# Estimating $L^{\infty}$ Norms by $L^{2 k}$ Norms for Functions on Orbits 

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#### Abstract

Let $G$ be a compact group acting in a real vector space $V$. We obtain a number of inequalities relating the $L^{\infty}$ norm of a matrix element of the representation of $G$ with its $L^{2 k}$ norm for a positive integer $k$. As an application, we obtain approximation algorithms to find the maximum absolute value of a given multivariate polynomial over the unit sphere (in which case $G$ is the orthogonal group) and for the assignment problem of degree $d$, a hard problem of combinatorial optimization generalizing the quadratic assignment problem (in which case $G$ is the symmetric group).


## 1. Introduction

A general optimization problem has to do with finding the maximum (minimum) value of a real-valued function $f: X \rightarrow \mathbb{R}$. Often, the set $X$ is endowed with a probability measure $\mu$ and the function $f$ possesses a certain degree of symmetry which allows one to compute the $k$ th moment $\int_{X} f^{k} d \mu$ efficiently at least for small values of $k$. Thus one may ask how well the $k$ th moment approximates the maximum value. In this paper, we describe a fairly general situation where some simple and meaningful relations between the maximum and moments can

[^0]be obtained. We provide two illustrations: one, continuous, has to do with optimization of multivariate polynomials on the unit sphere with possible applications to solving systems of real polynomial equations and the other, discrete, deals with optimization on the symmetric group, namely, with the assignment problem of degree $d$, a hard problem of combinatorial optimization.

### 1.1. The General Setting

Let $G$ be a compact group with the Haar probability measure $d g$ acting in a finitedimensional real vector space $V$. To avoid dealing with various technical details, we assume that the representation $G \rightarrow G L(V)$ is continuous, where the general linear group $G L(V)$ is considered in its standard topology.

Let us choose a vector $v \in V$ and a linear function $\ell: V \rightarrow \mathbb{R}$. We consider the orbit $\{g v: g \in G\}$ of $v$ and the resulting function $f: G \rightarrow \mathbb{R}$ defined by

$$
f(g)=\ell(g v)
$$

In other words, $f$ is a matrix element in the representation of $G$. We are interested in the relation between the following quantities:
The $L^{\infty}$ norm of $f$ :

$$
\|f\|_{\infty}=\max _{g \in G}|f(g)|=\max _{g \in G}|\ell(g v)| .
$$

The $L^{2 k}$ norm of $f$ for a positive integer $k$ :

$$
\|f\|_{2 k}=\left(\int_{G} f^{2 k}(g) d g\right)^{1 / 2 k}=\left(\int_{G} \ell^{2 k}(g v) d g\right)^{1 / 2 k}
$$

As we remarked earlier, for many examples in computational mathematics, the quantity $\|f\|_{\infty}$ is of considerable interest and is hard to compute, whereas $\|f\|_{2 k}$ is relatively easy to compute for moderate values of $k$. First, we relate $\|f\|_{\infty}$ and $\|f\|_{2}$.

Theorem 1. Let $G$ be a compact group acting in a finite-dimensional real vector space $V$ and let dg be the Haar probability measure on $G$. Let us fix a vector $v$ and a linear function $\ell: V \rightarrow \mathbb{R}$ and let us define a real-valued function $f: G \rightarrow \mathbb{R}$ by $f(g)=\ell(g v)$. Then

$$
\|f\|_{2} \leq\|f\|_{\infty} \leq \sqrt{\operatorname{dim} V} \cdot\|f\|_{2} .
$$

The bounds of Theorem 1 are generally sharp, see Section 2.1. To estimate how well $\|f\|_{2 k}$ approximates $\|f\|_{\infty}$ for a larger $k$, we invoke a general construction from the representation theory, see, e.g., Fulton and Harris [6, Lecture 6].

### 1.2. Tensor Power

For a positive integer $k$, let

$$
V^{\otimes k}=\underbrace{V \otimes \cdots \otimes V}_{k \text { times }}
$$

be the $k$ th tensor power of $V$. There is a natural action of $G$ in $V^{\otimes k}$, defined on decomposable tensors by

$$
g\left(v_{1} \otimes \cdots \otimes v_{k}\right)=g v_{1} \otimes \cdots \otimes g v_{k} \quad \text { for } \quad g \in G
$$

There is a natural action of the symmetric group $S_{k}$ permuting the components of the tensor product. Thus, for decomposable tensors, we have

$$
\sigma\left(v_{1} \otimes \cdots \otimes v_{k}\right)=v_{\sigma^{-1}(1)} \otimes \cdots \otimes v_{\sigma^{-1}(k)} \quad \text { for } \quad \sigma \in S_{k}
$$

The action of $S_{k}$ commutes with the action of $G$. Let $\operatorname{Sym}_{k}(V)$ be the symmetric part of $V^{\otimes k}$ consisting of the tensors $x$ such that $\sigma x=x$ for all $\sigma \in S_{k}$. It is known that

$$
\operatorname{dim} \operatorname{Sym}_{k}(V)=\binom{\operatorname{dim} V+k-1}{k}
$$

since $\operatorname{Sym}_{k}(V)$ can be thought of as the space of all real homogeneous polynomials of degree $k$ in $\operatorname{dim} V$ variables. Let

$$
v^{\otimes k}=\underbrace{v \otimes \cdots \otimes v}_{k \text { times }}
$$

be the $k$ th tensor power of $v$. Thus $v^{\otimes k} \in \operatorname{Sym}_{k}(V)$ and $g v^{\otimes k} \in \operatorname{Sym}_{k}(V)$ for all $g \in G$. It turns out that how well $\|f\|_{2 k}$ approximates $\|f\|_{\infty}$ depends on the dimension $D_{k}$ of the subspace spanned by the orbit $\left\{g v^{\otimes k}\right\}$. This dimension may be different for different $v \in V$. Roughly, if $D_{k}$ is small then $v$ lies in a certain algebraic variety constructed from the action of $G$ in $V$ and for such $v$ the functions $f$ are "smoother" than for those $v$ for which $D_{k}$ is large.

Thus we obtain the following corollary of Theorem 1 :
Corollary 2. Let $G$ be a compact group acting in a finite-dimensional real vector space $V$ and let dg be the Haar probability measure on G. Let us fix a vector $v$ and a linear function $\ell: V \rightarrow \mathbb{R}$ and let us define a real-valued function $f: G \rightarrow \mathbb{R}$ by $f(g)=\ell(g v)$. For a positive integer $k$, let

$$
D_{k}=\operatorname{dim} \operatorname{span}\left\{g v^{\otimes k}: g \in G\right\}
$$

be the dimension of the span of the orbit of $v^{\otimes k}$ in $V^{\otimes k}$. Then

$$
\|f\|_{2 k} \leq\|f\|_{\infty} \leq\left(D_{k}\right)^{1 / 2 k} \cdot\|f\|_{2 k}
$$

Again, generally speaking, the estimates of Corollary 2 cannot be improved.
H. Derksen informed the author that numbers $D_{k}$ turn out to be quite useful for the constructive invariant theory, see Derksen and Kraft [5].

