WEIGHTED VORONOI REGION ALGORITHMS
FOR POLITICAL DISTRICTING

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Abstract

Automated political districting shares with electronic voting the aim of preventing electoral manipulation and pursuing an impartial electoral mechanism. Political districting can be modelled as multiobjective partitioning of a graph into connected components, where population equality and compactness must hold if a majority voting rule is adopted. This leads to the formulation of combinatorial optimization problems that are extremely hard to solve exactly. We propose a class of heuristics, based on discrete weighted Voronoi regions, for obtaining compact and balanced districts, and discuss some formal properties of these algorithms. Their performance has been tested on randomly generated rectangular grids, as well as on real-life benchmarks; for the latter instances the resulting district maps are compared with the institutional ones adopted in the Italian political elections from 1994 to 2001.

Keywords: political districting, weighted Voronoi regions, graph partitioning, heuristics.

1 Introduction

Soon after modern democracies were established, gerrymandering practices, consisting of partisan manipulation of electoral district boundaries, began to occur in several states and countries. In order to oppose such practices, when the electronic computer became available researchers started thinking of automatic procedures for political districting, designed so as to be as neutral as possible. Commonly adopted criteria are:

Integrity - The territory to be subdivided into districts consists of territorial units (wards, townships, counties, etc.) and each unit cannot be split between two or more districts.

Contiguity - The units of each district should be geographically contiguous, that is, one can walk from any point in the district to any other point of it without ever leaving the district.

Population equality (or *population balance*) - Under the assumption that the electoral system is majoritarian with single-member districts, all districts should have roughly the same population (*one man – one vote* Principle).

Compactness - Each district should be compact, that is, "closely and neatly packed together" (Oxford Dictionary). Thus, a round-shaped district is deemed to be acceptable, while an octopus- or an eel-like one is not.

   A broad survey of political districting algorithms is given in (Grilli di Cortona et al., 1999). Later work focuses on local search (e.g., Ricca and Simeone, 2007; Bozkaya, Erkut, and Laporte, 2003). It is also worth mentioning the branch-and-price approach in (Mehrotra, Johnson, and Nemhauser, 1998). Here we propose a novel approach based on *weighted Voronoi regions* (or diagrams; WVR for short). This notion is not new in the literature, especially in the area of computational geometry (see, e.g., Aurenhammer and Edelsbrunner, 1984). Also the *discrete* version of the Voronoi regions is not new: for example, it was widely applied in network location problems [see, for example, Drezner and Hamacher, 2002]. What we believe to be new is our iterative updating of node weights. As we will show later, in the specific application to political districting, the computation of the WVR with respect to these updated weights is a useful tool to achieve both compactness and population balance of the districts. WVR procedures guarantee a flexible approach to the districting problem, since solutions with different trade-offs between compactness and population balance can be easily found: when population balance improves, compactness inevitably worsens; but, by stopping the procedure at different time points, one can always control the current trade-off between these two basic districting criteria. Figure 1 gives an idea of the shape of the districts obtained by a WVR procedure for a 20×20 grid. The grid is represented as a chessboard whose squares correspond to the grid nodes. In this case L_1 – distances in the grid were adopted.
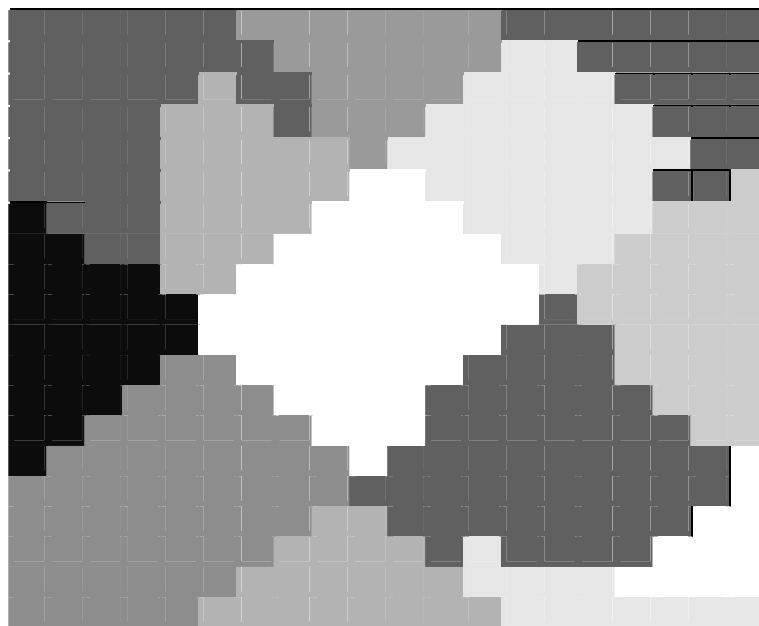

Figure 1 – A district map obtained by a WVR procedure for a 20×20 grid graph.

