 | -0.18 |
| BALOCCHI MAURIZIO | 38 | 0.831 | 0.557 | BRUNO EDUARDO | 91 | -0.905 | 0.426 |
| BAMPO PAOLO | 39 | 0.758 | 0.537 | BUFFO GLORIA | 92 | -0.76 | -0.067 |
| BANDOLI FULVIA | 40 | -0.817 | -0.081 | BUGLIO SALVATORE | 93 | -0.743 | 0.008 |
| BARBIERI ROBERTO | 41 | -0.794 | -0.031 | BUONTEMPO TEODORO | 94 | 0.529 | -0.391 |
| BARRAL LUCIO | 42 | 0.676 | 0.493 | BURANI MARIA | 95 | 0.222 | -0.19 |
| BARTOLICH ADRIA | 43 | -0.771 | -0.051 | BURLANDO CLAUDIO | 96 | -0.823 | -0.154 |
| BASSO MARCELLO | 44 | -0.773 | -0.081 | BUTTI ALESSIO | 97 | 0.444 | -0.584 |
| BASTIANONI STEFANO | 45 | -0.495 | -0.286 | BUTTIGLIONE ROCCO | 98 | -0.172 | -0.246 |
| BATTAGLIA AUGUSTO | 46 | -0.776 | -0.082 | CACCAVARI ROCCO | 99 | -0.795 | -0.066 |
| BECCHETTI PAOLO | 47 | 0.236 | -0.161 | CALDERISI GIUSEPPE | 100 | 0.287 | 0.076 |
| BENEDETTI DOMENICO | 48 | 0.445 | -0.468 | CALDEROLI ROBERTO | 101 | 0.86 | 0.511 |
| BENVENUTO GIORGIO | 49 | -0.759 | -0.088 | CALZAVARA FABIO | 102 | 0.737 | 0.407 |
| BERGAMO ALESSANDRO | 50 | 0.228 | -0.194 | Calzolaio valerio | 103 | -0.811 | -0.168 |

Table 14: Individual MPs coordinates - Continued.