A straightforward estimate of $D_{k} \leq \operatorname{dim} \operatorname{Sym}_{k}(V)$ produces the following corollary:

Corollary 3. Let $G$ be a compact group acting in a finite-dimensional real vector space $V$ and let $d g$ be the Haar probability measure on $G$. Let us fix a vector $v$ and a linear function $\ell: V \rightarrow \mathbb{R}$ and let us define a real-valued function $f: G \rightarrow \mathbb{R}$ by $f(g)=\ell(g v)$. Let $k$ be a positive integer. Then

$$
\|f\|_{2 k} \leq\|f\|_{\infty} \leq\binom{\operatorname{dim} V+k-1}{k}^{1 / 2 k} \cdot\|f\|_{2 k}
$$

There are examples showing that the bounds of Corollary 3 are "almost tight." For instance, if $G=S O(n)$ is the orthogonal group acting in $V=\mathbb{R}^{n}$, computations of Section 3.1 show that the upper bound for $\|f\|_{\infty}$ is tight up to a factor of $\sqrt{2}$ (uniformly on $k$ and $n$ ).

It follows from Corollary 3 that $\|f\|_{2 k} /\|f\|_{\infty} \rightarrow 1$ as long as $(\operatorname{dim} V) / k \rightarrow 0$. As we remarked earlier, in many cases we are able to compute $\|f\|_{2 k}$ efficiently if $k$ is not very large. Quite often (see examples of Sections 3 and 4), we can compute $\|f\|_{2 k}$ in polynomial time for any fixed $k$. The following estimate shows the type of bound that we can achieve if we fix $k$ in advance.

Corollary 4. For any $\varepsilon>0$ there exists a $k_{0}=k_{0}(\varepsilon)=O\left(\varepsilon^{-2}\right)$ such that for any positive integer $k>k_{0}$, for any compact group $G$ acting in a real vector space $V$ with $\operatorname{dim} V \geq k$, for any linear function $\ell: V \rightarrow \mathbb{R}$, for any $v \in V$ and for the function $f(g)=\ell(g v), f: G \rightarrow \mathbb{R}$, we have

$$
\|f\|_{2 k} \leq\|f\|_{\infty} \leq \varepsilon \sqrt{\operatorname{dim} V} \cdot\|f\|_{2 k}
$$

The paper is structured as follows. In Section 2, we prove Theorem 1 and Corollaries 2-4. In Section 3, we apply our results to the problem of finding the largest absolute value of a real homogeneous multivariate polynomial (form) on the unit sphere, in which case $G=S O(n)$, the orthogonal group. In particular, we present a simple polynomial time approximation algorithm to compute the largest absolute value on the sphere of a fewnomial, that is, a polynomial having only a small (fixed) number of monomials. More generally, our method is applicable to forms with small (fixed) dimensions of their Newton polytopes. The results have potential applications to constructing efficient algorithms for testing the feasibility of a given system of real polynomial equations, since the latter problem can be reduced to estimating the maximum absolute value on the unit sphere of an appropriately constructed "penalty" form $f$.

In Section 4, we discuss an NP-hard problem of combinatorial optimization, which we call the assignment problem of degree $d$, in which case $G=S_{n}$, the
symmetric group. In particular, our results lead to an approximation algorithm for finding a bijection between vertex sets of two hypergraphs $H_{1}$ and $H_{2}$, which maximizes the number edges of $H_{1}$ mapped onto the edges of $H_{2}$. For $d=2$, the corresponding optimization problem is known as the quadratic assignment problem, see, e.g., Çela [4]. In the quadratic assignment problem, the objective function $f: S_{n} \rightarrow \mathbb{R}$ is defined by $f(\sigma)=\sum_{i, j} a_{i j} b_{\sigma(i) \sigma(j)}$, where $A=\left(a_{i j}\right)$ and $B=\left(b_{i j}\right)$ are two given matrices. Interpreting $f$ as a matrix element in a representation of $S_{n}$, for any given $\varepsilon>0$, we obtain a polynomial time algorithm approximating $\|f\|_{\infty}=\max _{\sigma \in S_{n}}|f(\sigma)|$ within a factor of $\varepsilon n$.

We use the real model for computational complexity, counting the number of arithmetic operations performed by the algorithm, see Blum et al. [3]. Eventually, to compute $\|f\|_{2 k}$ from $\|f\|_{2 k}^{2 k}$, we need to extract a root of degree $2 k$, which we count as a single operation.

## 2. Proofs

In this section, we prove Theorem 1 and Corollaries $2-4$. We need some standard facts from the representation theory, see, e.g., Fulton and Harris [6].

Let $G$ be a compact group acting in a finite-dimensional real vector space $V$. As is known, $V$ possesses a $G$-invariant scalar product $\rangle$ :

$$
\langle u, v\rangle=\langle g u, g v\rangle \quad \text { for all } \quad u, v \in V \quad \text { and all } \quad g \in G .
$$

We introduce the corresponding Euclidean norm:

$$
\|x\|=\sqrt{\langle x, x\rangle} .
$$

The action (representation) is called irreducible if $V$ contains no proper $G$-invariant subspaces. As is known, if $G$ acts in a finite-dimensional real vector space $V$, then $V$ can be represented as a direct sum of pairwise orthogonal (with respect to a given $G$-invariant scalar product) invariant subspaces $V_{i}$ such that the action of $G$ in each $V_{i}$ is irreducible.

A somewhat "nonstandard" feature of our construction is that we consider representations over the real, rather than over the complex, numbers. The same results can be obtained by using the more standard representation theory over $\mathbb{C}$, but switching from real to complex numbers and back somewhat obscures the main simple ideas of the proof and makes it harder to see when the obtained bounds are sharp.

We need a "real" substitute for Schur's lemma. It comes in the form of the following observation. Suppose that $q: V \rightarrow \mathbb{R}$ is a $G$-invariant quadratic form, that is, $q(g x)=q(x)$ for all $x \in V$ and all $g \in G$. We claim that the eigenspaces of $q$ are $G$-invariant subspaces. A possible way to see this is to notice that the unit eigenvectors of $q$ are precisely the critical points of the restriction $q: \mathbb{S} \rightarrow \mathbb{R}$ where $\mathbb{S}=\{x:\|x\|=1\}$ is the unit sphere in $V$.

Our first lemma is a real version of the orthogonality relations for matrix elements.

Lemma 5. Let $G$ be a compact group acting in a finite-dimensional real vector space $V$ endowed with a $G$-invariant scalar product $\rangle$. Suppose that the representation of $G$ is irreducible and let $d g$ be the Haar probability measure on $G$. Then

$$
\int_{G}\langle x, g v\rangle^{2} d g=\frac{\|v\|^{2} \cdot\|x\|^{2}}{\operatorname{dim} V} \quad \text { for all } \quad x, v \in V
$$

Proof. Let us choose a vector $v \in V$ and let us define a quadratic form $q: V \rightarrow \mathbb{R}$ by

$$
q(x)=\int_{G}\langle x, g v\rangle^{2} d g
$$

Clearly, $q(x)$ is $G$-invariant: $q(g x)=q(x)$ for all $x \in V$ and all $g \in G$. Let $\lambda$ be the largest eigenvalue of $q$ and let $W$ be the corresponding eigenspace. Then $W$ is an invariant subspace of $V$ and hence $W=V$. Thus $q(x)=\lambda\|x\|^{2}$ for some $\lambda \geq 0$. To find $\lambda$, let us compute the trace of $q$.