In this paper both theoretical and experimental results are provided. In particular, in Section 2 we describe a general paradigm for our WVR algorithm and the specific features that characterize each different variant of it. In Section 3, after indicating some pathologies that may occur if some caution is not taken, we define some desirable properties to be met by WVR algorithms and we provide conditions under which key-properties, such as "geodesic consistency", hold. We also give finite termination results relying on the theory of Majorization introduced in 1934 by Hardy, Littlewood and Pólya. In Section 4 we present some preliminary computational experiments performed with a specific implementation of the algorithm on a sample of four Italian Regions and on a benchmark of rectangular grid graphs.

2 Weighted Voronoi Region Procedures

The input to the weighted Voronoi procedures to be described in this section is the following: a *contiguity graph* G = (V,E), whose nodes represent the territorial units and where there is an edge between two nodes if the two corresponding units are neighbouring; a positive integer *r*, the number of districts; a subset S ⊂ V of *r* nodes, called *centers* (all the remaining nodes will be called *sites*); positive integral node weights p_i, $i \in V$, representing territorial unit *population*s; positive real *lengths* l_{ij} for all edges (*i, j*). Usually, edge-lenghts represent road distances, so as to take into account orography and other geographical barriers. Given the edge-lengths, we can accordingly compute the *distance d_{ij}* between any two nodes *i* and *j* as the length of any shortest path on G with endpoints *i* and *j*. A path on G is called a *geodesic* if it is a shortest path between its two endpoints. By slightly perturbing edge-lengths, if necessary, we may, and shall, assume without loss of generality that:

(i)   the distance function is injective, i.e., $d_{ij} \neq d_{i'j'}$ whenever $(i, j) \neq (i', j')$;

(ii)   between any two nodes there is a unique geodesic.

In fact, let $0 < \varepsilon < \frac{1}{2} \min \{1, L\}$, where $L = \min \{|d_{ij} - d_{i'j'}| : d_{ij} \neq d_{i'j'}, \forall i, j, i', j'\}$. After numbering the *m* edges of G from 1 to *m*, let us assign to the *k*-th edge (*i,j*) the perturbed length $l_{ij}(\varepsilon) = l_{ij} + \varepsilon^{k+1}$, (*k* = 1, .., *m*). For every geodesic Q, let us denote by d_Q and $d_Q(\varepsilon)$ the total length of Q relative to the original and to the perturbed edge-lengths, respectively, and let

$\Delta_Q(\varepsilon) = d_Q(\varepsilon) - d_Q = \sum_{h \in Q} \varepsilon^{h+1}$. Finally, let $\chi(Q)$ be the characteristic vector of Q in E. Remark

that, since $0 < \varepsilon < 1/2$, $\Delta_Q(\varepsilon) = \varepsilon \sum_{h \in Q} \varepsilon^h < \varepsilon \left(\dfrac{1}{2} + \dfrac{1}{2^2} + \dfrac{1}{2^3} + ... + \dfrac{1}{2^{|Q|}} \right) < \varepsilon.$

Furthermore, $\Delta_Q(\varepsilon) < \Delta_{Q'}(\varepsilon)$ iff $\chi(Q)$ is lex-smaller than $\chi(Q')$. Hence, under the above

assumptions on ε, $d_Q = d_{Q'} \Rightarrow d_Q(\varepsilon) \neq d_{Q'}(\varepsilon)$, and $d_Q < d_{Q'} \Rightarrow d_Q(\varepsilon) < d_{Q'}(\varepsilon)$.