| Fullname | Code | X1 | X2 | Fullname | Code | X1 | X2 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| CAMBURSANO RENATO | 104 | -0.763 | 0.08 | COLOMBO FURIO | 154 | -0.767 | 0.026 |
| CAMOIRANO MAURA | 105 | -0.768 | -0.097 | COLOMBO PAOLO | 155 | 0.853 | 0.464 |
| CAMPATELLI VASSILI | 106 | -0.755 | 0.027 | colosimo elio | 157 | 0.317 | -0.471 |
| CANANZI RAFFAELE | 107 | -0.747 | -0.048 | COLUCCI GAETANO | 158 | 0.332 | -0.496 |
| CANGEMI LUCA | 108 | -0.493 | 0.87 | COMINO DOMENICO | 159 | 0.531 | 0.848 |
| CAPARINI DAVIDE | 109 | 0.825 | 0.438 | CONTE GIANFRANCO | 160 | 0.279 | -0.19 |
| CAPITELLI PIERA | 110 | -0.735 | -0.033 | CONTENTO MANLIO | 161 | 0.39 | -0.541 |
| CAPPELLA MICHELE | 111 | -0.747 | -0.093 | CONTI GIULIO | 162 | 0.391 | -0.54 |
| CARAZZI MARIA | 112 | -0.854 | 0.456 | COPERCINI PIERLUIGI | 163 | 0.842 | 0.54 |
| CARBONI FRANCESCO | 113 | -0.782 | 0.075 | CORDONI ELENA | 164 | -0.75 | 0.07 |
| CARDIELLO FRANCO | 114 | 0.367 | -0.391 | CORLEONE FRANCO | 165 | -0.789 | -0.046 |
| CARDINALE SALVATORE | 115 | -0.189 | -0.451 | CORSINI PAOLO | 166 | -0.809 | -0.001 |
| CARLESI NICOLA | 116 | 0.395 | -0.582 | CORVINO MICHELE | 167 | -0.762 | -0.269 |
| CARLI CARLO | 117 | -0.788 | -0.067 | COSENTINO NICOLA | 168 | 0.289 | -0.113 |
| CAROTTI PIETRO | 118 | -0.76 | -0.075 | COSSUTTA ARMANDO | 169 | -0.78 | 0.615 |
| CARRARA CARMELO | 119 | 0.174 | -0.307 | COSSUTTA MAURA | 170 | -0.895 | 0.447 |
| CARRARA NUCCIO | 120 | 0.237 | -0.187 | COSTA RAFFAELE | 171 | 0.353 | 0.074 |
| CARTA CLEMENTE | 121 | -0.848 | -0.529 | COVRE GIUSEPPE | 172 | 0.873 | 0.387 |
| CARUANO GIOVANNI | 122 | -0.743 | -0.127 | CREMA GIOVANNI | 173 | -0.792 | 0.099 |
| CARUSO ENZO | 123 | 0.426 | -0.405 | CRIMI ROCCO | 174 | 0.22 | -0.1 |
| CASCIO FRANCESCO | 124 | 0.271 | -0.205 | CRUCIANELLI FAMIANO | 175 | -0.773 | -0.082 |
| CASILLI COSIMO | 125 | -0.754 | -0.388 | Cuccu Paolo | 176 | 0.228 | -0.11 |
| CASINELLI CESIDIO | 126 | -0.758 | -0.083 | CUSCUNA' ANTONIO | 177 | 0.282 | -0.547 |
| CASINI PIERFERDINANDO | 127 | 0.192 | -0.01 | CUTRUFO MAURO | 178 | -0.65 | 0.226 |
| CASTELLANI GIOVANNI | 128 | -0.757 | -0.074 | D'ALEMA MASSIMO | 179 | -0.803 | -0.181 |
| CAVALIERE ENRICO | 129 | 0.829 | 0.559 | D'ALIA SALVATORE | 180 | 0.153 | -0.129 |
| CAVANNA MARIELLA | 130 | -0.156 | -0.73 | DALLACHIESA NANDO | 181 | -0.909 | 0.133 |
| CAVERI LUCIANO | 131 | -0.749 | 0.1 | DALLAROSA FIORENZO | 182 | 0.787 | 0.617 |
| CE' ALESSANDRO | 132 | 0.784 | 0.621 | DAMERI SILVANA | 183 | -0.741 | -0.101 |
| CENNAMO ALDO | 133 | -0.756 | -0.083 | D'Amico natale | 184 | -0.767 | 0.145 |
| CENTO PIER PAOLO | 134 | -0.901 | 0.434 | DANESE LUCA | 185 | -0.014 | -0.713 |
| CEREMIGNA ENZO | 135 | -0.708 | 0.118 | DANIELI FRANCO | 186 | -0.93 | 0.074 |
| CERULLI VINCENZO | 136 | -0.77 | -0.07 | DEBENETTI LINO | 187 | -0.921 | 0.233 |
| CESARO LUIGI | 137 | 0.202 | -0.184 | DECESARIS WALTER | 188 | -0.501 | 0.865 |
| CESETTI FABRIZIO | 138 | -0.743 | -0.089 | DEFRANCISCIS FERDINANDO | 189 | -0.039 | -0.43 |
| CHERCHI SALVATORE | 139 | -0.787 | -0.07 | DEGHISLANZONI GIACOMO | 190 | 0.239 | -0.107 |
| CHIAMPARINO SERGIO | 140 | -0.758 | -0.061 | DELUCA ANNAMARIA | 191 | 0.253 | -0.19 |
| CHIAPPORI GIACOMO | 141 | 0.844 | 0.469 | DEMITA CIRIACO | 192 | -0.757 | 0.319 |
| CHIAVACCI FRANCESCA | 142 | -0.765 | -0.035 | DEMURTAS GIOVANNI | 193 | -0.885 | 0.465 |
| CHINCARINI UMBERTO | 143 | 0.817 | 0.569 | DEPICCOLI CESARE | 194 | -0.758 | -0.078 |
| CHIUSOLI FRANCO | 144 | -0.756 | -0.05 | DESIMONE ALBERTA | 195 | -0.771 | -0.052 |
| CIANI FABIO | 145 | -0.753 | -0.082 | DEBIASIO LUISA | 196 | -0.758 | -0.083 |
| CIAPUSCI ELENA | 146 | 0.819 | 0.453 | DEDONI ANTONINA | 197 | -0.758 | -0.09 |
| CICU SALVATORE | 147 | 0.23 | -0.134 | DELBARONE GIUSEPPE | 198 | 0.168 | -0.353 |
| CIMADORO GABRIELE | 148 | -0.079 | -0.997 | DELBONO EMILIO | 199 | -0.735 | -0.117 |
| CITO GIANCARLO | 149 | 0.471 | -0.249 | DELFINO LEONE | 200 | -0.623 | 0.16 |
| COLA SERGIO | 150 | 0.367 | -0.558 | DELFINO TERESIO | 201 | 0.094 | -0.117 |
| COLLAVINI MANLIO | 151 | 0.233 | -0.09 | DELL'ELCE GIOVANNI | 202 | 0.204 | -0.205 |
| COLLETTI LUCIO | 152 | 0.317 | 0.019 | DELL'UTRI MARCELLO | 203 | 0.261 | -0.261 |
| COLOMBINI EDRO | 153 | 0.229 | -0.186 | DELMASTRO SANDRO | 204 | 0.475 | -0.644 |

Table 15: Individual MPs coordinates - Continued.