On one hand, we have $\operatorname{tr}(q)=\lambda \operatorname{dim} V$. Let $q_{g}(x)=\langle x, g v\rangle^{2}$. Then $q_{g}$ is a quadratic form of rank 1 with the nonzero eigenvalue $\|g v\|^{2}=\|v\|^{2}$ which corresponds to an eigenvector $x=g v$. Hence $\operatorname{tr}\left(q_{g}\right)=\|v\|^{2}$ and since $q(x)$ is the average of $q_{g}$, we have $\operatorname{tr}(q)=\|v\|^{2}$. Thus $\lambda=\|v\|^{2} / \operatorname{dim} V$. Hence

$$
q(x)=\frac{\|v\|^{2} \cdot\|x\|^{2}}{\operatorname{dim} V}
$$

and the proof follows.
Now we use that every representation is a sum of irreducible representations.
Lemma 6. Let $G$ be a compact group acting in a finite-dimensional real vector space $V$ endowed with a $G$-invariant scalar product $\rangle$. Let dg be the Haar probability measure on $G$. Let us fix a vector $v \in V$. Then there exists a decomposition $V=V_{1} \oplus \cdots \oplus V_{k}$ of $V$ into the direct sum of nonzero pairwise orthogonal invariant subspaces such that for every $x \in V$ we have

$$
\int_{G}\langle x, g v\rangle^{2} d g=\sum_{i=1}^{k} \frac{\left\|x_{i}\right\|^{2} \cdot\left\|v_{i}\right\|^{2}}{\operatorname{dim} V_{i}}
$$

where $x_{i}$ and $v_{i}$ are the orthogonal projections onto $V_{i}$ of $x$ and $v$, respectively.

Proof. Let us define a quadratic form $q: V \rightarrow \mathbb{R}$ by

$$
q(x)=\int_{G}\langle x, g v\rangle^{2} d g .
$$

Then $q$ is $G$-invariant, $q(g x)=q(x)$ for all $g \in G$ and all $x \in V$. Thus the eigenspaces of $q$ are $G$-invariant subspaces of $V$. Let us write every eigenspace as a direct sum of pairwise orthogonal invariant subspaces $V_{i}$ such that the action of $G$ in each $V_{i}$ is irreducible. Thus we obtain the decomposition $V=V_{1} \oplus \cdots \oplus V_{k}$ and we have

$$
q(x)=\sum_{i=1}^{k} \lambda_{i}\left\|x_{i}\right\|^{2}
$$

where $x_{i}$ is the orthogonal projection of $x$ onto $V_{i}$ and $\lambda_{i}$ are nonnegative numbers. To find $\lambda_{i}$, let us choose a nonzero $x \in V_{i}$. Then $\langle x, g v\rangle=\left\langle x, g v_{i}\right\rangle$ and, by Lemma 5, we get

$$
q(x)=\lambda_{i}\|x\|^{2}=\frac{\left\|v_{i}\right\|^{2} \cdot\|x\|^{2}}{\operatorname{dim} V_{i}}
$$

from which

$$
\lambda_{i}=\frac{\left\|v_{i}\right\|^{2}}{\operatorname{dim} V_{i}}
$$

The proof now follows.

Remark. A decomposition $V=V_{1} \oplus \cdots \oplus V_{k}$ of a representation into the direct sum of pairwise orthogonal irreducible components is not unique as long as some irreducible representation appears with a multiplicity greater than 1 (which means that the representations of $G$ in some subspaces $V_{i}$ are isomorphic). One can construct some simple examples showing that the decomposition of Lemma 6 indeed depends on $v$.

Now we are ready to prove Theorem 1.
Proof of Theorem 1. The inequality

$$
\|f\|_{2} \leq\|f\|_{\infty}
$$

is quite standard. Let us prove that

$$
\|f\|_{\infty} \leq \sqrt{\operatorname{dim} V} \cdot\|f\|_{2}
$$

Let $e$ be the identity in $G$. We note that it suffices to prove that

$$
|f(e)|=|\ell(v)| \leq \sqrt{\operatorname{dim} V} \cdot\|f\|_{2}
$$

because the inequality for $f(g)=\ell(g v)$ would follow by choosing a new vector $v$ :

$$
\text { new } v:=g(\text { old } v)
$$

Let us introduce a $G$-invariant scalar product $\rangle$ in $V$ so that $\ell(x)=\langle c, x\rangle$ for some $c \in V$ and all $x \in V$. Applying Lemma 6, we obtain a decomposition $V=V_{1} \oplus \cdots \oplus V_{k}$ of $V$ into the direct sum of pairwise orthogonal invariant subspaces $V_{i}$ such that

$$
\|f\|_{2}^{2}=\int_{G}\langle c, g v\rangle^{2} d g=\sum_{i=1}^{k} \frac{\left\|c_{i}\right\|^{2} \cdot\left\|v_{i}\right\|^{2}}{\operatorname{dim} V_{i}}
$$

where $c_{i}$ and $v_{i}$ are the orthogonal projections onto $V_{i}$ of $c$ and $v$, respectively. We have

$$
f(e)=\langle c, v\rangle=\sum_{i=1}^{k}\left\langle c_{i}, v_{i}\right\rangle
$$

and, hence,

$$
|f(e)| \leq \sum_{i=1}^{k}\left|\left\langle c_{i}, v_{i}\right\rangle\right| \leq \sum_{i=1}^{k}\left\|c_{i}\right\| \cdot\left\|v_{i}\right\|
$$

Let

$$
\alpha_{i}=\frac{\left\|c_{i}\right\| \cdot\left\|v_{i}\right\|}{\sqrt{\operatorname{dim} V_{i}}}
$$

Then

$$
|f(e)|^{2} \leq\left(\sum_{i=1}^{k} \alpha_{i} \sqrt{\operatorname{dim} V_{i}}\right)^{2} \leq\left(\sum_{i=1}^{k} \alpha_{i}^{2}\right)\left(\sum_{i=1}^{k} \operatorname{dim} V_{i}\right)=(\operatorname{dim} V) \cdot\|f\|_{2}^{2}
$$

and the proof follows.

### 2.1. When Is the Upper Bound for $\|f\|_{\infty}$ Attained?

Analyzing the proof of Theorem 1, it is not hard to find out when the bound $\|f\|_{\infty} \leq \sqrt{\operatorname{dim} V} \cdot\|f\|_{2}$ is sharp. For example, if the action of $G$ is irreducible, it follows by Lemma 5 that the equality $\|f\|_{\infty}=\sqrt{\operatorname{dim} V} \cdot\|f\|_{2}$ holds if and only if $\ell(x)$ is proportional to the scalar product of $x$ with one of the orbit vectors $g v$. Generally, the bound is sharp for the class of linear functions on the orbit of $v$ as long as the orbit of $v$ spans $V$. Here are some natural cases when the bound is attained.