It follows that the corresponding perturbed distance function $d(\varepsilon)$ is injective, and

$d_{ij} < d_{i'j'} \Rightarrow d_{ij}(\varepsilon) < d_{i'j'}(\varepsilon)$. Moreover, with the above-defined perturbed edge-lengths there is a

unique geodesic between any two nodes.

Under the injectivity assumption on d, for each site i there is always a unique center that is closest

to i. We denote by $\overline{P}$ the mean district population (= total population/r). In the remainder of the

paper, for brevity, we will denote by s both a center and a district centered in s (when this does not

cause any confusion).

The integrity criterion dictates that a district must be a subset of nodes; according to the contiguity

criterion, such subset must be connected. A *district map* is a partition of V into r connected subsets

(the districts), each containing exactly one center. Given any district map, we denote by D_s the

unique district containing center s. We look for a district map such that, informally speaking, the

district population imbalance is small and the districts are compact enough. According to our

approach, we first locate the subset $S \subset V$ of r centers and then we define the district map by

drawing on G the Voronoi diagram w.r.t. the distances d_{is}, $\forall i \in V - S, \forall s \in S$ (initial *discrete*

Voronoi regions), which can be seen as the graph-theoretic counterpart of the ordinary Voronoi

diagram in continuous space. More precisely, the *Voronoi region* (or *diagram*) of center s is the set

of all nodes i such that the closest center to i is s. Under the injectivity assumption, such closest

center is necessarily unique; moreover, since all edge-lengths are positive, for any center $s \in S$, we

have $d_{ss'} > 0$, $\forall s' \neq s$, thus s always belongs to D_s . Hence the Voronoi regions form a partition of

G. Furthermore, it can be shown (see Sec. 3) that Voronoi regions are connected.

The idea is that by taking as districts the Voronoi regions on G, a good compactness can be

achieved. Notice that the compactness of the district map strongly depends on the location of the r

centers. In order to get a good performance of the Voronoi approach w.r.t. compactness, we locate

the r centers by (heuristically) solving an unweighted r-center location problem on G (cf. Drezner

and Hamacher, 2002). In this way a good compactness level is usually achieved, but a poor

population balance might ensue. In order to re-balance district populations, one would like to promote site migration out of "heavier" districts (population-wise) and into lighter ones. Then, the basic idea is to consider weighted distances $d_{is}' = w_s d_{is}$, where each weight w_s is proportional to P_s, the population of district D_s, and to perform a Voronoi iteration (that is, the computation of the discrete Voronoi regions) w.r.t. the biased distances d_{is}'. Do this iteratively: at iteration k, $k = 1,2,...$, two different recursions may be taken into consideration, namely, a *static* one,

$$d_{is}^k = \frac{P_s^{k-1}}{\bar{P}} d_{is}^0 , \; i \in V - S, s \in S \tag{1}$$

and a *dynamic* one,

$$d_{is}^k = \frac{P_s^{k-1}}{\bar{P}} d_{is}^{k-1} , \; i \in V - S, s \in S , \tag{2}$$

where, in both cases, $d_{is}^0 = d_{is}$, P_s^0 is the population of the initial Voronoi region containing center s, and P_s^k is the population of D_s after iteration $k = 1,2,...$. Stop as soon as the districts become "stable", that is, the district map at some iteration coincides with the district map at the previous iteration. The above sketched algorithm will be called a *full transfer* one because, at each step, all the sites for which the closest center changes are transferred from their old district to the new one. Denote the set of these sites by $M \subseteq V - S$. In view of the possible finite termination difficulties of the full transfer algorithm (see Sec. 3), we also consider different versions for the WVR algorithm in which only a subset $Z \subset M$ of sites is actually allowed to migrate. Figure 2 shows the general paradigm of a WVR algorithm.

One may also consider a *single transfer* version of the WVR algorithm, by letting sites migrate one at a time from one district to another (in Figure 2 it corresponds to the case $|Z| = 1$); or a *partial transfer* version ($Z \subset M$), in which only a particular subset of sites (suitably selected according to some rule) migrates at each iteration. Here too, one may adopt either the static or the dynamic recursion defined above. So, one altogether gets six variants of the weighted Voronoi algorithm (static/dynamic recursion; full/partial/single transfer).