| Fullname | Code | X1 | X2 | Fullname | Code | X1 | X2 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| DEODATO GIOVANNI | 205 | 0.187 | -0.127 | FRIGERIO CARLO | 256 | 0.756 | 0.4 |
| DETOMAS GIUSEPPE | 206 | -0.706 | 0.002 | FRONZUTI GIUSEPPE | 257 | -0.117 | -0.487 |
| DIBISCEGLIE ANTONIO | 207 | $-0.757$ | -0.084 | FROSIO LUCIANA | 258 | 0.732 | 0.463 |
| DICAPUA FABIO | 208 | -0.746 | 0.108 | FUMAGALLI MARCO | 259 | -0.768 | -0.079 |
| DICOMITE FRANCESCO | 209 | 0.169 | -0.14 | FUMAGALLI SERGIO | 260 | -0.716 | 0.089 |
| DIFONZO GIOVANNI | 210 | -0.684 | 0.145 | GAETANI ROCCO | 261 | -0.816 | -0.069 |
| DILUCA ALBERTO | 211 | 0.281 | -0.149 | GAGLIARDI ALBERTO | 262 | 0.318 | -0.286 |
| DINARDO ANIELLO | 212 | -0.098 | -0.727 | GALATI GIUSEPPE | 263 | 0.216 | -0.049 |
| dirosa roberto | 213 | -0.767 | -0.06 | GALDELLI PRIMO | 264 | -0.833 | 0.486 |
| DISTASI GIOVANNI | 214 | -0.767 | -0.064 | GALEAZZI ALESSANDRO | 265 | 0.338 | -0.402 |
| DILIBERTO OLIVIERO | 215 | -0.836 | 0.549 | GALLETTI PAOLO | 266 | -0.976 | 0.217 |
| D'IPPOLITO IDA | 217 | 0.243 | -0.12 | GALLI DARIO | 267 | 0.737 | 0.676 |
| DIVELLA GIOVANNI | 218 | 0.245 | -0.185 | GAMBALE GIUSEPPE | 268 | -0.922 | -0.081 |
| DOMENICI LEONARDO | 219 | -0.762 | -0.027 | GAMBATO FRANCA | 269 | 0.66 | 0.524 |
| DONNER LUCIANO | 220 | 0.892 | 0.373 | GARDIOL GIORGIO | 270 | -0.95 | 0.312 |
| DOZZO GIANPAOLO | 221 | 0.818 | 0.464 | GARRA GIACOMO | 271 | 0.237 | -0.175 |
| DUCA EUGENIO | 222 | -0.797 | -0.092 | GASPARRI MAURIZIO | 272 | 0.331 | -0.345 |
| DUILIO LINO | 223 | -0.729 | 0.155 | GASPERONI PIETRO | 273 | -0.758 | -0.085 |
| DUSSIN GUIDO | 224 | 0.811 | 0.585 | GASTALDI LUIGI | 274 | 0.241 | -0.171 |
| DUSSIN LUCIANO | 225 | 0.831 | 0.556 | GATTO MARIO | 275 | -0.738 | -0.17 |
| ERRIGO DEMETRIO | 226 | -0.171 | -0.231 | GAZZARA ANTONINO | 276 | 0.177 | -0.076 |
| EVANGELISTI FABIO | 227 | -0.803 | -0.058 | GAZZILLI MARIO | 277 | 0.212 | -0.101 |
| FABRIS MAURO | 228 | 0.014 | -0.498 | GERARDINI FRANCO | 278 | -0.769 | -0.08 |
| FAGGIANO COSIMO | 229 | -0.77 | -0.083 | GIACALONE SALVATORE | 279 | -0.745 | -0.081 |
| FANTOZZI AUGUSTO | 230 | -0.777 | -0.185 | GIACCO LUIGI | 280 | -0.721 | -0.159 |
| FASSINO PIERO | 231 | -0.753 | -0.147 | GIANNATTASIO PIETRO | 281 | 0.239 | -0.22 |
| FAUSTINELLI ROBERTO | 232 | 0.776 | 0.429 | GIANNOTTI VASCO | 282 | -0.741 | -0.031 |
| FEI SANDRA | 233 | 0.371 | -0.442 | GIARDIELLO MICHELE | 283 | -0.729 | -0.111 |
| FERRARI FRANCESCO | 234 | -0.766 | -0.085 | GIORDANO FRANCESCO | 284 | -0.467 | 0.884 |
| FILOCAMO GIOVANNI | 235 | 0.351 | -0.317 | GIORGETTI ALBERTO | 285 | 0.361 | -0.542 |
| FINI GIANFRANCO | 236 | 0.273 | -0.269 | GIORGETTI GIANCARLO | 286 | 0.798 | 0.589 |
| FINO FRANCESCO | 237 | 0.343 | -0.453 | GIOVANARDI CARLO | 287 | 0.18 | -0.113 |
| FINOCCHIARO ANNA | 238 | -0.734 | -0.202 | GIOVINE UMBERTO | 288 | 0.235 | 0.022 |
| FIORI PUBLIO | 239 | 0.408 | -0.604 | GISSI ANDREA | 289 | 0.291 | -0.288 |
| FIORONI GIUSEPPE | 240 | -0.759 | -0.089 | GIUDICE GASPARE | 290 | 0.19 | -0.09 |
| FLORESTA ILARIO | 241 | 0.27 | -0.148 | GIULIANO PASQUALE | 291 | 0.162 | 0.048 |
| FOLENA PIETRO | 242 | -0.765 | -0.071 | GIULIETTI GIUSEPPE | 292 | -0.716 | -0.087 |
| FOLLINI MARCO | 243 | 0.187 | -0.084 | GNAGA SIMONE | 293 | 0.782 | 0.552 |
| FONGARO CARLO | 244 | 0.803 | 0.596 | GRAMAZIO DOMENICO | 294 | 0.476 | -0.334 |
| FONTAN ROLANDO | 245 | 0.761 | 0.649 | GRIGNAFFINI GIOVANNA | 295 | -0.769 | -0.097 |
| FONTANINI PIETRO | 246 | 0.725 | 0.469 | GRILLO MASSIMO | 296 | 0.082 | -0.098 |
| FORMENTI FRANCESCO | 247 | 0.721 | 0.693 | GRIMALDI TULLIO | 297 | -0.869 | 0.495 |
| FOTI TOMMASO | 248 | 0.462 | -0.493 | GRUGNETTI ROBERTO | 298 | 0.789 | 0.614 |
| FRAGALA' VINCENZO | 249 | 0.333 | -0.27 | GUARINO ANDREA | 299 | -0.703 | -0.1 |
| FRANZ DANIELE | 250 | 0.442 | -0.546 | GUERRA MAURO | 300 | -0.766 | -0.098 |
| FRATTA PIERALFONSO | 251 | 0.139 | -0.185 | GUERZONI ROBERTO | 301 | -0.767 | -0.085 |
| FRATTINI FRANCO | 252 | 0.26 | -0.071 | GUIDI ANTONIO | 302 | 0.156 | -0.081 |
| FRAU AVENTINO | 253 | 0.249 | 0.121 | IACOBELLIS ERMANNO | 303 | -0.065 | -0.548 |
| FREDDA ANGELO | 254 | -0.722 | 0.11 | INNOCENTI RENZO | 304 | -0.779 | -0.083 |
| FRIGATO GABRIELE | 255 | -0.746 | -0.087 | OTTI LEONILDE | 305 | -0.814 | -0.019 |

