Five Point Energy Minimization: A Summary

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August 14, 2019

Abstract

This paper is a condensation of my monograph [S0], which contains a complete proof that there is a constant $\psi \approx 15.0488$ such that the triangular bi-pyramid is the minimizer, amongst all 5 point configurations on the sphere, with respect to the power law potential $R_s(r) = \text{sign}(s)/r^s$, if and only if $s \in (-2, 0) \cup (0, \psi]$. In this paper we explain the main ideas and give proofs for some of the key lemmas.

1 Introduction

Let $S^2$ denote the unit sphere in $\mathbb{R}^3$ and let $X = \{\hat{p}_0, \ldots, \hat{p}_{n-1}\}$ be a finite list of distinct points on $S^2$. Given some function $f : (0, 2] \to \mathbb{R}$ we can compute the total $f$-potential

$$E_f(X) = \sum_{i<j} f(\|\hat{p}_i - \hat{p}_j\|).$$

For fixed $f$ and $n$, one can ask which configuration(s) $X$ minimize $E_f(X)$. This problem generally goes under the heading of Thomson's Problem, and some form of it originates in [Th].

For this problem, the energy functional $f = R_s$, where

$$R_s(r) = \text{sign}(s)r^{-s},$$

is a natural one to consider. When $s > 0$, this is called the Riesz potential. When $s < 0$ this is called the Fejes-Toth potential. The case $s = 1$ is specially called the Coulomb potential or the electrostatic potential.
There is a large literature on the energy minimization problem. See [Fö] and [C] for some early local results. See [MKS] for a definitive numerical study on the minimizers of the Riesz potential for \( n \leq 16 \). See [SK] and [RSZ] for results with an emphasis on the case when \( n \) is large. See [BBCGKS] for a survey of results about the higher dimensional case. For \( n = 4, 6, 12 \) it is known that the set of vertices of the corresponding platonic solid is the global minimizer with respect to any Riesz potential. See [KY], [A], [Y]. See also [CK] for a vast generalization.

The case \( n = 5 \) has been notoriously intractable. There is a general feeling that for a wide range of energy choices including the Riesz potentials, the global minimizer is either the triangular bi-pyramid \(^1\) (TBP) or else some pyramid with square base – an FP. Here is a resume of results.

- The paper [HS] has a rigorous computer-assisted proof that the TBP is the unique minimizer for \( R_{-1} \).
- My paper [S1] has a rigorous computer-assisted proof that the TBP is the unique minimizer for \( R_1 \) and \( R_2 \).
- The paper [DLT] gives a traditional proof that the TBP is the unique minimizer for the logarithmic potential.
- In [BHS, Theorem 7] it is shown that, as \( s \to \infty \), any sequence of 5-point minimizers w.r.t. \( R_s \) must converge (up to rotations) to the FP made from 5 vertices of the regular octahedron. So, the TBP is not a minimizer w.r.t \( R_s \) when \( s \) is very large.
- In 1977, T. W. Melnyk, O. Knop, and W. R. Smith, [MKS] conjectured the existence of the phase transition constant, around \( s = 15.04808 \), at which point the TBP ceases to be the minimizer w.r.t. \( R_s \).
- In [T], A. Tumanov gives a traditional proof of the following result. Define

\[
G_k(r) = (4 - r^2)^k, \quad k = 1, 2, 3, ... \tag{3}
\]

Let \( f = a_1G_1 + a_2G_2 \) with \( a_1, a_2 > 0 \). The TBP is the unique global minimizer with respect to \( f \). Moreover, a critical point of \( f \) must be the TBP. In particular, the TBP is a minimizer for \( G_2 \).

\(^1\)This is the configuration isometric to the one consisting of the north pole, the south pole, and three points placed on the equator in an equilateral triangle.
In my unpublished monograph [S0], I prove the following result.

