Sums of permanental minors using Grassmann algebra

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Abstract

We show that a formalism proposed by Creutz to evaluate Grassmann integrals provides an algorithm of complexity $O(2^n n^3)$ to compute the generating function for the sum of the permanental minors of a matrix of order $n$. This algorithm improves over the Brualdi-Ryser formula, whose complexity is at least $O(2^{5n})$. In the case of a banded matrix with band width $w$ and rank $n$ the complexity is $O(2^{\min(2w,n)}(w + 1)n^2)$.

Related algorithms for the matching and independence polynomials of graphs are presented.

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I. INTRODUCTION

Let $G$ be an undirected graph with $E$ edges and $v = |V|$ vertices. A matching of $G$ is a set of pairwise disjoint edges. The matching generating polynomial of $G$, i.e. the generating function of the number $N(i)$ of different matchings of $G$ containing $i$ edges, defined by

$$M(t) = \sum_{i=0}^{[v/2]} N(i)t^i$$

first appeared in combinatorics as the “rook” polynomial\(^2\). It was then introduced in statistical physics\(^3\) for the study of the monomer-dimer system on a lattice, in theoretical chemistry\(^4\) to compute the “Hosoya index” $Z(G) = M(1)$, i.e. the total number of matchings of $G$.

$M(-1)$ on a lattice graph can be interpreted as the Witten index of a supersymmetric dimer model defined on that lattice\(^5\).

For a generic graph, the best current algorithms for computing the matching generating polynomial are based\(^6\) on recurrence relations. Alternatively, for a bipartite graph the coefficients of the matching polynomial can be computed as the sum of the permanental minors of the reduced adjacency matrix of the graph (defined as the submatrix of the adjacency matrix from the even to the odd vertices). The best current algorithm for computing the sum of the permanental minors is the Brualdi-Ryser formula\(^7\). Notice that even in the case of the permanent of banded matrices the complexity of currently used algorithm is in general exponential. In Ref.\(^8\) it was shown that for banded matrices which are block factorizable the permanent can be computed in polynomial time.

The study of graph matchings can be naturally\(^9,10\) formalized by introducing anticommuting variables, so that the edges in a matching cannot overlap due to the Pauli exclusion principle. Creutz introduced an efficient algorithm for Grassmann integration\(^11\). In Ref.\(^12\) the Creutz algorithm is applied to a graph coloring problem.

In this paper, we shall present a simplified form of Creutz algorithm, in which Grassmann integration reduces to simple polynomial manipulations.

A hard object is represented as a product of even elements $\eta_i$ of a Grassmann algebra, associated to the nodes $i$ of the graph on which the objects lie. The $\eta_i$ elements are commuting and nilpotent and are represented as products $\eta_i = \bar{\theta}_i \theta_i$ of anticommuting variables $\theta_i, \bar{\theta}_i$. In the case of dimer systems, this notation was introduced\(^9,13\) as a starting point to
deduce the free-fermion interpretation of the close-packed dimer model on planar lattices.

The generating function that counts the hard objects is a Grassmann integral of a product on these objects.

Our algorithm to compute the generating function of the sums of the permanental minors of a matrix of order $n$ has time complexity $O(2^n n^3)$, while for the Brualdi-Ryser formula the complexity is larger than $O(2^{5n})$. In the case of banded matrices with a fixed band-width $w$ the former algorithm has quadratic complexity in $n$, $O(2^{2w}(w+1)n^2)$, the latter exponential.

In the case of dimers, the $\eta$ elements are associated to the end-point of the dimer and we obtain efficient prescriptions to compute the matching polynomial for both bipartite and non-bipartite graphs.

Another important graph polynomial associated to $G$, the independence polynomial, $I(t) = \sum_{i=0}^{\infty} a(i) t^i$, with $a(i)$ the number of independent subsets of $i$ vertices in $V$, can be similarly derived. In this case the hard object is represented by $\eta$ elements associated to the edges adjacent to a vertex.

Appendix A tabulates the values of the Witten indices in the cases of square and hexagonal lattices with relatively large size, for some of which we disagree with the results of the calculation in Ref.\[15\].

We provide an implementation of these algorithms in Python; examples of its usage are given in Appendix B.

II. AN ALGEBRAIC FORMALISM FOR COUNTING HARD OBJECTS ON A GRAPH

Define a “hard object” $a$ on a graph $G$ with set of vertices $V$ and of edges $E$ as a subset $V_a$ of the vertices in $V$. This is a generalization of the notion of dimer. Configurations of two or more hard objects onto $G$ are admissible provided they have no common vertices (i.e. the vertex subsets of the various objects are “independent”).