Table 16: Individual MPs coordinates - Continued.

| Fullname | Code | X1 | X2 | Fullname | Code | X1 | X2 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| IZZO DOMENICO | 306 | -0.812 | 0.269 | MANZIONE ROBERTO | 357 | -0.143 | -0.524 |
| IZZO FRANCESCA | 307 | -0.724 | -0.181 | MANZONI VALENTINO | 358 | 0.445 | -0.564 |
| JANNELLI EUGENIO | 308 | -0.748 | -0.073 | MARENGO LUCIO | 359 | 0.354 | -0.531 |
| JERVOLINO ROSA | 309 | -0.78 | -0.056 | MARIANI PAOLA | 360 | -0.759 | -0.085 |
| LAMALFA GIORGIO | 310 | -0.758 | -0.616 | MARINACCI NICANDRO | 361 | 0.081 | -0.208 |
| LARUSSA IGNAZIO | 311 | 0.338 | -0.135 | MARINI FRANCO | 362 | -0.787 | -0.198 |
| LABATE GRAZIA | 312 | -0.74 | -0.154 | MARINO GIOVANNI | 363 | 0.321 | -0.448 |
| LADU SALVATORE | 313 | -0.741 | -0.093 | MARONGIU GIANNI | 364 | -0.722 | 0.101 |
| LAMACCHIA BONAVENTURA | 314 | -0.78 | -0.102 | MARONI ROBERTO | 365 | 0.814 | 0.581 |
| LANDI GIAMPAOLO | 315 | 0.49 | -0.376 | MAROTTA RAFFAELE | 366 | 0.175 | -0.154 |
| LANDOLFI MARIO | 316 | 0.398 | -0.45 | MARRAS GIOVANNI | 367 | 0.276 | -0.1 |
| LAVAGNINI ROBERTO | 317 | 0.203 | -0.139 | MARTINELLI PIERGIORGIO | 369 | 0.814 | 0.58 |
| LECCESE VITO | 318 | -0.943 | 0.145 | MARTINI LUIGI | 370 | 0.33 | -0.428 |
| LEMBO ALBERTO | 319 | 0.754 | 0.352 | MARTINO ANTONIO | 371 | 0.361 | -0.626 |
| LENTI MARIA | 320 | -0.488 | 0.873 | MARTUSCIELLO ANTONIO | 372 | 0.251 | -0.158 |
| LENTO GUGLIELMO | 321 | -0.761 | 0.077 | MARZANO ANTONIO | 373 | 0.357 | -0.302 |
| LEONE ANTONIO | 322 | 0.321 | -0.185 | MASELLI DOMENICO | 374 | -0.795 | -0.077 |
| LEONI CARLO | 323 | -0.753 | -0.1 | MASI DIEGO | 375 | -0.537 | 0.178 |
| LICALZI MARIANNA | 324 | -0.412 | -0.23 | MASIERO MARIO | 376 | 0.34 | -0.339 |
| LIOTTA SILVIO | 325 | -0.143 | 0.226 | MASSA LUIGI | 377 | -0.759 | -0.081 |
| LOJUCCO DOMENICO | 326 | 0.23 | -0.146 | MASSIDDA PIERGIORGIO | 378 | 0.257 | -0.097 |
| LOPORTO GUIDO | 327 | 0.352 | -0.294 | MASTELLA CLEMENTE | 379 | -0.372 | -0.482 |
| LOPRESTI ANTONINO | 328 | 0.316 | -0.55 | MASTROLUCA FRANCESCO | 380 | -0.759 | -0.085 |
| LODDO ANTONIO | 329 | -0.651 | -0.053 | MATACENA AMEDEO | 381 | 0.274 | -0.255 |