**Theorem 1.1 (Main)** There exists a computable number
\[ \varpi = 15.0480773927797... < 15 + 25/512 \]
with the following properties:

- For \( s \in (-2, 0) \cup (0, \varpi) \) the TBP is the unique minimizer w.r.t. \( R_s \).
- For \( s = \varpi \), the TBP and some FP are the two minimizers w.r.t. \( R_s \).
- For \( s \in (\varpi, 15 + 25/512] \), some FP is the unique minimizer w.r.t \( R_s \).
- For \( s \geq 15 + 25/512 \) the TBP is not the global minimizer w.r.t. \( R_s \).

It is worth remarking that the fourth item is pretty easy to prove with elementary methods, because it just involves a direct comparison between the TBP and FPs.

The purpose of this paper is to give a condensed account of the work in [S0], concentrating on the overall strategy and providing proofs of, or at least insight into, the key technical lemmas. §2 of this paper gives an outline of the proof and then subsequent chapters fill in the outline.

I would also like to mention that my Java Program [S3], which the reader can download, is a companion to the monograph. This program illustrates most of the ideas discussed here.

I would like to thank Ed Saff for suggesting that I write this condensation. I would also like to thank two anonymous referees for some helpful suggestions.

## 2 Outline of the Proof

### 2.1 Interpolation

We consider two configurations the same if they are isometric, and as we go on we will normalize our configurations in particular ways. Given the function \( f \) and a 5-point configuration \( X \) we often write \( f(X) \) in place of \( \mathcal{E}_f(X) \) for ease of notation. Let \( R_s \) be as in the introduction.

Let \( T \) denote the triangular bi-pyramid. Let \( I \subset \mathbb{R} \) denote an interval, which we think of as an interval of power law exponents. We say that a
triple \((\Gamma_2, \Gamma_3, \Gamma_4)\) of potentials is forcing on the interval \(I\) if the following implication holds for any 5-point configuration \(X\):

\[
\Gamma_i(X) > \Gamma_i(T) \quad i = 2, 3, 4 \quad \implies \quad R_s(X) > R_s(T) \quad \forall s \in I. \tag{4}
\]

Recall that \(G_k(r) = (4 - r^2)^k\). In the paper \([T]\), A. Tumanov observes that \((G_2, G_3, G_5)\) is forcing on \((-2, 0)\) and \((0, 2)\). He does not supply a proof for his observation, but we will take up a related idea. Consider the following functions:

\[
\begin{align*}
G_5^\circ &= G_5 - 25G_1 \\
G_{10}^{\#} &= G_{10} + 13G_5 + 68G_2 \\
G_{10}^{##} &= G_{10} + 28G_5 + 102G_2
\end{align*}
\]

(5)

In the §3 we will sketch the ideas behind a proof of the following result.

**Lemma 2.1 (Forcing)** The following is true.

1. \((G_2, G_3, G_5)\) is forcing on \((-2, 0)\).
2. \((G_2, G_4, G_6)\) is forcing on \((0, 6)\).
3. \((G_2, G_5, G_{10}^{##})\) is forcing on \([6, 13]\).
4. \((G_2, G_5^\circ, G_{10}^{##})\) is forcing on \([13, 15.05]\).

**Remarks:**

(1) Item 4 in the Forcing Lemma plays a different role in the proof of the Main Theorem because \(T\) is not the global minimizer for \(G_{10}^{##}\). While Items 1-3 play a role in the proof of the Big Theorem, Item 4 plays a role in the proof of the Small Theorem.

(2) We don’t bother proving that \((G_2, G_3, G_5)\) is forcing on \((0, 2)\) because we don’t need this fact.

(3) Ideas like the Forcing Lemma are used in many papers on energy minimization – see e.g. \([CK]\), \([BDHSS]\), and (for my inspiration) \([T]\). I believe that the idea goes back to Yudin \([Y]\).