Let us associate to each object $a$ the expression

$$O_a = 1 + w_a \prod_{i \in V_a} \eta_i$$

(2)

where $w_a$ is a weight factor, and the product runs on a set of elements $\eta_i = \bar{\theta}_i \theta_i$, where $\theta_i$, $\bar{\theta}_i$ are Grassmann anticommuting variables.
Define
\[ < A > := \int \prod_{i=1}^{n} d\bar{\theta}_i \exp(\sum \bar{\theta}_i \theta_i) A \] (3)
where the Berezin integration\cite{footnote1} over anticommuting variables is used.

The \( \eta \)-elements satisfy the rules
\[ \eta_i^2 = 0 \] (4)
\[ \eta_i \eta_j = \eta_j \eta_i \] (5)
\[ < \eta_{i_1} \ldots \eta_{i_k} > = 1 \] (6)
when \( i_1, \ldots, i_k \) are all distinct. Consider now the product of all admissible objects onto the graph \( G \)
\[ Z_G = < \prod_a O_a > \] (7)

One can write
\[ Z_G = < \prod_a (1 + w_a \prod_{i \in V_a} \eta_i) >= \int d\theta d\bar{\theta} \exp(S) \] (8)
with
\[ S = \sum_{i \in V} \bar{\theta}_i \theta_i + \sum_a w_a \prod_{i \in V_a} \bar{\theta}_i \theta_i \] (9)

If \( w_a = t \) for all \( a \), where \( t \) is a variable, \( Z_G(t) \) is the generating function of the number of ways to settle the hard objects onto the graph.

Let us observe that after performing a partial product \( \prod_b O_b \), if an element \( \eta_i \) does not occur in the remaining products in \( Z_G \), then one can replace \( \eta_i \) with 1. This reduces the number of possible monomials, thus simplifying the product. To save memory and improve performance, the product should be ordered in such a way that only few elements \( \eta_i \) are present for any partial product. This algorithm is a simplified version of Creutz algorithm\cite{footnote11}, since the Grassmann variables \( \theta_i \) and \( \bar{\theta}_i \) appear only in \( \eta_i \), we can avoid introducing the Fock space for fermionic operators and use only simple polynomial manipulations.

A. Sums of permanental minors

The permanent of a \( n \times n \) matrix \( A \) is the coefficient\cite{footnote18} of the \( x_1 \ldots x_n \) monomial in
\[ \prod_{i=1}^{n} \sum_{j=1}^{n} A_{ij} x_j \] (10)
so obviously

\[ \text{perm}(A) = < (\sum_{i_1} A_{1,i_1} \eta_{i_1}) (\sum_{i_2} A_{2,i_2} \eta_{i_2}) \ldots >. \] (11)

The sum of permanental \( k \)-minors of a square matrix of size \( n \) is

\[ p_k(A) = \sum \text{perm}(A_{r,s}) \] (12)

where \( r, s \) are all the order \( n - k \) subsets of \( (1, \ldots, n) \) and \( A_{r,s} \) is the minor obtained eliminating the rows \( r \) and the columns \( s \). Using directly this formula, since the complexity for computing the permanent of \( A \) using the Ryser algorithm is \( O(2^n n) \), the complexity for computing the case \( k = n/2 \) for \( n \) even is \( (\binom{n}{n/2})^2 O(2^{n/2} n) \simeq O(2^{5n/2}) \).

The generating function of the sums of permanental minors of the \( m \times n \) matrix \( A \) is

\[ \sum_k p_k(A) t^k = < (1 + t \sum_{i_1=1}^n A_{1,i_1} \eta_{i_1}) (1 + t \sum_{i_2=1}^n A_{2,i_2} \eta_{i_2}) \ldots (1 + t \sum_{i_m=1}^n A_{m,i_m} \eta_{i_m}) > \] (13)

To compute Eq.(13), after evaluating the \( i \)-th partial product, there are \( 2^i \) monomials in \( \eta \), each of them multiplied by a polynomial of degree at most \( i \) in \( t \). Multiplying by \( (1 + t \sum_{j=1}^n A_{i,j} \eta_j) \) and expanding the product one gets \( i2^n n \) terms, so that the complexity of computing the generating function for the sum of permanental minors is \( O(2^n m^2 n) \). In all the estimates of the time complexity, we have neglected the contribution of number multiplication.