Suppose, for example, that we have an absolutely irreducible representation $\rho$ of $G$ in a real vector space $W$ (i.e., the representation remains irreducible after complexification). Thus, for every $g \in G, \rho(g)$ is an operator in $W$. We interpret $\rho(g)$ as a point in the space $V=\operatorname{End}(W)=W^{*} \otimes W$ of all linear transformations $W \rightarrow W$. Let $\chi(g)=\operatorname{tr}(\rho(g))$ be the character of the representation. We think of $\chi(g)$ as of a linear function on the orbit of the identity operator $I \in \operatorname{End}(W)$ under
the action $g(x)=\rho(g) x$ for all $x \in \operatorname{End}(W)$ (in other words, $G$ acts identically in $W^{*}$ and by $\rho$ in $W$ in the tensor product $W^{*} \otimes W$ ). Then $\|\chi\|_{\infty}=\operatorname{dim} W=$ $\sqrt{\operatorname{dim} V}$. The orthogonality relations for the characters, see, e.g., Fulton and Harris [6, Lecture 2], state that $\|\chi\|_{2}=1$ and hence the bound of Theorem 1 holds with equality.

As another example, let us consider a finite group $G$ of cardinality $|G|$ and an arbitrary function $f: G \rightarrow \mathbb{R}$. The function $f$ can be thought of as a linear function on the orbit of a point in the regular representation of $G$. The space $V$ in this case is the vector space of all linear functions $f: G \rightarrow \mathbb{R}$ where $G$ acts by shifts: $g f(x)=f\left(g^{-1} x\right)$. Let $v \in V$ be the delta-function at the identity: $v(e)=1$ where $e$ is the identity in $G$ and $v(g)=0$ for all $g \neq e$. Then $f$ is a linear function on the orbit of $v$ and $\operatorname{dim} V=|G|$. Of course, in this case, the inequality $\|f\|_{\infty} \leq \sqrt{|G|} \cdot\|f\|_{2}$ is the best we can hope for as it is sharp on delta-functions.

To prove Corollary 2, we use the construction of the tensor power (see Section 1.2).

Proof of Corollary 2. Let us define a function $h: G \rightarrow \mathbb{R}$ by

$$
h(g)=f^{k}(g)=\ell^{\otimes k}\left(g v^{\otimes k}\right)
$$

Thus $h$ is a linear function on the orbit of $v^{\otimes k}$. Let

$$
W=\operatorname{span}\left\{g v^{\otimes k}: g \in G\right\}
$$

be the span of the orbit of $v^{\otimes k}$. Hence $\operatorname{dim} W=D_{k}$. Applying Theorem 1 to the linear function $h$ on the orbit of $v^{\otimes k}$ in $W$, we get

$$
\|h\|_{2} \leq\|h\|_{\infty} \leq \sqrt{D_{k}} \cdot\|h\|_{2}
$$

Now we note that $\|h\|_{\infty}=\|f\|_{\infty}^{k}$ and that $\|h\|_{2}=\|f\|_{2 k}^{k}$.
Remark. The bound $\|f\|_{\infty} \leq\left(D_{k}\right)^{1 / 2 k}\|f\|_{2 k}$ is rarely sharp. One example when it is sharp is provided by a generic orbit in the regular representation of a finite group $G$, see Section 2.1. In Section 3.1, we present a series of examples of matrix elements of $G=S O(n)$ for which the estimate is sharp up to a constant factor uniformly on $k$ and $n$.

Corollary 3 follows by a general estimate of $D_{k}$.
Proof of Corollary 3. We apply Corollary 2. The orbit $\left\{g v^{\otimes k}\right\}$ lies in the symmetric part $\operatorname{Sym}_{k}(V)$ of the tensor product $V^{\otimes k}$ and hence

$$
D_{k}=\operatorname{dim} \operatorname{span}\left\{g v^{\otimes k}\right\} \leq \operatorname{dim} \operatorname{Sym}_{k}(V)=\binom{\operatorname{dim} V+k-1}{k}
$$

### 2.2. Obtaining Sharp Estimates

As follows from Section 3.1, the upper bound for $\|f\|_{\infty}$ is sharp up to a constant factor uniformly on $k$ if $G=S O(n), V=\mathbb{R}^{n}$, and $G$ acts in $V$ by its defining representation.

We describe below classes of functions $f: G \rightarrow \mathbb{R}$ for which some sharp estimates can be obtained. Let us fix a representation $\rho$ of $G$ in a real vector space $W$, let $V=\operatorname{End}(W)$, and let $f$ be a linear function on the orbit of the identity operator $I \in V$, see Section 2.1. Then for some constant $C(\rho, k)$ we have $\|f\|_{\infty} \leq C(\rho, k)\|f\|_{2 k}$. Let us assume that $\rho$ is absolutely irreducible (see Section 2.1). In principle, the best possible value of $C(\rho, k)$ can be computed from the representation theory of $G$ as follows. Shifting $f$, if necessary, we may assume that the maximum absolute value of $f$ is attained at the identity $e$ of $G$. Let us define $h: G \rightarrow \mathbb{R}$ by

$$
h(x)=\int_{G} f\left(g^{-1} x g\right) d g \quad \text { for all } \quad x \in G
$$

Then $\|h\|_{\infty}=\|f\|_{\infty}$ and $\|h\|_{2 k} \leq\|f\|_{2 k}$ for all positive integers $k$. Thus the largest ratio $\|f\|_{\infty} /\|f\|_{2 k}$ is attained when $f$ satisfies $f\left(g^{-1} x g\right)=f(x)$ for all $g \in G$ and all $x \in G$, from which it follows that $f$ is a multiple of the character $\chi(g)=\operatorname{tr}(\rho(g))$, see Section 2.1. We observe that $\|\chi\|_{\infty}=\operatorname{dim} W$. Moreover, the orthogonality relations, see Fulton and Harris [6, Lecture 6], imply that $\|\chi\|_{2 k}^{2 k}$ is the sum of squares of multiplicities of the irreducible components of the tensor power $\rho^{\otimes k}$. Summarizing, we conclude that to compute the best possible constant $C(\rho, k)$ such that $\|f\|_{\infty} \leq C(\rho, k)\|f\|_{2 k}$, it suffices to know how the tensor power $\rho^{\otimes k}$ decomposes into the sum of absolutely irreducible representations. This example also shows that the algebra of tensor powers is indeed relevant for estimating $\|f\|_{\infty}$ in terms of $\|f\|_{2 k}$.

Finally, Corollary 4 follows by an estimate of the binomial coefficient.
Proof of Corollary 4. Let us choose a $k_{0}$ such that $(k!)^{1 / k}>2 \varepsilon^{-2}$ for all $k>k_{0}$. By Stirling's formula, we can choose $k_{0}=O\left(\varepsilon^{-2}\right)$. Then

$$
\begin{aligned}
& \binom{\operatorname{dim} V+k-1}{k}^{1 / 2 k}=\left(\frac{\operatorname{dim} V \cdot(\operatorname{dim} V+1) \cdots(\operatorname{dim} V+k-1)}{k!}\right)^{1 / 2 k} \\
& \leq\left(\frac{2^{k} \operatorname{dim}^{k} V}{k!}\right)^{1 / 2 k}=2^{1 / 2}(k!)^{-1 / 2 k} \cdot \sqrt{\operatorname{dim} V} \leq \varepsilon \sqrt{\operatorname{dim} V}
\end{aligned}
$$

The proof follows by Corollary 3.

## 3. Applications to Polynomials

In this section, we apply our results to approximate the maximum absolute value of a homogeneous multivariate polynomial on the unit sphere.