(4) I found these triples through computer experimentation, taking Tumanov’s observation as a starting point. The reader can do experiments themselves, using my Java program \([S3]\). One other guide to the experiments is that \(T\) is not the minimizer for \(G_k\) when \(k = 7, 8, 9, \ldots\), so one might look at more complicated combinations of these functions.
2.2 Divide and Conquer

As we explain in §4, we use a divide and conquer algorithm, together with a certain energy estimate, to give a rigorous computer-assisted proof of

**Theorem 2.2 (Big)** The TBP is the unique global minimizer w.r.t.

\[ G_3, G_4, G_5^b, G_6, G_{10}^{##} \]

We prove the energy estimate in §5.

We also recall that Tumanov’s result implies that the TBP is a minimizer for both \( G_1 \) and \( G_2 \). We note that \( G_5 \) is a positive combination of \( G_1 \) and \( G_5^b \), so \( G_5 \) also satisfies the conclusion of the Big Theorem. The Big Theorem combines with Tumanov’s result and the Forcing Lemma to prove the Main Theorem for all exponents \( s \in (-2, 0) \cup (0, 13] \). To finish the proof, we just have to deal with the exponent interval \((13, 15 + 25/512] \). The TBP is not the global minimizer for \( G_{10}^{##} \), but we can still squeeze information out of this function.

The first idea of our proof is to transfer the main problem from the sphere to the plane. For this purpose, there are a variety of coordinate systems we could use. We will use steeographic projection because this map has several virtues. First of all, it is a birational map: It is given by rational functions and so is its inverse. This is very useful when we want to do computer calculations with rational quantities. Were we to use spherical coordinates, for instance, this would not work as well. Second of all, stereographic projection maps generalized circles – i.e. circles or lines – to generalized circles. Our various energy estimates are made easier by the fact that, in the guts of the proofs, we work with lines and circles.

Stereographic projection is the map

\[
\Sigma(x, y, z) = \left( \frac{x}{1-z}, \frac{y}{1-z} \right). \tag{6}
\]

The domain of this map is \( S^2 - \{(0, 0, 1)\} \). Here \((0, 0, 1)\) is the north pole. We rotate our configurations \( \hat{p}_0, \hat{p}_1, \hat{p}_2, \hat{p}_3, \hat{p}_4 \) so that \( \hat{p}_4 = (0, 0, 1) \). We then have the correspondence

\[
\hat{p}_0, \hat{p}_1, \hat{p}_2, \hat{p}_3, (0, 0, 1) \in S^2 \iff p_0, p_1, p_2, p_3 \in \mathbb{R}^2. \tag{7}
\]

Here \( p_k = \Sigma(\hat{p}_k) \). We write \( p_k = (p_{k1}, p_{k2}) \). We always normalize so that \( p_0 \) lies in the positive \( x \)-axis and \( \|p_0\| \geq \|p_k\| \) for \( k = 1, 2, 3 \). Let \( \Omega \) denote those
5-point configurations which are represented by 4-tuples \((p_0, p_1, p_2, p_3)\) such that

1. \(\|p_0\| \geq \|p_k\|\) for \(k = 1, 2, 3\).
2. \(512p_0 \in [433, 498] \times [0, 0]\).
3. \(512p_1 \in [-16, 16] \times [-464, -349]\).
4. \(512p_2 \in [-498, -400] \times [0, 24]\).
5. \(512p_3 \in [-16, 16] \times [349, 364]\).

This domain is pretty tight. We tried hard to get as far away from the TBP configuration as possible. Figure 2.1 shows a picture of the sets corresponding to the definition of \(\Omega\). The grey circle is the unit circle. Note that \(T \notin \Omega\). The 4 black dots and the 4 white dots are the two nearby normalized TBP configurations. As we discuss in §4, we prove the following companion to the Big Theorem.

**Theorem 2.3 (Small)** Let \(X\) be some 5-point configuration. Suppose that we have \(G^\#_{10}(X) \leq G^\#_{10}(T)\). Then either \(X = T\) or \(X \in \Omega\).