| LOMBARDI GIANCARLO | 330 | -0.752 | 0.056 | MATRANGA CRISTINA | 382 | 0.129 | -0.095 |
| LORENZETTI MARIARITA | 331 | -0.762 | -0.056 | MATTARELLA SERGIO | 383 | -0.78 | -0.025 |
| LORUSSO ANTONIO | 332 | 0.287 | -0.209 | MATTEOLI ALTERO | 384 | 0.268 | -0.425 |
| LOSURDO STEFANO | 333 | 0.278 | -0.45 | MATTIOLI FRANCESCO | 385 | -0.773 | -0.056 |
| LUCA' MIMMO | 334 | -0.764 | -0.092 | MAURO MASSIMO | 386 | -0.824 | -0.1 |
| LUCCHESE PAOLO | 335 | 0.163 | -0.044 | MAZZOCCHI ANTONIO | 387 | 0.381 | -0.62 |
| LUCIDI MARCELLA | 336 | -0.769 | -0.094 | MAZZOCCHIN GIANANTONIO | 388 | -0.747 | -0.084 |
| LUMIA GIUSEPPE | 337 | -0.769 | -0.066 | MELANDRI GIOVANNA | 389 | -0.775 | -0.032 |
| LUONGO ANTONIO | 338 | -0.807 | -0.422 | MELOGRANI PIERO | 390 | 0.248 | -0.123 |
| MACCANICO ANTONIO | 339 | -0.752 | -0.074 | MELONI GIOVANNI | 391 | -0.889 | 0.459 |
| MAGGI ROCCO | 340 | -0.777 | -0.06 | MENIA ROBERTO | 392 | 0.405 | -0.531 |
| MAIOLO TIZIANA | 341 | 0.334 | -0.039 | MERLO GIORGIO | 393 | -0.757 | -0.085 |
| MAIONE FRANCESCO | 342 | 0.328 | -0.578 | MERLONI FRANCESCO | 394 | -0.709 | -0.053 |
| MALAGNINO UGO | 343 | -0.756 | 0.086 | MESSA VITTORIO | 395 | 0.447 | -0.427 |
| MALAVENDA MARA | 344 | 0.744 | 0.668 | MICCICHE' GIANFRANCO | 396 | 0.23 | -0.108 |
| MALENTACCHI GIORGIO | 345 | -0.47 | 0.883 | MICHELANGELI MARIO | 397 | -0.866 | 0.422 |
| MALGIERI GENNARO | 346 | 0.346 | -0.67 | MICHELINI ALBERTO | 398 | 0.145 | 0.009 |
| MAMMOLA PAOLO | 347 | 0.19 | -0.19 | MICHIELON MAURO | 400 | 0.787 | 0.474 |
| MANCA PAOLO | 348 | -0.671 | 0.111 | MIGLIAVACCA MAURIZIO | 401 | -0.756 | -0.081 |
| MANCINA CLAUDIA | 349 | -0.778 | 0.003 | MIGLIORI RICCARDO | 402 | 0.297 | -0.54 |
| MANCUSO FILIPPO | 350 | 0.481 | 0.148 | MIRAGLIA NICOLA | 403 | -0.053 | -0.581 |
| MANGIACAVALLO ANTONINO | 351 | -0.81 | -0.004 | MISURACA FILIPPO | 404 | 0.281 | -0.242 |
| MANTOVANI RAMON | 353 | -0.467 | 0.884 | MITOLO PIETRO | 405 | 0.414 | -0.332 |
| MANTOVANO ALFREDO | 354 | 0.395 | -0.316 | MOLGORA DANIELE | 406 | 0.773 | 0.635 |
| MANZATO SERGIO | 355 | -0.759 | -0.083 | MOLINARI GIUSEPPE | 407 | -0.736 | -0.097 |
| MANZINI PAOLA | 356 | -0.743 | 0.023 | MONACO FRANCESCO | 408 | -0.747 | -0.155 |