Let $p$ be a homogeneous polynomial of degree $d$ in $n$ real variables $\xi_{1}, \ldots, \xi_{n}$. Thus we can write

$$
p(x)=\sum_{1 \leq i_{1}, \ldots, i_{d} \leq n} \gamma_{i_{1} \cdots i_{d}} \xi_{i_{1}} \cdots \xi_{i_{d}} \quad \text { for } \quad x=\left(\xi_{1}, \ldots, \xi_{n}\right)
$$

where $\gamma_{i_{1} \cdots i_{d}}$ are some real numbers.
Let $\mathbb{R}^{n}$ be the $n$-dimensional Euclidean space and let $x=\left(\xi_{1}, \ldots, \xi_{n}\right) \in \mathbb{R}^{n}$ be a point. Then $V=\left(\mathbb{R}^{n}\right)^{\otimes d}$ can be identified with the space $\mathbb{R}^{n^{d}}$. The coordinates of a typical point (tensor) $X \in V$ are

$$
\left(X_{i_{1} \cdots i_{d}}: 1 \leq i_{1}, \ldots, i_{d} \leq n\right)
$$

and the scalar product in $V$ is defined by

$$
\langle X, Y\rangle=\sum_{1 \leq i_{1}, \ldots, i_{d} \leq n} X_{i_{1} \cdots i_{d}} Y_{i_{1} \cdots i_{d}}
$$

For $x=\left(\xi_{1}, \ldots, \xi_{n}\right) \in \mathbb{R}^{n}$, the coordinates of $x^{\otimes d}$ are

$$
\left(\xi_{i_{1}} \cdots \xi_{i_{d}} \text { for } 1 \leq i_{1}, \ldots, i_{d} \leq n\right)
$$

Therefore, we can write

$$
p(x)=\left\langle c, x^{\otimes d}\right\rangle \quad \text { where } \quad c=\left(\gamma_{i_{1} \cdots i_{d}}\right) .
$$

Let $G=S O(n)$ be the group of orientation-preserving orthogonal transformations of $\mathbb{R}^{n}$. Then $G$ acts in $V$ by the $d$ th tensor power of its defining representation in $\mathbb{R}^{n}$. Let us choose $w=(1,0, \ldots, 0) \in \mathbb{R}^{n}$. Then, for any $g \in G$, we have

$$
\left\langle c, g w^{\otimes d}\right\rangle=p(g w)
$$

and the orbit $\{g w: g \in G\}$ is the unit sphere $\mathbb{S}^{n-1} \subset \mathbb{R}^{n}$. Thus the values of $p(x)$, as $x$ ranges over the unit sphere in $\mathbb{R}^{n}$, are the values of the linear function

$$
\ell\left(g w^{\otimes d}\right)=\left\langle c, g w^{\otimes d}\right\rangle=\left\langle c, x^{\otimes d}\right\rangle
$$

as $g$ ranges over the orthogonal group $S O(n)$.
Moreover, the push-forward of the Haar probability measure $d g$ on $G$ is the probability measure $d x$ on $\mathbb{S}^{n-1}$. Thus we connect the values of a polynomial on the unit sphere with the values of a linear function on the orbit of the group $G=S O(n)$.

Corollary 7. Let p be a homogeneous polynomial of degree d in $n$ real variables, let $\mathbb{S}^{n-1}$ be the unit sphere in $\mathbb{R}^{n}$, and let dx be the rotation invariant probability measure on $\mathbb{S}^{n-1}$. For a positive integer $k$, let us define the $L^{2 k}$ norm of $p$ by

$$
\|p\|_{2 k}=\left(\int_{\mathbb{S}^{n-1}} p^{2 k}(x) d x\right)^{1 / 2 k}
$$

and let us define the $L^{\infty}$ norm by

$$
\|p\|_{\infty}=\max _{x \in \mathbb{S}^{n-1}}|p(x)|
$$

Then

$$
\|p\|_{2 k} \leq\|p\|_{\infty} \leq\binom{ k d+n-1}{k d}^{1 / 2 k}\|p\|_{2 k}
$$

Proof. We apply Corollary 2. Let $w=(1,0, \ldots, 0)$ be as above. Then, for $v=w^{\otimes d}$, we can write $p(g w)=\ell(g v)$ for some linear function $\ell: V \rightarrow \mathbb{R}$ and all $g \in G$. The dimension $D_{k}$ of the span of the orbit $\left\{g v^{\otimes k}=g w^{\otimes k d}: g \in G\right\}$ is that of the space of homogeneous polynomials of degree $k d$ in $n$ variables. Hence

$$
D_{k}=\binom{k d+n-1}{k d}
$$

We have

$$
\int_{G} \ell^{2 k}(g v) d g=\int_{\mathbb{S}^{n-1}} p^{2 k}(x) d x
$$

The proof now follows.
One way to integrate polynomials over the unit sphere is to take the sum of the integrals of the monomials. The following result is certainly known, but for the sake of completeness, we sketch its proof here.

Lemma 8. Let $p(x)=\xi_{1}^{\alpha_{1}} \cdots \xi_{n}^{\alpha_{n}}$ be a monomial. If at least one of the $\alpha_{i}$ 's is odd, then

$$
\int_{\mathbb{S}^{n-1}} p(x) d x=0 .
$$

If $\alpha_{i}=2 \beta_{i}$, where $\beta_{i}$ are nonnegative integers for $i=1, \ldots, n$, then

$$
\int_{\mathbb{S}^{n}-1} p(x) d x=\frac{\Gamma(n / 2) \prod_{i=1}^{n} \Gamma\left(\beta_{i}+\frac{1}{2}\right)}{\pi^{n / 2} \Gamma\left(\beta_{1}+\cdots+\beta_{n}+n / 2\right)}
$$

where $d x$ is the Haar probability measure on $\mathbb{S}^{n-1}$.
Sketch of Proof. If $\alpha_{i}$ is odd, then

$$
p\left(\xi_{1}, \ldots, \xi_{i-1},-\xi_{i}, \xi_{i+1}, \ldots, \xi_{n}\right)=-p\left(\xi_{1}, \ldots, \xi_{i-1}, \xi_{i}, \xi_{i+1}, \ldots, \xi_{n}\right)
$$

and hence the average value of $p$ over the unit sphere is 0 .
Assuming that $\alpha_{i}=2 \beta_{i}$ for $i=1, \ldots, n$, we get

$$
\int_{\mathbb{R}^{n}} p(x) e^{-\|x\|^{2}} d \mu=\prod_{i=1}^{n} \int_{\mathbb{R}} \xi^{2 \beta_{i}} e^{-\xi^{2}} d \xi=\prod_{i=1}^{n} \Gamma\left(\beta_{1}+\frac{1}{2}\right),
$$

where $\mu$ is the standard Lebesgue measure in $\mathbb{R}^{n}$.

On the other hand, passing to the polar coordinates and using that $p$ is homogeneous of degree $d=2\left(\beta_{1},+\cdots+\beta_{n}\right)$, we get

$$
\int_{\mathbb{R}^{n}} p(x) e^{-\|x\|^{2}} d \mu=\left|\mathbb{S}^{n-1}\right| \cdot\left(\int_{\mathbb{S}^{n-1}} p(x) d x\right) \cdot\left(\int_{0}^{+\infty} r^{d+n-1} e^{-r^{2}} d r\right)
$$

where $\left|\mathbb{S}^{n-1}\right|=2 \pi^{n / 2} / \Gamma(n / 2)$ is the Euclidean volume of the unit sphere in $\mathbb{R}^{n}$. The proof now follows.