2.3 Symmetrization

Let \(K_4\) denote the 2-dimensional set of configurations whose stereographic projections are rhombi with points in the coordinate axes. (Here \(K_4\) stands for “Klein 4 symmetry”.) The TBP and all FPs have normalizations which lie in \(K_4\).
Note that $K_4 \cap \Omega$ is the set of configurations $X = (p_0, p_1, p_2, p_3)$ such that
\[-p_2 = p_0 = (x, 0), \quad -p_1 = p_3 = (0, y), \quad x \geq y, \quad x, y \in \left[\frac{348}{512}, \frac{495}{512}\right].\]
The point $(x, y) = (1, 1/\sqrt{3})$, outside our slice, represents the TBP.

We consider the following map from $\Omega$ to $K_4$. Starting with $X$, we let $(p'_0, p'_1, p'_2, p'_3)$ be the configuration which is obtained by rotating $X$ about the origin so that $p'_{02} = p'_{22}$ and $p'_{21} < p'_{01}$. We then define
\[-p_2^* = p_0^* = \left(\frac{p'_{01} - p'_{21}}{2}, 0\right), \quad -p_3^* = p_1^* = \left(0, \frac{p'_{12} - p'_{32}}{2}\right) \tag{8}\]

The points $(p_0^*, p_1^*, p_2^*, p_3^*)$ define the symmetrized configuration $X^*$. In §6 we give some details about the following crucial result.

**Lemma 2.4 (Symmetrization)** Let $s \in [12, 15 + 25/512]$ and suppose that $X \in \Omega$. Then $R_s(X^*) \leq R_s(X)$ with equality iff $X = X^*$.

One strange thing is that the map $X \to X^*$ is not clearly related to spherical geometry. Rather, it is a linear projection with respect to the stereographic coordinates we impose on the moduli space. I found the map $X \to X^*$ experimentally, after trying many alternatives. Combining the Symmetrization Lemma and the Small Theorem, we get the following result:

**Corollary 2.5** Let $s \in [13, 15 + 25/512]$. Suppose that $R_s(X) \leq R_s(T)$ and $X \neq T$. Then $X \in K_4 \cap \Omega$.

This corollary practically finishes the proofs of the Main Theorem. It leaves us with the exploration of a 2-dimensional rectangle in the configuration space. We refer the reader to [S0] for a discussion of the endgame.

## 3 Interpolation

### 3.1 The General Approach

The purpose of this chapter is to explain the proof of Lemma 2.1, the Forcing Lemma. Recall that $G_k(r) = (4 - r^2)^k$. 
Let \( \Gamma_0 \) be the constant function and let \( \Gamma_1 = G_1 \). Recall that \( T \) is the TBP. We have \( \Gamma_0(X) = \Gamma_0(T) \) and \( \Gamma_1(X) \geq \Gamma_1(T) \) for all 5-point configurations \( X \). Indeed, as is well known, the minimizers for \( \Gamma_1 \) are precisely those configurations whose center of mass is the origin. See also \([T]\).

The distances involved in \( T \) are \( \sqrt{2}, \sqrt{3}, \sqrt{4} \). (Writing \( \sqrt{4} \) for 2 makes the equations look nicer.) Let \( R \) be some function defined on \((0, 2]\). Suppose that \((\Gamma_2, \Gamma_3, \Gamma_4)\) is one of the triples from the Forcing Lemma. Suppose we can find a combination

\[ \Gamma = a_0\Gamma_0 + \ldots + a_4\Gamma_4, \quad a_1, a_2, a_3, a_4 > 0 \]  

such that

\[ \Gamma(x) \leq R(x), \quad \Gamma(\sqrt{m}) = R(\sqrt{m}), \quad m = 2, 3, 4. \]

Suppose also that \( \Gamma_j(X) \geq \Gamma_j(T) \) for \( j = 2, 3, 4 \), with strict inequality for at least one index. Then

\[ R(X) \geq \Gamma(X) = \sum a_i\Gamma_i(X) > \sum a_i\Gamma_i(T) = \Gamma(T) = R(T). \]

If the above conditions are satisfied we call our triple \textit{good} for \( R \).