In the case of a square band matrix of size \( n \) with entries \( M_{ij} = 0 \) for \(|i - j| > w\), at the end of the \( i \)-th partial product, the number of \( \eta \)-elements is \( \nu = \min(2w, n) \), so that there are \( 2^{\min(2w,n)} \) monomials in \( \eta \), each multiplied by a polynomial of degree \( i \); the computational complexity is \( O(2^{\min(2w,n)} (w + 1) n^2) \) i.e., for fixed \( w \), it is polynomial in \( n \), while the algorithm in Ref.[7] is exponential.

If one is interested only in computing the permanent using Ryser algorithm[17], the complexity is \( O(n2^n) \), even in the case of banded matrices, while with our algorithm it is \( O(n^2) \).

If the matrix is “almost banded”, i.e. it has \( h \) non-zero elements elements outside the band, with \( h \) small, we have to replace \( 2w \) with \( 2w + h \) in the above estimate of the complexity.
B. Matching generating polynomial

The matching generating polynomial of $G$, i.e. the generating function of the number $N(i)$ of different matchings of $G$ containing $i$ edges, is defined by

$$M(t) = \sum_{i=0}^{[v/2]} N(i) t^i$$

(Equivalently one defines the matching polynomial

$$\mu(t) = \sum_{i=0}^{[v/2]} (-1)^i N(i) t^{v-2i}$$

related to the former by $\mu(x) = x^v M(-x^{-2})$)

Consider now Eq.(7) in the case of dimers:

$$Z_G = \langle \prod_{<i,j>} (1 + w_{i,j} \eta_i \eta_j) \rangle$$

The matching generating polynomial $M(t)$ is obtained from $Z_G$, setting $w_{i,j} = t$.

To make clear by an example the algebraic manipulations used, consider the graph $A$ in Fig.1.

Evaluate the partial product $O_0 O_1 O_2$ ($O_6$ is associated to the edge $E_0$ etc.): one has

$$M_G = \langle 1 + 2t + 2t^2 \eta_3 + (t + t^2) \eta_4 + t^2 \eta_3 \eta_4 \rangle$$

In the last step $\eta_0$ has been replaced with 1 since it does not occur in the remaining terms $O_2, ..., O_6$. Similarly in the next step, after expanding the partial product $O_0 O_1 O_2$ we can set $\eta_1 = 1$

$$M_G = \langle 1 + t + t^2 \eta_2 + t \eta_3 + t^2 \eta_2 \eta_3 \rangle$$

In the partial product $O_0 O_1 O_2 O_3$, a term $(t + t^2) \eta_2 \eta_3$ is added. After expanding the partial product $O_0 O_1 O_2 O_3 O_4$, we can set $\eta_2 = 1$

$$M_G = \langle 1 + 2t + (2t + 2t^2) \eta_3 + (t + t^2) \eta_4 + t^2 \eta_3 \eta_4 \rangle$$

In the partial product $O_0 O_1 O_2 O_3 O_4 O_5$, we can set $\eta_3 = 1$:

$$M_G = \langle 1 + 2t + 2t^2 + (t + t^2) \eta_4 + (t + t^2) \eta_5 + (t^2 + t^3) \eta_4 \eta_5 \rangle$$
Finally

\[ M_G = 1 + 7t + 11t^2 + 3t^3 \]  \hspace{1cm} (17)

In each step there are at most two \( \eta \) elements.

Given a graph \( G \) with \( V \) vertices and \( E \) edges, start with the empty graph \( G_0 \) on \( V \)
vertices, add an edge to get \( G_1 \); then continue to add edges, until \( G_E = G \). For a graph \( G_i \) in this sequence, an “active node” is by definition a node which is incident with at least one edge, and has a degree less than the degree that the node has in \( G \). The active node number \( \nu \) is the maximum number of active nodes in the sequence \( G_0, ..., G_E \). In general the size of the computer memory used by the algorithm grows with a factor \( 2^\nu \).

The graph in the above example has active node number \( \nu = 2 \).

The time complexity for computing the matching polynomial for graphs with small active node number is \( O(2^{\nu^3}) \), analogously to the case of the sums of permanental minors for band matrices; the space complexity is \( O(2^{\nu^2}) \). For fixed \( \nu \), the complexity is polynomial in \( \nu \). Therefore one can deal with large graphs, provided \( \nu \) is small.