The estimates of Corollary 7 are probably not optimal (apart from the case of $k=1$ ), but the following simple example shows that, in some sense, they are close to being optimal.

### 3.1. Powers of Linear Functions

Let $p$ be the power of a linear function, for example, $p(x)=\xi_{1}^{d}$. Then $\|p\|_{\infty}=1$ and, by Lemma 8,

$$
\|p\|_{2 k}=\left(\frac{\Gamma(n / 2) \Gamma(k d+1 / 2)}{\sqrt{\pi} \Gamma(k d+n / 2)}\right)^{1 / 2 k}
$$

Then Corollary 7 gives us the estimate

$$
\begin{aligned}
\|p\|_{\infty} & \leq\left(\frac{\Gamma(n / 2) \Gamma(k d+1 / 2) \Gamma(k d+n)}{\sqrt{\pi} \Gamma(k d+n / 2) \Gamma(n) \Gamma(k d+1)}\right)^{1 / 2 k} \leq\left(\frac{\Gamma(n / 2) \Gamma(k d+n)}{\Gamma(k d+n / 2) \Gamma(n)}\right)^{1 / 2 k} \\
& =\left(\frac{n(n+1) \cdots(k d+n-1)}{(n / 2)(n / 2+1) \cdots(k d+n / 2-1)}\right)^{1 / 2 k} \leq 2^{d / 2}
\end{aligned}
$$

Hence, among all homogeneous polynomials of a given degree $d$, powers of linear functions give the largest ratio $\|f\|_{\infty} /\|f\|_{2 k}$ up to a constant factor depending on the degree of $f$ and independent of the number of variables $n$ and the value of $k$. In particular, if $d=1$, the upper bound of Corollary 3 exceeds the tight bound by at most a factor of $\sqrt{2}$.

This may serve as an indication that the bounds of Corollary 3 are not too bad, see Sections 2.1 and 2.2. G. Blekherman [2] pointed out to the author that the powers, in general, do not provide exactly the largest ratio $\|f\|_{\infty} /\|f\|_{2 k}$ among all polynomials of a given degree $d$. Such "extremal" polynomials $f$ were computed by G. Blekherman when some of the parameters $n, d$, and $k$ are small.

Suppose we want to approximate $\|p\|_{\infty}$ by $\|p\|_{2 k}$ for a sufficiently large $k$. Let us see what trade-off between the computational complexity and accuracy we can achieve.

### 3.2. Low Degree Polynomials

Let us fix the degree $d$ and allow the number $n$ of variables to vary. Suppose that we are given a homogeneous polynomial $p$ of degree $d$ and that we want to estimate $\|p\|_{\infty}$. This problem is NP-hard already for $d=4$ (one can infer it from the results of Blum et al. [3, Part 1] and is suspected to be hard for $d=3$.

Let $m$ be the number of monomials in $p$, so $m=O\left(n^{d}\right)$. We observe that for any fixed $k$, the direct computation of $p^{2 k}(x)$ and computing $\|p\|_{2 k}$ via Lemma 8 has $O\left(m^{2 k}\right)$ complexity. One the other hand, using Corollary 7, we get that

$$
\|p\|_{\infty} \leq C(k) n^{d / 2}\|p\|_{2 k} \quad \text { where } \quad C(k)=O\left(k^{-1 / 2}\right)
$$

In other words, for any fixed $\varepsilon>0$, there is a polynomial time algorithm estimating $\|p\|_{\infty}$ within a factor of $\varepsilon n^{d / 2}$. If we want a better estimate, we have to take a larger $k$. Thus, for any constant $C>1$, from Corollary 7 we get that

$$
\|p\|_{\infty} \leq C\|p\|_{2 k} \quad \text { for some } \quad k=O(n)
$$

Since $p^{2 k}(x)$ contains at most $\binom{2 k d+n-1}{2 k d}$ monomials, we can compute $\|p\|_{2 k}$ by Lemma 8 in $2^{O(n)}$ time. Summarizing, for any $C>1$ there exists a $\gamma>0$ such that we can approximate $\|p\|_{\infty}$ within a factor $C$ in $2^{\gamma n}$ time.

### 3.3. Fewnomials and Their Extensions

Suppose that we do not fix the degree $d$ of $p$ but fix instead the number $m$ of monomials in $p$. Thus we can write

$$
p(x)=\sum_{i=1}^{m} p_{i}(x)
$$

where

$$
p_{i}(x)=\gamma_{i} \xi_{1}^{\alpha_{i 1}} \cdots \xi_{n}^{\alpha_{i n}}
$$

are monomials. For a positive integer $k$, by the multinomial expansion, we get

$$
p^{2 k}=\sum_{\substack{r_{1}, \ldots, r_{m} \geq 0 \\ r_{1}+\cdots+r_{m}=2 k}} \frac{(2 k)!}{r_{1}!\cdots r_{m}!} p_{1}^{r_{1}} \cdots p_{m}^{r_{m}}
$$

Thus $p^{2 k}$ contains at most $\binom{m+2 k-1}{m-1}$ monomials, which is a polynomial in $k$ when $m$ is fixed. Using Lemma 8 , we compute $\|p\|_{2 k}$ in $O\left(d n(2 k)^{m}\right)$ time. Given an $\varepsilon>0$, for some $\delta>0$ let us choose an integer $k=O\left(\varepsilon^{-1} n^{1+\delta} \ln d\right)$ such that

$$
\frac{n-1}{2 k} \ln (k d+1)<\ln (1+\varepsilon)
$$

Using Corollary 7 and a simple estimate

$$
\binom{k d+n-1}{n-1}=\frac{(k d+1)(k d+2) \cdots(k d+n-1)}{1 \cdot 2 \cdots(n-1)} \leq(k d+1)^{n-1}
$$

we conclude that

$$
\|p\|_{2 k} \leq\|p\|_{\infty} \leq(1+\varepsilon) \cdot\|p\|_{2 k}
$$

Hence as long as the number of monomials is fixed, we get a polynomial time approximation algorithm, which, for any given $\varepsilon>0$, computes the maximum absolute value of a given polynomial ("fewnomial") over the unit sphere within a relative error of $\varepsilon$, in time polynomial in $\varepsilon^{-1}$, the number of variables $n$ and the degree $d$ of the polynomial. In fact, the only place where we have to use polynomially many in $d$ arithmetic operations is when we compute gamma-functions (factorials) in Lemma 8. Apart from this, the running time of the algorithm is polynomial in $\ln d$.