Here is how we find the coefficients \( \{a_i\} \). We impose the 5 conditions

- \( \Gamma(x) = R(x) \) for \( x = \sqrt{2}, \sqrt{3}, \sqrt{4} \).
- \( \Gamma'(x) = R'(x) \) for \( x = \sqrt{2}, \sqrt{3} \).

Here \( R' = dR/dx \) and \( \Gamma' = d\Gamma/dx \). These 5 conditions give us 5 linear equations in 5 unknowns. In the cases described below, the associated matrix is invertible and there is a unique solution. The proof of the Forcing Lemma, in each case, involves solving the matrix equation, checking positivity of the coefficients, and checking the under-approximation property.

In this chapter we will present solutions to the matrix equation and some discussion of the work in \([S0, \S 3-6]\), which establishes all the claims about our triples. My monograph \([S0, \S 3]\) has computer plots for each case. The reader can see much more extensive and interactive plots using my Java program \([S3]\).
3.2 The Matrix Solutions

For the triple is \((G_2, G_3, G_5)\). The solution is

\[
\begin{bmatrix}
0 & 0 & 0 & -144 & 0 & 0 & 0 \\
-312 & -96 & 408 & 24 & 80 & 0 & 0 \\
684 & -288 & -396 & -54 & -144 & 0 & 0 \\
-402 & 264 & 138 & 33 & 68 & 0 & 0 \\
30 & -24 & -6 & -3 & -4 & 0 & 0 \\
2496 & 768 & -3264 & -192 & -640 & -144 & 0 \\
\end{bmatrix}
\begin{bmatrix}
2^{-s/2} \\
3^{-s/2} \\
4^{-s/2} \\
2^{s-1/2} \\
3^{s-1/2} \\
4^{s-1/2} \\
\end{bmatrix}
\]

It turns out that the triple is good for \(R_s\) when \(s \in (-2, 0]\).

For the triple \(G_2, G_4, G_6\) the solution is given by

\[
\begin{bmatrix}
0 & 0 & 0 & 0 & 0 & 0 & 0 \\
-792 & 1152 & -1944 & -54 & -288 & 0 & 0 \\
-1254 & -396 & -1350 & 87 & 376 & 0 & 0 \\
-66 & -68 & 18 & 6 & 10 & 0 & 0 \\
-6336 & -9216 & 15552 & 432 & 2304 & 792 & 0 \\
\end{bmatrix}
\begin{bmatrix}
2^{-s/2} \\
3^{-s/2} \\
4^{-s/2} \\
2^{s-1/2} \\
3^{s-1/2} \\
4^{s-1/2} \\
\end{bmatrix}
\]

For the triple \(G_2, G_5, G_{10}^{\#}\) we have

\[
\begin{bmatrix}
0 & 0 & 0 & 268536 & 0 & 0 & 0 \\
88440 & 503040 & -591480 & -4254 & 65728 & 0 & 0 \\
-77586 & -249648 & 327234 & 2361 & 63896 & 0 & 0 \\
-41808 & -19440 & -22368 & -2430 & -9076 & 0 & 0 \\
-402 & 264 & 138 & 33 & 68 & 0 & 0 \\
-707520 & -4024320 & 4731840 & 34032 & 525824 & 268536 & 0 \\
\end{bmatrix}
\begin{bmatrix}
2^{-s/2} \\
3^{-s/2} \\
4^{-s/2} \\
2^{s-1/2} \\
3^{s-1/2} \\
4^{s-1/2} \\
\end{bmatrix}
\]

For the triple \(G_2, G_6, G_{10}^{\#}\) we have

\[
\begin{bmatrix}
0 & 0 & 0 & 268536 & 0 & 0 & 0 \\
947112 & 131520 & -1078632 & -50694 & -25072 & 0 & 0 \\
-91254 & -240672 & 331926 & 3483 & 68208 & 0 & 0 \\
-35778 & -15480 & -20298 & -1935 & -8056 & 0 & 0 \\
-402 & 264 & 138 & 33 & 68 & 0 & 0 \\
174268608 & 24199680 & -198468288 & -9326796 & -47669248 & 268536 & 0 \\
\end{bmatrix}
\begin{bmatrix}
2^{-s/2} \\
3^{-s/2} \\
4^{-s/2} \\
2^{s-1/2} \\
3^{s-1/2} \\
4^{s-1/2} \\
\end{bmatrix}
\]