We have not yet devised a general prescription to determine an ordering for which \( \nu \) is close to minimum. A simple greedy procedure to get a sequence with small (but generally non optimal) \( \nu \) is the following: as long as it is possible, add an edge at the time without increasing the value of \( \nu \); otherwise add an edge of one among the shortest paths in \( G - \) \((\text{non-active vertices})\), which join active vertices.

As an example of a sequence of random graphs with fixed \( \nu \), take a sequence of regular bipartite graphs constructed in the following way. Let \( G_0 \) be a cycle with \( k \) vertices, with \( k \) even. Add another cycle with \( k \) vertices; the odd (even) vertices of this cycle are linked respectively to the even (odd) vertices of the previous cycle in a random way, obtaining \( G_1 \); continue adding cycles in this way, obtaining the sequence \( G_i \). The sequence of the corresponding regular bipartite graphs is obtained linking the vertices of the first and last cycle. Since the active node number is \( \nu = 2k \) (the vertices on the first and the last cycle are active), computing the matching polynomials for a sequence of \( N \) cycle graphs takes \( O(N^2) \). The code for computing a sequence with \( k = 6 \) is included in the examples reported in Ref. [20].

In Ref. [11] an ordering of fermionic variables, with the insertion of a projector excluding a fermionic operator, is similarly chosen. Since this formalism is applied to fermions on a finite square lattice (a grid), there is a natural way to establish a longitudinal direction and a transverse direction. Only fermionic modes on the transverse direction appear in the computation, so one can deal with long grids with few modes in the transverse direction.
Let us observe that from Eq.(16), distributing a term $(1 + w_{i,j} \eta_i \eta_j)$ we get

$$Z_G(t) = \prod_{<i,j> \neq <k,l>}(1 + w_{i,j} \eta_i \eta_j) + w_{k,l} \prod_{<i,j>, i \neq k, j \neq l}(1 + w_{i,j} \eta_i \eta_j)$$

which gives a recursion relation for matching generating polynomials

$$Z_G(t) = M_{G-<k,l>}(t) + t M_{G-k-l}(t)$$

If the graph $G$ is bipartite, let us indicate by $y_i$ the elements $\eta_i$ associated to the even sites, with $\eta_i$ those associated to the odd sites; then

$$Z_G = \langle (1 + \sum_{i_1} y_{1i} w_{1i} \eta_{i_1})(1 + \sum_{i_2} y_{2i} w_{2i} \eta_{i_2})... \rangle$$

Since the $y_{ij}$ element occurs only in the $j$th term of the product, it can be set equal to unity, so that

$$Z_G = \langle (1 + \sum_{i_1} w_{1i} \eta_{i_1})(1 + \sum_{i_2} w_{2i} \eta_{i_2})... \rangle$$

which gives Eq.(13) in the case $w_{i,j} = tA_{i,j}$.

As an application, we have computed $M(-1)$ for some periodical square lattices and for some hexagonal lattices in the brick-wall representation considered in Ref.[15]; we disagree with Ref.[15] in some cases, see Appendix A.

C. Independence polynomial

The independence polynomial $I_G(t) = \sum_i a(i)t^i$ is the generating function for the number $a(i)$ of ways of choosing $i$ independent vertices on $G$. A hard object is made associating to a vertex the product of the $\eta$ elements on the edges incident with that vertex. The greedy algorithm for ordering the product consists in choosing a short path in $G - (non - active vertices)$.

The matching generating polynomial of a graph $G$ is the independence polynomial of the line graph of $G$.

As an example, consider the line graph of the graph $B$ in Fig[1]. Evaluate the partial product $O_9O_1$, set $\eta_8 = 1$

$$I = I(L(G)) = \langle O_0...O_6 \rangle = \langle (1 + t\eta_9 + t\eta_4 \eta_7)O_2...O_6 \rangle$$

$$I = \langle (1 + t\eta_9 + t\eta_4 \eta_7)O_2...O_6 \rangle$$

$$I = \langle (1 + t\eta_9 + t\eta_4 \eta_7)O_2...O_6 \rangle$$
Evaluate the partial product $O_0 O_1 O_2$, set $\eta_0 = 1$

$$I = < (1 + t + t \eta_4 \eta_7 + t \eta_5 \eta_6 + t^2 \eta_4 \eta_5 \eta_6 \eta_7) O_3 ... O_6 >$$