Table 17: Individual MPs coordinates - Continued.

| Fullname | Code | X1 | X2 | Fullname | Code | X1 | X2 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| MONTECCHI ELENA | 409 | -0.804 | -0.125 | PECORAROSCANIO ALFONSO | 459 | -0.913 | -0.408 |
| MORGANDO GIANFRANCO | 410 | -0.759 | -0.007 | PECORELLA GAETANO | 460 | 0.266 | -0.178 |
| MORONI ROSANNA | 411 | -0.874 | 0.487 | PENNA RENZO | 461 | -0.758 | -0.078 |
| MORSELLI STEFANO | 412 | 0.371 | -0.597 | PENNACCHI LAURA | 462 | -0.795 | -0.141 |
| MUSSI FABIO | 413 | -0.788 | -0.091 | PEPE ANTONIO | 463 | 0.316 | -0.58 |
| MUSSOLINI ALESSANDRA | 414 | 0.334 | -0.598 | PEPE MARIO | 464 | -0.763 | 0.003 |
| MUZIO ANGELO | 415 | -0.846 | 0.521 | PERETTI ETTORE | 465 | 0.169 | -0.055 |
| NAN ENRICO | 416 | 0.289 | -0.289 | PERUZZA PAOLO | 466 | -0.768 | 0.092 |
| NANIA DOMENICO | 417 | 0.309 | -0.632 | PETRELLA GIUSEPPE | 467 | -0.758 | -0.082 |
| NAPOLI ANGELA | 418 | 0.426 | -0.402 | PETRINI PIERLUIGI | 468 | -0.877 | 0.13 |
| NAPPI GIANFRANCO | 419 | -0.767 | -0.097 | PEZZOLI MARIO | 469 | 0.392 | -0.017 |
| NARDINI MARIACELESTE | 420 | -0.481 | 0.877 | PEZZONI MARCO | 470 | -0.752 | -0.089 |
| NARDONE CARMINE | 421 | -0.778 | 0.041 | PICCOLO SALVATORE | 471 | -0.782 | -0.095 |
| NEGRI LUIGI | 422 | -0.394 | -0.222 | PILO GIOVANNI | 472 | 0.318 | -0.285 |
| NERI SEBASTIANO | 423 | 0.224 | -0.518 | PINZA ROBERTO | 473 | -0.754 | -0.144 |
| NESI NERIO | 424 | -0.903 | 0.43 | PIROVANO ETTORE | 474 | 0.697 | 0.717 |
| NICCOLINI GUALBERTO | 425 | 0.302 | -0.153 | PISANU BEPPE | 475 | 0.293 | -0.323 |
| NIEDDA GIUSEPPE | 426 | -0.765 | -0.081 | PISAPIA GIULIANO | 476 | -0.491 | 0.871 |
| NOCERA LUIGI | 427 | -0.189 | -0.507 | PISCITELLO RINO | 477 | -0.673 | 0.164 |
| NOVELLI DIEGO | 428 | -0.832 | 0.145 | PISTELLI LAPO | 478 | -0.751 | -0.09 |
| OCCHETTO ACHILLE | 429 | -0.721 | -0.076 | PISTONE GABRIELLA | 479 | -0.922 | 0.386 |
| OCCHIONERO LUIGI | 430 | -0.679 | -0.308 | PITTELLA GIOVANNI | 480 | -0.74 | -0.167 |
| OLIVERIO GERARDO | 431 | -0.766 | -0.042 | PITTINO DOMENICO | 481 | 0.762 | 0.609 |
| OLIVIERI LUIGI | 432 | -0.761 | -0.058 | PIVA ANTONIO | 482 | 0.257 | 0.04 |
| OLIVO ROSARIO | 433 | -0.741 | -0.157 | PIVETTI IRENE | 483 | -0.326 | -0.48 |
| ORLANDO FEDERICO | 434 | -0.736 | -0.057 | POLENTA PAOLO | 484 | -0.741 | -0.088 |
| ORTOLANO DARIO | 435 | -0.861 | 0.466 | POLIBORTONE ADRIANA | 485 | 0.47 | -0.572 |
| OSTILLIO MASSIMO | 436 | -0.117 | -0.559 | POLIZZI ROSARIO | 486 | 0.326 | -0.261 |
| OZZA EUGENIO | 437 | 0.334 | -0.412 | POMPILI MASSIMO | 487 | -0.76 | 0.007 |
| PACE CARLO | 438 | 0.414 | -0.523 | PORCU CARMELO | 488 | 0.411 | -0.532 |
| PACE GIOVANNI | 439 | 0.362 | -0.51 | POSSA GUIDO | 489 | 0.322 | -0.251 |
| PAGANO SANTINO | 440 | -0.209 | -0.811 | POZZA ELISA | 490 | -0.667 | -0.242 |
| PAGLIARINI GIANCARLO | 441 | 0.821 | 0.563 | PRESTAMBURGO MARIO | 491 | -0.741 | -0.174 |
| PAGLIUCA NICOLA | 442 | 0.208 | -0.201 | PRESTIGIACOMO STEFANIA | 492 | 0.242 | -0.075 |
| PAGLIUZZI GABRIELE | 443 | 0.452 | -0.415 | PREVITI CESARE | 493 | 0.208 | -0.437 |
| PAISSAN MAURO | 444 | -0.933 | 0.163 | PROCACCI ANNAMARIA | 494 | -0.987 | 0.16 |
| PALMA PAOLO | 445 | -0.807 | -0.098 | PRODI ROMANO | 495 | -0.743 | -0.129 |
| PALMIZIO MASSIMO | 446 | 0.227 | -0.174 | PROIETTI LIVIO | 496 | 0.394 | -0.665 |
| PALUMBO GIUSEPPE | 447 | 0.245 | -0.102 | Rabbito gaetano | 497 | -0.768 | -0.084 |
| PAMPO FEDELE | 448 | 0.38 | -0.555 | RADICE ROBERTO | 498 | 0.295 | -0.158 |
| PANATTONI GIORGIO | 449 | -0.779 | -0.018 | RaFFAELLI PAOLO | 499 | -0.782 | -0.019 |
| PANETTA GIOVANNI | 450 | 0.076 | -0.081 | RAFFALDINI FRANCO | 500 | -0.759 | -0.086 |
| PAOLONE BENITO | 451 | 0.304 | -0.398 | RALLO MICHELE | 501 | 0.413 | -0.401 |
| PARENTI TIZIANA | 452 | -0.18 | -0.355 | RANIERI UMBERTO | 502 | -0.78 | -0.09 |
| PARISI ARTURO | 453 | -0.869 | -0.496 | RASI GAETANO | 503 | 0.381 | -0.463 |
| PAROLI ADRIANO | 454 | 0.251 | -0.131 | RAVA LINO | 504 | -0.758 | -0.085 |
| PAROLO UGO | 455 | 0.806 | 0.536 | REBECCHI ALDO | 505 | -0.799 | -0.376 |
| PARRELLI ENNIO | 456 | -0.787 | -0.085 | REBUFFA GIORGIO | 506 | 0.021 | -0.149 |
| PASETTO GIORGIO | 457 | -0.763 | -0.062 | REPETTO ALESSANDRO | 507 | -0.752 | -0.074 |
| PASETTO NICOLA | 458 | 0.478 | -0.592 | RICCI MICHELE | 508 | -0.757 | 0.176 |

Table 18: Individual MPs coordinates - Continued.