One can define a wider class of polynomials $p$ for which a similar polynomial time approximation scheme exists. For a monomial $x^{a}=\xi_{1}^{\alpha_{1}}, \ldots, \xi_{n}^{\alpha_{n}}$, let $a=\left(\alpha_{1}, \ldots, \alpha_{n}\right) \in \mathbb{Z}^{n}$ be its multidegree considered as a point in $\mathbb{R}^{n}$. For a polynomial $p$, let us define the subspace $\mathcal{L}_{p} \subset \mathbb{R}^{n}$ to be the span of all $a$ such that the monomial $x^{a}$ has a nonzero coefficient in $p$. In other words, $\mathcal{L}_{p}$ is the span of the Newton polytope of $p$, see Khovanskii [7]. Let us fix a positive integer $m$ and consider the class of homogeneous polynomials $p$ such that $\operatorname{dim} \mathcal{L}_{p} \leq m$. We note that for any $k$, the multidegrees of monomials in $p^{k}$ are points from $\mathcal{L}_{p} \cap \mathbb{Z}^{n}$ contained in the cube $0 \leq \alpha_{i} \leq k d$ with $d=\operatorname{deg} p$. In particular, the number of monomials in $p^{k}$ grows polynomially in $k$. Using that, one can compute the monomial expansion of $p^{k}$ recursively in time polynomials in $k, d$, and $n$. Hence, for any $\varepsilon>0$, one can estimate $\|p\|_{\infty}$ within relative error $\varepsilon$ in time polynomials in $\varepsilon^{-1}, n$, and $d$.

Computing or approximating the maximum absolute value of a polynomial on the unit sphere can be used for testing whether a given system of real polynomial equations has a real solution, a difficult (NP-hard) and important problem, see e.g., Blum et al. [3] and Renegar [9]. Suppose that $p_{i}: i=1, \ldots, s$, are given homogeneous polynomials of degree $d$ in $n$ variables $x=\left(\xi_{1}, \ldots, \xi_{n}\right)$ and that we would like to test whether the system

$$
p_{i}(x)=0 \quad \text { for } \quad i=1, \ldots, s
$$

has a real solution $x \neq 0$. Let

$$
q=\sum_{i=1}^{s} p_{i}^{2}(x) .
$$

Thus we want to test whether

$$
\min _{x \in \mathbb{S}^{n-1}} q(x)=0
$$

Let us choose a

$$
\gamma>\max _{x \in \mathbb{S}^{n-1}} q(x)
$$

and let

$$
p=\gamma\|x\|^{2 d}-q
$$

Thus the problem reduces to checking whether

$$
\max _{x \in \mathbb{S}^{n-1}}|p(x)|=\gamma
$$

If the polynomials $p_{i}$ of the original system do not have too many monomials or belong to other easy cases described above, we can try to approximate $\|p\|_{\infty}$ by $\|p\|_{2 k}$ for a reasonably large $k$, see Section 3.3. Similarly, to choose an appropriate $\gamma$, we can compute $\|q\|_{2 k}$ for a sufficiently large $k$. The number of monomials in the system is relevant to the "topological complexity" of the set of real solutions (Khovanskii [7]), so it should not be surprising that it is also relevant to the computational complexity of the decision problem. In particular, this approach may be useful for detecting "badly unsolvable" systems (systems for which the value of $\|p\|_{\infty}$ is substantially smaller than $\gamma$ ) of fewnomial equations.

## 4. Applications to Combinatorial Optimization

Let us fix a number $d$ and let $V=\left(\mathbb{R}^{n}\right)^{\otimes d}=\mathbb{R}^{n^{d}}$ be the vector space of $d$ dimensional arrays (tensors)

$$
X=\left(x_{i_{1} \cdots i_{d}}: 1 \leq i_{1}, \ldots, i_{d} \leq n\right)
$$

To simplify the notation somewhat, we denote the coordinates of $X$ by $x_{I}$, where $I=\left(i_{1}, \ldots, i_{d}\right)$.

We introduce the scalar product by

$$
\langle X, Y\rangle=\sum_{I} x_{I} y_{I} \quad \text { for } \quad I=\left(1 \leq i_{1}, \ldots, i_{d} \leq n\right)
$$

Let $G=S_{n}$ be the symmetric group of all permutations $g$ of the set $\{1, \ldots, n\}$. We introduce the action of $S_{n}$ on $V$ by the $d$ th tensor power of the natural action of $S_{n}$ in $\mathbb{R}^{n}$ :

$$
Y=g X \quad \text { provided } \quad x_{I}=y_{g I} \quad \text { where } \quad g\left(i_{1}, \ldots, i_{d}\right)=\left(g\left(i_{1}\right), \ldots, g\left(i_{d}\right)\right)
$$

Let us choose two tensors $A, B \in V$ and let

$$
f(g)=\langle B, g A\rangle=\sum_{1 \leq i_{1}, \ldots, i_{d} \leq n} a_{i_{1} \cdots i_{d}} b_{g\left(i_{1}\right) \cdots g\left(i_{d}\right)}, \quad f: S_{n} \rightarrow \mathbb{R}
$$

be the corresponding matrix element.

We call the problem of optimizing $f$ the assignment problem of degree $d$. It is straightforward for $d=1$. For $d=2$, it is known as the quadratic assignment problem, one of the most general problems of combinatorial optimization, see, e.g., Çela [4]. The quadratic assignment problem is NP-hard and viewed as one of the "hardest" among NP-hard problems, compared, for example, with the traveling salesman problem.

### 4.1. An Application: Matching Edges of Hypergraphs

Recall that a $d$-hypergraph $H$ on the set $\{1, \ldots, n\}$ is a set of subsets $E \subset$ $\{1, \ldots, n\}$, called edges of $H$, such that $|E| \leq d$ for the cardinality $|E|$ of every edge $E$ of $H$. A hypergraph is called uniform provided $|E|=d$ for every edge $E$ of $H$. Let $H_{1}$ and $H_{2}$ be uniform $d$-hypergraphs with the set of vertices $\{1, \ldots, n\}$. Let us define the adjacency tensor $A=\left(a_{i_{1} \cdots i_{d}}\right)$ of $H_{1}$ by

$$
a_{i_{1} \cdots i_{d}}= \begin{cases}1 & \text { if }\left\{i_{1}, \ldots, i_{d}\right\} \text { is an edge of } H_{1} \\ 0 & \text { otherwise }\end{cases}
$$

Let us define $B=\left(b_{i_{1} \cdots i_{d}}\right)$ by

$$
b_{i_{1} \cdots i_{d}}= \begin{cases}1 / d! & \text { if }\left\{i_{1}, \ldots, i_{d}\right\} \text { is an edge of } H_{2} \\ 0 & \text { otherwise }\end{cases}
$$

A permutation $g$ of the set $\{1, \ldots, n\}$ is interpreted as a bijection between the vertices of $H_{2}$ and the vertices of $H_{1}$, and the value of

$$
f(g)=\langle B, g A\rangle
$$

is the number of edges of $H_{2}$ mapped onto the edges of $H_{1}$. The value of $\|f\|_{\infty}$ is the maximum number of edges of $H_{1}$ and $H_{2}$ that can be matched by a bijection of the vertices of $H_{1}$ and $H_{2}$. If $H_{1}$ and $H_{2}$ are not uniform, we can modify $B$ by letting

$$
b_{i_{1}, \ldots, i_{d}}=\frac{k_{1}!\cdots k_{r}!}{d!}
$$

provided $\left\{i_{1}, \ldots, i_{d}\right\}$ is an edge of $H_{2}$ and the multiplicities of the elements in the multiset $\left\{\left\{i_{1}, \ldots, i_{d}\right\}\right\}$ are $k_{1}, \ldots, k_{r}$, so that $k_{1}+\cdots+k_{r}=d$. Then again the value of $\|f\|_{\infty}$ is equal to the maximum number of edges of $H_{1}$ and $H_{2}$ that can be matched by a bijection of the vertex sets.