In all cases, we also keep track of

\[
\delta = 2G'(2) - 2R'(2).
\] (12)

This quantity is also positive on the relevant intervals. We will use the positivity of \(\delta\) in our under-approximation proof below.

Even though the expressions involved above are not polynomials in general, we establish the various claims by showing that certain polynomials are positive on certain domains. There are a variety of computational positivity certificates for polynomials; I will discuss one that I thought of myself and have used on several occasions in other work. We first discuss the general approach, and then we turn the technique towards the verification of the claims in this section.
3.3 Positive Dominance

Here we discuss some positivity certificates for polynomials in one and several variables. See [S0, Chapter 5] and also [S2] for the proofs of the lemmas in this section.

Given a multi-index \( I = (i_1, ..., i_k) \in (\mathbb{N} \cup \{0\})^k \) we let
\[
x^I = x_1^{i_1} \cdots x_k^{i_k}.
\]
Any polynomial \( F \in \mathbb{R}[x_1, ..., x_k] \) can be written succinctly as
\[
F = \sum a_I x^I, \quad a_I \in \mathbb{R}.
\]

For \( I' = (i'_1, ..., i'_k) \) we write \( I' \leq I \) if \( i'_j \leq i_j \) for all \( j = 1, ..., k \). We call \( F \) weak positive dominant (WPD) if
\[
A_I := \sum_{I' \leq I} a_{I'} \geq 0 \quad \forall I,
\]
and if the sum of all the coefficients is positive. We call \( f \) positive dominant (PD) if we have strict inequality in Equation 15 for all indices.

**Lemma 3.1 (Positivity Criterion)** If \( P \) is PD, then \( P > 0 \) on \([0, 1]^k\). If \( P \) is WPD, then \( P > 0 \) on \((0, 1]^k\).

For the application we have in mind, Lemma 3.1 is not that useful, because the polynomials of interest to us are not WPD or PD in general. However, Lemma 3.1 feeds into a powerful divide-and-conquer algorithm. We define the maps
\[
A_{j,1}(x_1, ..., x_k) = (x_1, ..., x_{j-1}, \frac{x_j + 0}{2}, x_{i+1}, ..., x_k),
\]
\[
A_{j,2}(x_1, ..., x_k) = (x_1, ..., x_{j-1}, \frac{x_j + 1}{2}, x_{j+1}, ..., x_k).
\]

We define the \( j \)th subdivision of \( P \) to be the set
\[
\{P_{j,1}, P_{j,2}\} = \{P \circ A_{j,1}, P \circ A_{j,2}\}.
\]

**Lemma 3.2** For any index \( j = 1, ..., k \), the following is true. \( P > 0 \) on \([0,1]^k\) if and only if \( P_{j,1} > 0 \) and \( P_{j,2} > 0 \) on \([0,1]^k\).
We say that a marker is a non-negative vector $(a_1, \ldots, a_k) \in \mathbb{Z}^k$ such that $a_{i+1} \in \{a_i - 1, a_i\}$ for all $i = 0, \ldots, k - 1$. We order the markers lexicographically. We define the youngest entry in the marker to be the first minimum entry going from left to right. The successor of a marker is the marker obtained by adding one to the youngest entry. For instance, the successor of $(2, 2, 1, 1, 1)$ is $(2, 2, 2, 1, 1)$. We have made bold the youngest entry in our example. Let $\mu_+$ denote the successor of $\mu$.