| Fullname | Code | X1 | X2 | Fullname | Code | X1 | X2 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| RICCIO EUGENIO | 509 | 0.476 | -0.675 | SCHMID SANDRO | 560 | -0.774 | -0.088 |
| RICCIOTTI PAOLO | 510 | -0.557 | 0.199 | SCIACCA ROBERTO | 561 | -0.771 | -0.091 |
| RISARI GIANNI | 511 | -0.701 | -0.253 | SCOCA MARETTA | 562 | -0.166 | -0.434 |
| RIVA LAMBERTO | 512 | -0.736 | -0.243 | SCOZZARI GIUSEPPE | 563 | -0.708 | 0.064 |
| RIVELLI NICOLA | 513 | 0.286 | 0.071 | SCRIVANI OSVALDO | 564 | -0.768 | -0.08 |
| RIVERA GIOVANNI | 514 | -0.781 | -0.139 | SEDIOLI SAURO | 565 | -0.759 | -0.083 |
| RIVOLTA DARIO | 515 | 0.28 | -0.126 | SELVA GUSTAVO | 566 | 0.358 | -0.532 |
| RIZZA ANTONIETTA | 516 | -0.715 | -0.112 | SERAFINI ANNAMARIA | 567 | -0.75 | -0.066 |
| RIZZI CESARE | 517 | 0.74 | 0.562 | SERRA ACHILLE | 568 | 0.145 | -0.047 |
| RIZZO ANTONIO | 518 | 0.377 | -0.588 | SERVODIO GIUSEPPINA | 569 | -0.816 | -0.1 |
| RIZZO MARCO | 519 | -0.843 | 0.538 | SESTINI GRAZIA | 570 | 0.261 | 0.078 |
| RODEGHIERO FLAVIO | 520 | 0.806 | 0.437 | SETTIMI GINO | 571 | -0.757 | -0.193 |
| ROGNA SERGIO | 521 | -0.755 | -0.16 | SGARBI VITTORIO | 572 | 0.071 | 0.03 |
| ROMANI PAOLO | 522 | 0.262 | -0.109 | SICA VINCENZO | 573 | -0.79 | -0.086 |
| ROMANO DOMENICO | 523 | -0.729 | -0.014 | SIGNORINI STEFANO | 574 | 0.609 | 0.572 |
| ROSCIA DANIELE | 524 | 0.576 | 0.806 | SIGNORINO ELSA | 575 | -0.728 | -0.072 |
| ROSSETTO GIUSEPPE | 525 | 0.276 | -0.15 | SIMEONE ALBERTO | 576 | 0.33 | -0.395 |
| ROSSI EDO | 526 | -0.481 | 0.877 | SINISCALCHI VINCENZO | 577 | -0.729 | 0.254 |
| ROSSI GUIDO | 527 | 0.799 | 0.004 | SINISI GIANNICOLA | 578 | -0.706 | -0.157 |
| ROSSI ORESTE | 528 | 0.645 | 0.76 | SIOLA UBERTO | 579 | -0.751 | -0.201 |
| ROSSIELLO GIUSEPPE | 529 | -0.757 | -0.082 | SOAVE SERGIO | 580 | -0.752 | 0.101 |
| ROSSO ROBERTO | 530 | 0.349 | -0.028 | SODA ANTONIO | 581 | -0.745 | -0.044 |
| ROTUNDO ANTONIO | 531 | -0.745 | -0.001 | SOLAROLI BRUNO | 582 | -0.759 | -0.044 |
| RUBERTI ANTONIO | 532 | -0.757 | 0.017 | SORIERO GIUSEPPE | 583 | -0.736 | -0.195 |
| RUBINO ALESSANDRO | 533 | 0.313 | -0.146 | SORO ANTONELLO | 584 | -0.735 | -0.095 |
| RUBINO PAOLO | 534 | -0.781 | 0.12 | SOSPIRI NINO | 585 | 0.369 | -0.47 |
| RUFFINO ELVIO | 535 | -0.762 | -0.074 | SPINI VALDO | 586 | -0.743 | -0.07 |
| RUGGERI RUGGERO | 536 | -0.764 | -0.247 | STAGNO FRANCESCO | 587 | 0.284 | -0.2 |
| RUSSO PAOLO | 537 | 0.25 | -0.165 | STAJANO ERNESTO | 588 | -0.623 | 0.125 |
| RUZZANTE PIERO | 538 | -0.761 | -0.09 | STANISCI ROSA | 589 | -0.759 | -0.082 |
| SABATTINI SERGIO | 539 | -0.742 | -0.069 | STEFANI STEFANO | 590 | 0.756 | 0.427 |
| SAIA ANTONIO | 540 | -0.888 | 0.46 | STELLUTI CARLO | 591 | -0.76 | -0.079 |
| SALES ISAIA | 541 | -0.701 | -0.097 | STORACE FRANCESCO | 592 | 0.39 | -0.56 |
| SALVATI MICHELE | 542 | -0.699 | 0.051 | STRADELLA FRANCESCO | 593 | 0.34 | -0.09 |
| SANTANDREA DANIELA | 543 | 0.774 | 0.633 | STRAMBI ALFREDO | 594 | -0.915 | 0.404 |
| SANTORI ANGELO | 545 | 0.222 | -0.143 | STUCCHI GIACOMO | 595 | 0.823 | 0.569 |
| SANZA ANGELO | 546 | -0.115 | -0.356 | SUSINI MARCO | 596 | -0.76 | -0.073 |
| SAONARA GIOVANNI | 547 | -0.692 | 0.094 | TABORELLI ALBERTO | 597 | 0.265 | -0.155 |
| SAPONARA MICHELE | 548 | 0.209 | -0.203 | TARADASH MARCO | 598 | 0.393 | -0.185 |
| SARACA GIANFRANCO | 549 | -0.211 | -0.351 | TARDITI VITTORIO | 599 | 0.25 | -0.19 |
| SARACENI LUIGI | 550 | -0.858 | 0.132 | TARGETTI FERDINANDO | 600 | -0.751 | -0.03 |
| SAVARESE ENZO | 551 | 0.304 | -0.5 | TASSONE MARIO | 601 | 0.038 | -0.055 |
| SAVELLI GIULIO | 552 | -0.081 | -0.322 | TATARELLA GIUSEPPE | 602 | 0.342 | -0.383 |
| SBARBATI LUCIANA | 553 | -0.651 | 0.096 | TATARELLA SALVATORE | 603 | 0.414 | -0.486 |
| SCAJOLA CLAUDIO | 554 | 0.096 | -0.153 | TATTARINI FLAVIO | 604 | -0.734 | -0.042 |
| SCALIA MASSIMO | 555 | -0.919 | 0.281 | TERZI SILVESTRO | 605 | 0.822 | 0.57 |
| SCALTRITTI GIANLUIGI | 556 | 0.237 | -0.177 | TESTA LUCIO | 606 | -0.723 | -0.327 |
| SCANTAMBURLO DINO | 557 | -0.813 | 0.087 | TORTOLI ROBERTO | 607 | 0.247 | -0.15 |
| SCARPA PAOLO | 558 | 0.291 | -0.145 | TOSOLINI RENZO | 608 | 0.417 | -0.522 |
| SCHIETROMA GIANFRANCO | 559 | -0.793 | 0.185 | TRABATTONI SERGIO | 609 | -0.766 | 0.032 |