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WEIGHTED VORONOI REGION ALGORITHM

INPUT: G=(V,E), r, $p_i$  $\forall i \in V$, $d_{ij}$ $\forall i, j \in V$

OUTPUT: a connected partition of G

1.      Locate the set S of r centers in G
2.      $k = 0$
3.      Let $d_{is}^o = d_{is}$, $\forall i \in V - S$, $\forall s \in S$
4.      Compute the discrete Voronoi regions w.r.t. $d_{is}$ (initial district map)
5.      repeat
           $k = k + 1$
           update the distances $d_{is}^k$, $\forall i \in V - S$, according to $P_s^{k-1}$, $\forall s \in S$
           compute the subset M of sites that are candidates for migrating
           select a subset $Z \subseteq M$ and perform the corresponding migrations
           compute the discrete Voronoi regions w.r.t. $d_{is}^k$ (current district map)
        until M is empty
6.      output the last district map.
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Figure 2 – General paradigm of a WVR algorithm.

In particular, the implementation of the single transfer algorithm is the following:

At iteration k, some district D_t with minimum population, $P_t^{k-1} = \min\{P_s^{k-1} : s = 1, ..., r\}$, is selected as the destination district. Then, a set C of candidate sites for migrating into D_t is selected according to the following rule: site $i \notin D_t$ is a candidate for migrating into D_t if $d_{it}^k = \min\{d_{is}^k : s = 1, ..., r\}$. Finally, site i is chosen for migrating from D_q (the district it belongs to) to D_t if the following two conditions hold: (i) $d_{it}^k = \min\{d_{jt}^k : j \in C, \ j \notin D_t\}$; (ii) $P_t^k < P_q^k$. Notice that if site i belongs to district D_q and it is a candidate for migrating to D_t, then $d_{it}^k < d_{iq}^k$. The algorithm stops when the set of candidates is empty.

One possible implementation of the partial transfer algorithm is called *path transfer* and is defined as follows. Voronoi regions are calculated at the beginning. At each iteration an auxiliary network $\mathcal{N}$ is constructed such that nodes in $\mathcal{N}$ correspond to the current districts, while there is an arc between two nodes D and D' of $\mathcal{N}$ if and only if there exists a site j that can be moved from D to D'. At each iteration a suitable path P is selected in $\mathcal{N}$ and for each arc (D, D') of P a site migrates from D to D'.

3 Pathologies and Desirable Properties for WVR Algorithms

The WVR algorithms described above may encounter some pathological situations, such as those discussed below. In Figure 3, one case of lack of termination is presented in which the full transfer algorithm may loop. The numbers next to the nodes represent node-weights, those next to the edges edge-weights. At the beginning, both sites 1 and 2 are assigned to center s, thus generating a map with population equal to 199 and to 1 for districts s and t, respectively. In this extremely unbalanced situation, the iterative distance updating results in the repeated transfer of both sites 1 and 2 from district s to district t and back without termination.
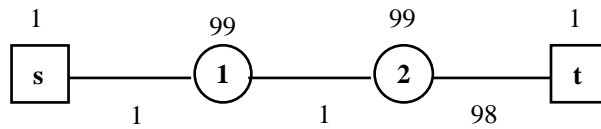

Figure 3 – Lack of termination for the dynamic full transfer WVR algorithm.

Figure 4 shows an example where lack of contiguity might arise when the site-to-center distances are completely arbitrary. Suppose that all the nodes have the same population. It is easy to check that the Voronoi regions are {1,3} and {2,4}. This district map is perfectly balanced, but not contiguous. Here, however, the distance function is neither metric nor injective.

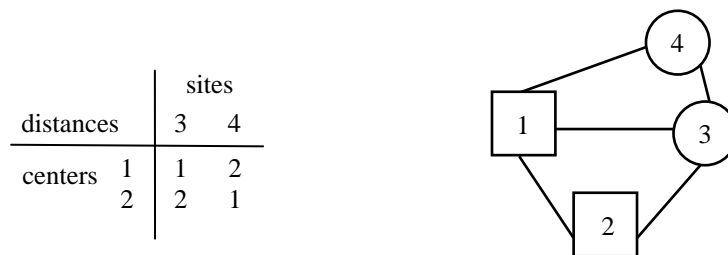| distances		sites	
		3	4
centers	1	1	2
	2	2	1

Figure 4 – An example of lack of contiguity, where all the nodes have the same population and the site-to-center distances are given in the table.