We say that a marked polynomial is a pair $(P, \mu)$, where $P$ is a polynomial and $\mu$ is a marker. Let $j$ be the position of the youngest entry of $\mu$. We define the subdivision of $(P, \mu)$ to be the pair

$$\{(P_{j,1}, \mu_+), (P_{j,2}, \mu_+)\}.$$  

(18)

Geometrically, we are cutting the domain in half along the longest side, and using a particular rule to break ties when they occur. Now we have assembled the ingredients needed to explain the Positive Dominance Algorithm.

1. Start with a list LIST of marked polynomials. Initially, LIST consists only of the marked polynomial $(P, (0, \ldots, 0))$.

2. Let $(Q, \mu)$ be the last element of LIST. We delete $(Q, \mu)$ from LIST and test whether $Q$ is positive dominant.

3. Suppose $Q$ is positive dominant. We go back to Step 2 if LIST is not empty. Otherwise, we halt.

4. Suppose $Q$ is not positive dominant. We append to LIST the two marked polynomials in the subdivision of $(Q, \mu)$ and then go to Step 2.

If the algorithm halts, it constitutes a proof that $P \succ 0$ on $[0, 1]^k$ or $(0, 1]^k$. For the case of strict positivity, the algorithm halts if and only if $P \succ 0$ on $[0, 1]^k$. There is also a parallel version whose halting constitutes a proof that $\max(P_1, \ldots, P_m) \succ 0$, where $P_1, \ldots, P_m$ is a finite list of polynomials. In the parallel version we pass a block if one of the functions is PD (or WPD) on it.

Remark: Below we will be somewhat imprecise about whether we use the WPD version of the algorithm of the PD version. These fine details are taken care of in [S0, §4-6].
3.4 Positivity of the Coefficients

The coefficients from the previous section all have the form

\[ A(s) = \lambda_2 2^{-s/2} + \lambda_3 3^{-s/2} + \lambda_4 4^{-s/2} + \mu_2 s 2^{-s/2} + \mu_3 s 3^{-s/2} + \mu_4 s 4^{-s/2} \tag{19} \]

for \( \lambda_i, \mu_i \in Q \) for all \( i = 2, 3, 4 \). We wish to show that such an expression is positive on some interval \( I \subset (-2, 16] \). We break the interval \((-2, 16]\) into the intervals \((-2, -1], [-1, 0], [0, 1], \text{etc.}, \) and consider the problem separately on each such interval.

Let \( I \) be one of these unit intervals. Let \( k \) be the even endpoint of \( I \). For each \( m \in \{2, 3, 4\} \) we have Taylor’s Theorem with Remainder:

\[ m^{-s/2} = \sum_{j=0}^{11} \frac{(-1)^j \log(m)^j}{m^{k2j}j!} (s - 2k)^j + \frac{E_s}{12!} (s - 2k)^{12}. \tag{20} \]

Here \( E_s \) is the 12th derivative of \( m^{-s/2} \) evaluated at some point in the interval. (Going up to 12 derivatives is a somewhat arbitrary but convenient cutoff.) We replace the powers of \( \log(m) \) by nearby rational approximations in order to get close rational under-approximations and rational over-approximations to \( m^{-s/2} \) on \( I \). We then feed these approximations back into Equation 19 to get a rational polynomial \( A(s) \leq A(s) \) on \( I \). We then take the relevant sub-interval \( J \subset I \) and use the Positive Dominance Algorithm to show that \( A(s) > 0 \) on \( J \). Usually we have \( J = I \) but in Case 4 of the Forcing Lemma we sometimes have \( J = [15, 15 + \frac{25}{512}] \) and \( I = [15, 16] \).

3.5 Under Approximation

Let \( \Gamma \) be as in Equation 9. Let \( \Omega \) be the exponent interval of interest to us. For instance, in Case 1, \( \Omega = (0, 6] \). We want to show that \( \Gamma(r) \leq R_s(r) \) for all \( r \in (0, 2] \) and all \( s \in \Omega \). Define

\[ H(r, s) = 1 - \frac{\Gamma(r)}{R_s(r)} = 1 - r^s \Gamma(r), \tag{21} \]

We just have to show that \( H \geq 0 \) on \((0, 2) \times \Omega \). In Cases 1 and 2 the following lemma, has a traditional proof that exploits the low degree of the polynomials involved. The proof we sketch here is an industrial strength proof that works (one case at a time) for all the cases.
Lemma 3.3 For each fixed value $s \in \Omega$, the function $\partial_r H(r, s)$ has 4 roots in $(0, 2)$, and these roots are all simple.