Evaluate the partial product $O_0 O_1 O_2 O_3$, set $\eta_6 = \eta_7 = 1$

$$I = < (1 + t + t \eta_4 + t \eta_5 + (t + t^2) \eta_2 \eta_3 + (t + t^2) \eta_4 \eta_5 \eta_6 \eta_7) O_4 ... O_6 >$$

Evaluate the partial product $O_0 O_1 O_2 O_3 O_4$, set $\eta_3 = \eta_4 = 1$

$$I = < (1 + 2 t + (t^2 + t) \eta_0 + (t + t^2) \eta_2 + (t + t^2) \eta_5 + t^2 \eta_0 \eta_5) O_5 O_6 >$$

Evaluate the partial product $O_0 O_1 O_2 O_3 O_4 O_5$, set $\eta_2 = \eta_5 = 1$

$$I = < 1 + 4 t + 2 t^2 + (t + 2 t^2) \eta_0 + (t + 2 t^2) \eta_1 + (t^2 + t^3) \eta_0 \eta_1) O_6 >$$

Finally one gets the same as in Eq. (17).

As a check, we computed $I(-1)$ for the hexagonal lattices in the brick-wall representation considered in table VII of Ref. [15], which can be interpreted as the Witten index of the quantum hexagonal model.

As another application, we computed $I(1)$ for square grids of size up to $35 \times 35$. The results agree with Ref. [18] where results are reported up to the size $33 \times 33$.

For the $34 \times 34$ square grid, we get

$I_{34 \times 34}(1) = 387891128933234889019525245048798489818497881776634515543429025520$

$6346721638717020250480108304893087882913564262766592538500796108515840997971$

$52548773065607505250668587876084152495126750481594564582092827282$

In the $35 \times 35$ case, we get

$I_{35 \times 35}(1) = 7212249471271214286776360359845549941067616972563902046316263757$

$753676843828248033036148852945185908035253110516357208809079131881311521646$

$489569003946048223764207235363675779986619848116510736835320875797768521522$

$195$

In theoretical chemistry $I(1)$ is called the Merrifield-Simmons index\(^{24}\). In Appendix B we have computed the Merrifield-Simmons index of the Buckminster fullerene.

### III. CONCLUSIONS

We have shown that a simplified version of Creutz algorithm can be used to compute sums of permanental minors, matching and independence polynomials. In the case of the sums
of permanental minors, we have shown that this algorithm has lower complexity than using
the Brualdi-Ryser formula. The algorithms are in general exponential, but they can become
polynomial in particular cases. For example, sums of permanental minors have polynomial
complexity if the matrix is banded. It is then important to be able to recognize whether
a matrix can be brought to banded form permuting its rows and its columns. A similar
ordering problem is met when computing the matching and independence polynomials. We
did not address the problem of finding an optimal ordering: presumably it is related to the
tree decomposition of graphs\textsuperscript{19}.

IV. APPENDIX A: THE WITTEN INDEX FOR RECTANGULAR AND HEXAGONAL
PERIODIC LATTICES

In Ref.\textsuperscript{[15]} the Witten index $W = \sum (-1)^i N(i)$ is evaluated for the supersymmetric dimer
model. For the largest lattices we agree with these results only modulo $2^{32}$. We think it
likely that in Ref.\textsuperscript{[15]} the large integer arithmetic was inadequately managed. We checked
only the cases $m \times n$ for $m, n \geq 4$ and even. The disagreeing values are listed in Tables I
and II.

TABLE I: Comparison of the values of the Witten index $W(G)$ for a square grid $G$ of size $m \times n$
obtained by the algorithm introduced in this paper with the disagreeing results in table III in
Ref.\textsuperscript{[15]}.

| $m$ | $n$ | $W$ | $|W|^{(1/(mn))}$ | $W$ of Ref.\textsuperscript{[15]} | $|W|^{(1/(mn))}$ in Ref.\textsuperscript{[15]} |
|-----|-----|-----|-----------------|-----------------|-----------------|
| 10  | 8   | -14550253471 | 1.340           | -1665351583     | 1.304           |
| 10  | 10  | 3235851927936 | 1.334           | 1741554048      | 1.237           |

We have computed the index $W(G)$ also for the larger lattices indicated in Table II. The

TABLE II: The value of the Witten index $W$ for a square grid of size $m \times n$ larger than those
considered in Ref.\textsuperscript{[15]}.

| $m$ | $n$ | $W$ | $|W|^{(1/(mn))}$ |
|-----|-----|-----|-----------------|
| 12  | 10  | -139080563404700 | 1.312           |
| 12  | 12  | 988571682202805376 | 1.333           |
quantity \( |W|^{(1/(mn))} \) should be compared with the expression \( W = 2r^{mn}\cos (mn\theta + \theta_0) \) of Ref.\[15\], where \( r = 1.33 \pm 0.01 \).