In order to prevent these and other pathologies from occurring, we introduce four desirable properties to be met by weighted Voronoi algorithms – or at least by some variants of them.

1) *order invariance*: at each step of the algorithm, the order relation on the sites w.r.t. their distances to any given center s does not change. Formally, at iteration $k = 1, 2, ...,$ we have

$$d_{is}^k < d_{js}^k \quad \Leftrightarrow \quad d_{is} < d_{js}, \quad s \in S; \ i, j \in V - S. \tag{3}$$

2) *re-balancing*:

at iteration $k = 1, 2,...,$ site i migrates from D_q to D_t only if $P_q^{k-1} > P_t^{k-1}$. \hfill (4)

3) *geodesic consistency*: at any iteration, if site j belongs to district D_s and site i lies on the geodesic between j and s, then i also belongs to D_s.

4) *finite termination*: the algorithm stops after a finite number of iterations.

Proposition 1. Order invariance holds for the full transfer WVR algorithms.

Proof. In the static case, the statement directly follows from the updating formula (1); in the dynamic case, it inductively follows from (2). $\qquad\qquad\square$

Proposition 2. Re-balance holds for the dynamic full transfer WVR algorithm.

Proof. If at iteration k site i is assigned to center s^*, then one must necessarily have

$$d_{is^*}^k = \min_{s=1,2,...,r} d_{is}^k,$$

that is, $d_{is^*}^k \leq d_{is}^k$, $\forall s \neq s^*$, and in particular, for a given center $\bar{s} \neq s^*$,

$$d_{is^*}^k \leq d_{i\bar{s}}^k. \tag{5}$$

Similarly, if at iteration $k+1$ site i belongs to the district with center $\bar{s}$, it must be

$$d_{i\bar{s}}^{k+1} = \min_{s=1,2,...,r} d_{is}^{k+1},$$

and thus

$$d_{i\bar{s}}^{k+1} \leq d_{is^*}^{k+1}. \tag{6}$$

If the dynamic updating formula (2) is adopted, one gets from (6)

$$d_{i\bar{s}}^k P_{\bar{s}}^k = d_{i\bar{s}}^{k+1} \leq d_{is^*}^{k+1} = d_{is^*}^k P_{s^*}^k$$

so, since (5) holds, one must have

$$P_{\bar{s}}^k \leq P_{s^*}^k,$$

implying that at iteration k the population of district $\bar{s}$ into which node i migrates was no larger than that of district s^* where i migrates from. $\qquad\qquad\square$

Remark 1. Notice that re-balance is not guaranteed for the static full transfer WVR algorithm. In this case, however, reasoning as in the above proof one gets the following slightly different property: at iteration k+1, site i migrates from district s* to district $\bar{s}$ only if

$$\frac{P_{\bar{s}}^{k}}{P_{\bar{s}}^{k-1}} \leq \frac{P_{s*}^{k}}{P_{s*}^{k-1}} \; .$$

Proposition 3. Geodesic consistency holds for the full transfer WVR algorithms.

Proof. Let us firstly show that the property holds w.r.t. the initial distances. So, suppose that node *j* is assigned to center *s* and that *i* lies on the geodesic from *j* to *s*. If *i* were assigned to center $s' \neq s$, then one would have $d_{is'} < d_{is}$ by the injectivity of *d*. But then one would have also $d_{js'} < d_{js}$, a contradiction. Then by order invariance the property must hold at each iteration *k*. □

Proposition 4. Geodesic consistency implies contiguity.

Proof. Suppose that vertex *i* belongs to district *s*. By geodesic consistency, every vertex of the geodesic between *i* and *s* must also be assigned to *s*. Thus for every *i* in district *s* there is a path from *i* to *s* entirely contained in district *s*, implying that the district is connected. □

Although Propositions 1-4 provide good properties for the full transfer WVR algorithm (at least for the dynamic version) the example in Figure 4 shows that finite termination does not hold in general for this class of algorithms. Hence, in our experiments, we have taken into consideration only the single and the partial transfer versions of WVR algorithms. As we shall see, the implementations of these algorithms satisfy order invariance, geodesic consistency, re-balancing and finite termination properties. The last two results rely on the Theory of Majorization introduced in 1934 by Hardy, Littlewood and Pólya (cf. Marshall and Olkin, 1979). They introduce the following definition of *transfer*:

Given a positive real vector $a = (a_1, a_2, ..., a_n)$, and given a pair *i* and *j* such that $a_i < a_j$, a *transfer* is an operation which transforms vector *a* into a new vector $\alpha = (\alpha_1, \alpha_2, ..., \alpha_n)$ as follows:

$$\alpha_s = a_s + \delta, \quad \text{if } s = i$$
$$a_s - \delta, \quad \text{if } s = j$$
$$a_s, \qquad \text{if } s \neq i, j$$

where $0 < \delta < (a_j - a_i)/2$. A transfer operation involving the pair (a_i, a_j), with $a_i < a_j$, returns the pair (α_i, α_j) such that $\alpha_i < \alpha_j$, that is, the order relation between elements i and j does not change.

One says that a nonnegative vector a is *strictly majorized* by another nonnegative vector b (notation: $a \prec b$) if a can be obtained from b through a finite number of transfers. The relation $\prec$ is a strict preorder (i.e., an irreflexive, asymmetric and transitive relation) in R_+^n.

The integer counterpart of a transfer in the sense of Hardy, Littlewood and Pólya (HLP transfer for short) can be defined when $a = (a_1, a_2, ..., a_n)$ and δ are all integers. Since populations are integers anyhow, we shall always perform integral transfers in our WVR algorithms.

Let $P^k = (P_1^k, P_2^k, ..., P_r^k)$ be the integer vector of the populations of the r districts at a generic iteration k. Then the following results hold.

Proposition 5. At iteration k, the single transfer algorithm performs a HLP transfer on P^{k-1}.

Proof. The result trivially follows from the definition of single transfer. $\square$

Consider the path transfer algorithm and denote by $q_1, ..., q_h$ the sequence of h consecutively adjacent districts that correspond to the nodes of a path in the auxiliary network $\mathcal{N}$. For every pair of consecutive districts, let $\delta_{q_i q_{i+1}}$ be the total amount of population that migrates from q_i to q_{i+1}, $i = 1, ..., h-1$, (that is, the population of the site that migrates from q_i to q_{i+1}).

Proposition 6. At iteration k, the path transfer algorithm performs a sequence of HLP transfers on P^{k-1} if $0 < \delta_{q_i q_{i+1}} < \delta = \min_{i=1,...,h-1} (P_{q_i}^{k-1} - P_{q_{i+1}}^{k-1})/2$, for all $i = 1, ..., h-1$.

Proof. According to the definition of path transfer, we must have $P_{q_1}^{k-1} > P_{q_2}^{k-1} > ... > P_{q_h}^{k-1}$. If $0 < \delta_{q_i q_{i+1}} < \delta = \min_{i=1,...,h-1} (P_{q_i}^{k-1} - P_{q_{i+1}}^{k-1})/2$ holds for $i = 1, ..., h-1$, then for every pair of consecutive districts in the path we have $P_{q_i}^k > P_{q_{i+1}}^k$, thus implying the result. $\square$

Notice that, if in a path transfer we have $\delta_{q_i q_{i+1}} = \delta > 0$ for all $i = 1, ..., h-1$, then the transfer modifies only the populations of the first and the last district in the path, respectively.

After Propositions 5 and 6 re-balancing is guaranteed by the very constructions for both the single and the path transfer algorithm, since HLP transfers always move a site from a heavier district to a lighter one. Moreover, order invariance and geodesic consistency hold by similar arguments as in Propositions 1 and 3, respectively. Finally, the following result guarantees finite termination for these algorithms.

Proposition 7. The single and path transfer weighted Voronoi algorithms halt after a finite number of steps.

Proof. On the basis of the previous results, during the execution of these algorithms the vector of district populations, $P^k = (P_1^k, P_2^k, ..., P_r^k)$, decreases w.r.t. the strict preorder $\prec$, so it cannot be encountered twice. Since the total number of partitions of V is finite, finite termination is guaranteed both for single transfer and path transfer weighted Voronoi algorithm. □

The following table summarizes the main theoretical results presented in this section. Here we denote the still open questions by "?".

Table 1 – Properties of the WVR algorithms.