One can extend this construction to oriented hypergraphs whose edges are ordered subsets of $\{1, \ldots, n\}$. By introducing weights on the edges of $H_{1}$ and $H_{2}$ we can introduce "prices" for matching (or mismatching) particular edges.

Applying Corollary 3, we get the inequality

$$
\|f\|_{2 k} \leq\|f\|_{\infty} \leq\binom{ n^{d}+k-1}{k}^{1 / 2 k}\|f\|_{2 k}
$$

for the function $f$ of a general assignment problem of degree $d$.
In various special cases, the bound can be somewhat improved by using Corollary 2. For example, if the coordinates of $A$ (or $B$ ) are 0 's and 1's, one can prove that

$$
\|f\|_{2 k} \leq\|f\|_{\infty} \leq D^{1 / 2 k}(n, d, k) \cdot\|f\|_{2 k} \quad \text { where } \quad D(n, d, k)=\sum_{j=1}^{k}\binom{n^{d}}{j}
$$

We claim that for small (fixed) values of $k$ the value of $\|f\|_{2 k}$ can be computed relatively easily (in polynomial time). First, we observe that computation of $\|f\|_{2 k}$ reduces to computation of the average of a matrix element for larger tensors.

Lemma 9. Let us fix two tensors $A=\left(a_{I}\right)$ and $B=\left(b_{I}\right)$ for $I=(1 \leq$ $i_{1}, \ldots, i_{d} \leq n$ ). For a positive integer $m$ (in particular, for $m=2 k$ ), let us define tensors $X=A^{\otimes m}$ and $Y=B^{\otimes m}$ as follows:

$$
X=\left(x_{J}\right) \quad \text { and } \quad Y=\left(y_{J}\right) \quad \text { where } \quad J=\left(1 \leq j_{1}, \ldots, j_{d m} \leq n\right)
$$

and where

$$
x_{J}=a_{I_{1}} \cdots a_{I_{m}} \quad \text { and } \quad y_{J}=b_{I_{1}} \cdots b_{I_{m}} \quad \text { provided } \quad J=\left(I_{1}, \ldots, I_{m}\right) .
$$

Then

$$
\frac{1}{n!} \sum_{g \in S_{n}}\langle B, g A\rangle^{m}=\frac{1}{n!} \sum_{g \in S_{n}}\langle Y, g X\rangle
$$

Proof. The proof follows by observation that

$$
\langle B, g A\rangle^{m}=\left\langle B^{\otimes m}, g A^{\otimes m}\right\rangle=\langle Y, g X\rangle .
$$

Next, we show how to compute the average.

Lemma 10. Let us fix a positive integer $l$ (in particular, $l=m d=2 k d$ ). For a partition $\Sigma=\left\{\Sigma_{1}, \ldots, \Sigma_{r}\right\}$ of the set $\{1, \ldots, l\}$ into nonempty disjoint subsets, we say that a sequence $I=\left(i_{1}, \ldots, i_{l}\right)$ has type $\Sigma$ if for each $\Sigma_{p}$ the indices $i_{j}: j \in \Sigma_{p}$ are all equal and if for each pair of subsets $\Sigma_{p}$ and $\Sigma_{q}$ the indices $i_{j}: j \in \Sigma_{p}$ and $i_{j}: j \in \Sigma_{q}$ are different.

Let $X=\left(x_{I}\right)$ and $Y=\left(y_{I}\right), I=\left(1 \leq i_{1}, \ldots, i_{l} \leq n\right)$ be tensors (in particular, we can have $X=A^{\otimes m}=A^{\otimes 2 k}$ and $\left.Y=B^{\otimes m}=B^{\bar{\otimes} 2 k}\right)$.

Let us define the tensors $\bar{X}=\left(\bar{x}_{I}\right)$ and $\bar{Y}=\left(\bar{y}_{I}\right), I=\left(1 \leq i_{1}, \ldots, i_{l} \leq n\right)$ by

$$
\bar{x}_{I}=\frac{(n-r)!}{n!} \sum_{J: \text { type } J=\text { type } I} x_{J} \quad \text { provided type } I=\left(\Sigma_{1}, \ldots, \Sigma_{r}\right)
$$

and

$$
\bar{y}_{I}=\frac{(n-r)!}{n!} \sum_{J: \text { type } J=\text { type } I} y_{J} \quad \text { provided type } I=\left(\Sigma_{1}, \ldots, \Sigma_{r}\right)
$$

Then

$$
\frac{1}{n!} \sum_{g \in S_{n}}\langle Y, g X\rangle=\langle\bar{Y}, \bar{X}\rangle
$$

Proof. The two index sets $I=\left(i_{1}, \ldots, i_{l}\right)$ and $J=\left(j_{1}, \ldots, j_{l}\right)$ belong to the same orbit of the action $I \mapsto g I$ of $S_{n}$ if and only if they have the same type $\left\{\Sigma_{1}, \ldots, \Sigma_{r}\right\}$. Moreover, the stabilizer of $I$ consists of $(n-r)$ ! permutations. Hence

$$
\bar{X}=\frac{1}{n!} \sum_{g \in S_{n}} g X \quad \text { and } \quad \bar{Y}=\frac{1}{n!} \sum_{g \in S_{n}} g Y
$$

We have

$$
\frac{1}{n!} \sum_{g \in S_{n}}\langle Y, g X\rangle=\left\langle\frac{1}{n!} \sum_{g \in S_{n}} g Y, \frac{1}{n!} \sum_{g \in S_{n}} g X\right\rangle
$$

and the proof follows.

Combining Lemmas 9 and 10, we observe that as long as $d$ and $k$ are fixed, we can compute $\|f\|_{2 k}$ in $O\left(n^{2 k d}\right)$ time, that is, in polynomial in $n$ time.

In particular, from Corollary 4, we conclude that for any fixed $d$ and for any fixed $\varepsilon>0$ there exists a polynomial in $n$ algorithm for estimating $\|f\|_{\infty}$ within a factor of $\varepsilon n^{d / 2}$. This result seems to be new already for $d=2$, see Arkin et al. [1].

### 4.2. Finding a Permutation

So far we have shown how to approximate $\|f\|_{\infty}$ by $\|f\|_{2 k}$ but we did not discuss how to find a particular permutation $g$ which gives the value of $|f(g)|$ close to $\|f\|_{\infty}$. In fact, it is not hard to construct a permutation $g \in S_{n}$ for which $|f(g)| \geq\|f\|_{2 k}$ and, hence, $|f(g)|$ approximates $\|f\|_{\infty}$ within a factor of $\varepsilon n^{d / 2}$ at the cost of some extra work, which still results in a polynomial time algorithm when $k$ is fixed. The idea is to use the "method of conditional probabilities," see, e.g., Motwani and Raghavan [8, Section 5.6]. We split the symmetric group $S_{n}$ into the union of cosets $S_{j}=\{g: g(1)=j\}$ and then compute the average value of $f^{2 k}$
over each coset separately (this would require some straightforward modification of Lemma 10). Then a coset should be chosen which gives the largest average. Thus we have determined $g(1)=j$ and we proceed to determine $g(2), \ldots, g(n)$ successively.

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