We have compared our evaluations of the Witten indices with the results in Ref.\[15\] also in the case of hexagonal lattices. Our results agree only modulo \( 2^{32} \) with Ref.\[15\] (table VIII in that reference); the disagreeing values are shown in Table [III].

**TABLE III:** The values of the Witten index \( W \) for hexagonal grids of size \( m \times n \), disagreeing with those considered in Ref.\[15\].

| \( m \) | \( n \) | \( W \) | \( W \) of Ref.\[15\] |
|---|---|---|---|
| 10 | 14 | 7711439360 | -878495232 |
| 10 | 16 | -655517342208 | 1612654080 |
| 12 | 12 | 94909515776 | 420235264 |
| 12 | 14 | 6459966411264 | 335598080 |
| 12 | 16 | 100182729294336 | -1677852160 |
| 14 | 10 | 11948085184 | -936816704 |
| 14 | 12 | 6736033699456 | 1524979328 |
| 14 | 14 | 742553681809408 | 1080971264 |
| 14 | 16 | -1384901745575424 | 1869085184 |

We have computed the index \( W \) also for the larger lattice indicated in Table [IV]. For \( 10 \leq m, n \leq 16 \) the Witten index per site \( |W|^{(1/(mn))} \) is between 1.156 and 1.192 with an average value 1.18, which lies below the interval \( r = 1.4 \pm 0.1 \) reported in Ref.\[15\].

**TABLE IV:** The value of the Witten index \( W \) for hexagonal grids of size \( m \times n \) larger than those considered in Ref.\[15\].

| \( m \) | \( n \) | \( W \) |
|---|---|---|
| 16 | 14 | -119633551609600 |
| 16 | 16 | -600918132969537536 |

10 \leq m, n \leq 16 the Witten index per site \( |W|^{(1/(mn))} \) is between 1.156 and 1.192 with an average value 1.18, which lies below the interval \( r = 1.4 \pm 0.1 \) reported in Ref.\[15\].
V. APPENDIX B: USAGE OF THE PYTHON MODULE “HOBJ”

The module “hobj” can be downloaded from Ref. [20]. It can be used without any dependence; the code for univariate polynomials, represented as arrays, is adapted from SymPy [21]. Here is the example graph A in Figure 1:

```python
>>> from hobj import dup_matching_generating_poly
>>> d = {0: [1, 2], 1: [0, 3], 2: [0, 3, 4], 3: [1, 2, 5], 4: [2, 5], 5: [3, 4]}
>>> dup_matching_generating_poly(d)
[3, 11, 7, 1]
```

Since it is a bipartite graph, one can compute it also using the reduced adjacency matrix:

```python
>>> from hobj import dup_permanental_minor_poly
>>> from domains import ZZ
>>> m = [[1, 1, 0], [1, 1, 1], [0, 1, 1]]
>>> dup_permanental_minor_poly(m, ZZ)
[3, 11, 7, 1]
```

In the case of bipartite graphs the second way is often faster.

The following examples take a fraction of a second on a current personal computer. Compute the sum of the permanental minors of a banded matrix:

```python
>>> from hobj import dup_permanental_minor_poly
>>> from domains import ZZ
>>> m = [[i*j if abs(i-j) < 6 else 0 for i in range(20)] for j in range(20)]
>>> sum(dup_permanental_minor_poly(m, ZZ))
11936810897247956264161397956481650508142206788L
>>> dup_permanental_minor_poly(m, ZZ, 1)
11936810897247956264161397956481650508142206788L
```

the second way is faster and uses less memory because it avoids constructing the polynomial. Similarly in the following examples.

Let us call “hobj” from Sage [22] and compute the sum of the coefficients of the matching polynomial for the Buckminster fullerene $C_{60}$ (truncated icosahedron) computed first in Ref. [23]
sage: from hobj import dup_matching_generating_poly
sage: d = graphs.BuckyBall().to_dictionary()
sage: sum(dup_matching_generating_poly(d))
1417036634543488
sage: dup_matching_generating_poly(d, val=1)
1417036634543488

Same for the independence polynomial

sage: from hobj import dup_independence_poly
sage: sum(dup_independence_poly(d))
217727997152
sage: dup_independence_poly(d, val=1)
217727997152
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