Property	Static			Dynamic		
	Single transfer	**Path transfer**	**Full transfer**	**Single transfer**	**Path transfer**	**Full transfer**
Order invariance	yes	yes	yes	yes	yes	yes
Re-balancing	yes	yes	?	yes	yes	yes
Geodesic consistency	yes	yes	yes	yes	yes	yes
Finite termination	yes	yes	?	yes	yes	no

4 Preliminary Experimental Results

In this section we present some preliminary results we obtained with one of our WVR algorithms, namely, the single transfer WVR.

Following the pseudocode presented in Figure 2, the implementation of a WVR algorithm requires, first of all, the definition of the procedure for locating r centers in G.

In our implementation we compared two different methods to locate the centers: on the one hand, the location was performed by solving an unweighted r-center problem on G; on the other hand, the centers were located as far apart from each other as possible (we refer to this particular location as "sparse centers"). Both approaches are heuristics, but they produce very different results (see, e.g., the example reported in Figure 5), showing that the location of the centers at the beginning is a crucial step for a WVR algorithm. According to our experimental results, in general, the r-center

approach performs better than the other. This behaviour is confirmed also by the numerical results shown in Tables 2-5 which follow.
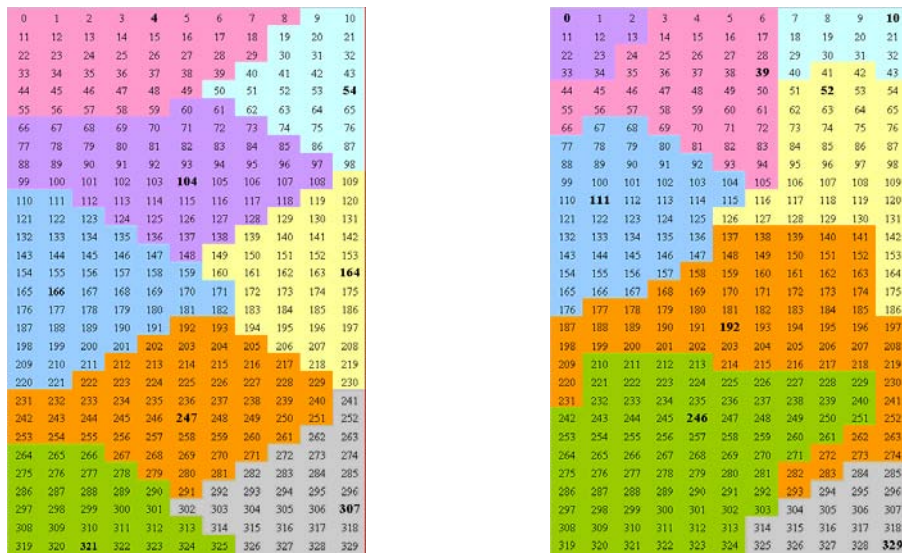

Figure 5 – Different district maps obtained on a rectangular 30×11 grid graph
according to different procedures for the location of the *r* centers.

Tables 2-5 show the results obtained with the single transfer algorithm on four contiguity graphs corresponding to a sample of Italian regions, both when the *r*-center and the sparse centers approaches are adopted. Population equality and compactness are measured by suitable indices varying between 0 and 1. For these indices (for more details, see Ricca and Simeone, 2007), a value close to 0 corresponds to a very balanced and very compact district map, respectively, thus meaning that the result is very good. On the other hand, values close to 1 suggest that the performance of the algorithm was poor. In fact, both indices can be read as percentages of *lack of population equality* and *lack of compactness*, respectively.

The "Initial value" refers to the partition obtained at the beginning by computing the discrete Voronoi regions in G. It is easy to recognize that very good values of the index of compactness are associated to these initial solutions, due to the fact that Voronoi diagrams generally produce "compact-shaped" regions. Unfortunately, the poor performance w.r.t. population equality – and the importance of such criterion in political districting – prevente us to stop the algorithm at this very preliminary step, and forces towards the search of better compromises through the application of a WVR procedure.

Table 2 – Latium 374 nodes, 2012 arcs, 19 districts.