Figure 6: Individual MPs coordinates - Continued.

| FulIname | Code | X1 | X2 |
| :--- | ---: | ---: | ---: |
| TRANTINO ENZO | 610 | 0.345 | -0.478 |
| TREMAGLIA MIRKO | 611 | 0.428 | -0.345 |
| TREMONTI GIULIO | 612 | 0.394 | 0 |
| TREU TIZIANO | 613 | -0.786 | -0.176 |
| TRINGALI PAOLO | 614 | 0.415 | -0.62 |
| TUCCILLO DOMENICO | 615 | -0.768 | -0.112 |
| TURCI LANFRANCO | 616 | -0.723 | -0.196 |
| TURCO LIVIA | 617 | -0.83 | -0.107 |
| TURRONI SAURO | 618 | -0.788 | 0.434 |
| URBANI GIULIANO | 619 | 0.206 | -0.104 |
| URSO ADOLFO | 620 | 0.404 | -0.52 |
| VALDUCCI MARIO | 621 | 0.251 | -0.119 |
| VALENSISE RAFFAELE | 622 | 0.354 | -0.511 |
| VALETTO MARIAPIA | 623 | -0.778 | -0.06 |
| VALPIANA TIZIANA | 624 | -0.443 | 0.897 |
| VANNONI MAURO | 625 | -0.764 | -0.068 |
| VASCON LUIGINO | 626 | 0.784 | 0.582 |
| VELTRI ELIO | 627 | -0.699 | 0.157 |
| VELTRONI VALTER | 628 | -0.848 | -0.29 |
| VENDOLA NICHI | 629 | -0.505 | 0.863 |
| VENETO ARMANDO | 630 | -0.753 | -0.187 |
| VENETO GAETANO | 631 | -0.77 | -0.132 |
| VENTURA MICHELE | 632 | -0.758 | -0.335 |
| VIALE EUGENIO | 633 | 0.389 | -0.146 |
| VIGNALI ADRIANO | 634 | -0.756 | -0.074 |
| VIGNERI ADRIANA | 635 | -0.769 | -0.247 |
| VIGNI FABRIZIO | 636 | -0.762 | -0.067 |
| VILLETTI ROBERTO | 637 | -0.77 | 0.123 |
| VISCO VINCENZO | 638 | -0.771 | -0.216 |
| VITA VINCENZO | 639 | -0.729 | -0.094 |
| VITALI LUIGI | 640 | 0.239 | -0.112 |
| VITO ELIO | 641 | 0.202 | -0.165 |
| VOGLINO VITTORIO | 642 | -0.794 | -0.09 |
| VOLONTE' LUCA | 643 | -0.045 | -0.256 |
| VOLPINI DOMENICO | 644 | -0.754 | -0.101 |
| VOZZA SALVATORE | 645 | -0.774 | -0.075 |
| ZELDEMANN JOHANNGEORG | 646 | -0.785 | 0.171 |
| ZACCHEO VINCENZO | 647 | 0.364 | -0.612 |
| ZACCHERA MARCO | 648 | 0.297 | -0.362 |
|  | 649 | -0.742 | -0.093 |
| ZANATTI ALFREDO | 0.14 |  |  |
|  |  |  |  |

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[^1]:    ${ }^{1}$ The source is the House itself, through its web-site: http://www.camera.it
    ${ }^{2}$ A detailed explanation of the PR methodology is in Poole [9].

[^2]:    ${ }^{3}$ A more general version, (DW-)NOMINATE, allows for more sophisticated estimations of MPs bliss points, as it takes into account their voting patterns across legislatures. As data about the Italian Parliament have just started to unfold, at present we can only confine ourselves to the static version of this procedure.
    ${ }^{4} \mathrm{PCC}$ is just the percentage of the correct predictions that are generated by the estimated model. APRE instead is explaining how the model can improved with respect to the trivial prediction all MPs vote according to the majority. It is computed in the

[^3]:    ${ }^{5}$ In the appendix we report both TESEO and OURS classifications. Clausen and Peltzman can be found, for example, in Poole and Rosenthal [10].
    ${ }^{6}$ Specifically, if we let $p_{i}$ denote the percentage of bills within category $i=1, \cdots, n$, the Herfindahl index is computed as $H=\sum_{i} p_{i}^{2}$. Whenever the bills are evenly split, $H=1 / n$. Whenever bills are concentrated on one category only, $H=1$. Our normalization generates $N H=(n H-1) /(n-1)$, which ranges between 0 and 1.

[^4]:    ${ }^{7}$ Specifically, we followed this two steps process: first, we computed the average (median) of MP's coordinates by group in each bill, considering only those MPs that actually voted on that bill. Our score is the group average across all bills.
    ${ }^{8}$ At the extreme right-end of the spectrum, we have a noticeable exception, the Northern League (LN). In the following paragraphs we will comment about this party's position.

[^5]:    ${ }^{9}$ Specifically, let $s_{i j}, n_{i j}, a_{i j}$ and $h_{i j}$ be, respectively, the number of MPs for group $j$ that vote yes, no, abstain or stay home for roll call $i=1, \ldots m$. The Rice index for group $j$ is

    $$
    R_{j}=\frac{1}{m} \sum_{i=1}^{m} \frac{\left|s_{i j}-n_{i j}\right|}{s_{i j}+n_{i j}} ;
    $$