Sparse centers	Initial value	Final value	Variation (%)	Number of iterations
Population Equality	0.540	0.297	- 45%	486
Compactness	0.295	0.419	+ 42%	
r-center	**Initial value**	**Final value**	**Variation (%)**	**Number of iterations**
Population Equality	0.364	0.199	- 45 %	769
Compactness	0.081	0.263	+ 2.25 %	

Table 3 – Piedmont 1208 nodes, 7055 arcs, 28 districts.

Sparse centers	Initial value	Final value	Variation (%)	Number of iterations
Population Equality	0.444	0.264	- 40 %	3201
Compactness	0.506	0.612	+ 21%	
r-center	**Initial value**	**Final value**	**Variation (%)**	**Number of iterations**
Population Equality	0.363	0.187	- 48 %	3087
Compactness	0.150	0.421	+ 1,81 %	

Table 4 – Abruzzi 305 nodes, 1694 arcs, 11 districts.

Sparse centers	Initial value	Final value	Variation (%)	Number of iterations
Population Equality	0.280	0.130	54 %	200
Compactness	0.338	0.451	+ 33 %	
r-center	**Initial value**	**Final value**	**Variation (%)**	**Number of iterations**
Population Equality	0.361	0.216	- 40 %	239
Compactness	0.134	0.305	+ 1.28 %	

Table 5 – Trentino 339 nodes, 1876 arcs. 8 districts.

Sparse centers	Initial value	Final value	Variation (%)	Number of iterations
Population Equality	0.357	0.165	- 54 %	239
Compactness	0.368	0.532	+ 45 %	
r-center	**Initial value**	**Final value**	**Variation (%)**	**Number of iterations**
Population Equality	0.412	0.306	- 26 %	314
Compactness	0.202	0.327	+ 62 %	

The final values found by our algorithm show that in the best case (*r*-center approach) we are able to fairly improve population equality without worsening compactness too much. However, if, on one side, the values of compactness are very good, it must be noticed that the lack of population equality here remains still too high to be compared with the corresponding values typically associated to a political district map. Table 6 reports the values for the population equality and compactness indices associated to the institutional district map of the four Italian regions that was actually adopted in Italy for the political elections of the Chamber of Deputies until 2001[1]. While it is immediately recognized that our compactness values are definitively better than the institutional ones, we cannot say the same for population equality. It must be also pointed out that, according to the electoral law[2], population equality is the main criterion in the design of the districts, while compactness is not considered at all. Nevertheless, better results can be attained for population equality, even in combination with good values for compactness. Actually, in a previous

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[1] This was the last election of the Chamber of Deputies performed in Italy with the old mixed system (Law 277/1993) for which a map of single-member districts was available. In 2005 the Italian electoral law was reformed and a proportional electoral system was adopted (Law 270/2005).
[2] Law 277/1993.

experimental work (Ricca and Simeone, 2007) we obtained such results on the same territories by applying local search techniques. On these grounds, we believe that local search could be successfully combined with WVR to obtain an algorithm that is able to reach both a good compactness together with good balance in population. We shall leave this project to future work.

Table 6 – Italian institutional district map.

Region	Population Equality	Compactness
Latium	0.06	0.68
Piedmont	0.1	0.88
Abruzzi	0.08	0.63
Trentino	0.04	0.70

Additional results are provided in Table 7, showing the outcome of the application of the single transfer WVR algorithm to rectangular grids. A rectangular grid shares with the contiguity graph of a real territory the properties that it is planar and has low vertex degree.

Here we show two examples related to medium-size grids with a different number of districts. The results confirm the good performance of the single transfer WVR algorithm, combined with the *r*-center approach, since the values obtained for the population equality and compactness indices are all very low (between 12% and 15%).

Table 7 – Rectangular grid graphs (*r*-center approach).

Grid 20x20 (15 districts)	Initial value	Final value	Variation (%)
Population Equality	0.206	0.148	- 28 %
Compactness	0.098	0.157	+ 59 %
Grid 30x11 (8 districts)	**Initial value**	**Final value**	**Variation (%)**
Population Equality	0.185	0.157	- 15 %
Compactness	0.107	0.120	+ 12 %

To conclude, on the basis of our results, the class of WVR algorithms appears to be a useful tool for the design of impartial electoral district maps. Even if the (very preliminary) experimental results are still not fully satisfactory, we believe that better results can be achieved through the implementation of more sophisticated variants of WVR, such as the path transfer, also in view of the possibility of combining it with local search.

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