Proof: (Sketch.) The positive roots of $\partial_r H(r, s)$ are the same as the positive roots of the polynomial $\psi(r, s) = s \Gamma(r) + r \Gamma'(r)$. To show that $\psi$ has simple roots, we just have to show that $\psi(r, s)$ and $\partial_r \psi(r, s)$ do not simultaneously vanish for $(r, s) \in (0, 2) \times \Omega$. Considered as polynomials in $r$, the coefficients are functions of $s$ having the same form as in Equation 19.

We work separately with each unit integer interval $I$ that intersects $\Omega$. Using the same technique as in the previous section we produce rational 2-variable polynomials $u, \overline{u}, u, \overline{v}$ such that

$$u(r, s) \leq \psi(r, s) \leq \overline{u}(r, s), \quad \psi(r, s) \leq \partial_r \psi(r, s) \leq \overline{v}(r, s)$$

for all $(r, s) \in (0, 2] \times I$. We then apply the parallel version of the Positive Dominance Algorithm to the set of functions $\{u, -\overline{u}, u, -\overline{v}\}$. This time we simply work with $I$. There is no need to restrict to a sub-interval $J \subset I$ even in Case 4. The algorithm halts, and this constitutes a proof that $\psi(r, s)$ and $\partial_r \psi(r, s)$ do not both vanish at the same point $(r, s) \in (0, 2] \times I$.

Now we turn to the number of roots. We first observe that $\partial_r H(r, s) < 0$ for $r > 0$ sufficiently small, and also $\partial_r H(2, s) < 0$. The second fact comes from the fact that the quantity $\delta$ in Equation 12 is positive for all $s \in \Omega$. We will abbreviate these two conditions as the asymptotic conditions.

The asymptotic conditions imply that $\partial_r H(r, s)$ has an even number of simple roots for $r \in (0, 2)$. The asymptotic conditions also imply the following phenomenon: As $s$ changes, the number of roots of $\partial_r H(r, s)$ can change only if 2 non-real and conjugate roots converge to the interval $(0, 2)$. But this would give $\partial_r H(r, s)$ a double root for some $s$. In other words, the asymptotic conditions imply that the number of simple roots is independent of $s$. We then check in each case, for a single value of $s$, that this number is 4. ♠

Lemma 3.4 $\partial^2_r H(r, s) > 0$ for $r = \sqrt{2}$ and $r = \sqrt{3}$.

Proof: We check this in each case for a single value of $s$. If the condition ever failed for some other value of $s$, then there would be a value of $s$ such that $\partial^2_r H(r, s) = 0$. But then $\partial_r H(r, s)$ would not have all simple roots. ♠

Lemma 3.4 combines with the asymptotic conditions and Lemma 3.3 to show that $H(r, s) \geq 0$ for all $(r, s) \in (0, 2] \times I$. 

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4 Divide and Conquer

4.1 Normalized Configurations

The purpose of this chapter is to explain the proofs of Theorem 2.2, the Big Theorem, and Theorem 2.3, the Small Theorem.

Recall that \( \hat{p}_0, \ldots, \hat{p}_4 \) is our configuration and \( p_k = \Sigma(\hat{p}_k) \) for \( k = 0, 1, 2, 3 \). As usual \( \hat{p}_4 = (0, 0, 1) \). Setting \( p_k = (p_{k1}, p_{k2}) \), we normalize so that

- \( p_0 \) lies in the positive \( x \)-axis. That is, \( p_{01} > 0 \) and \( p_{02} = 0 \).
- \( \|p_0\| \geq \max(\|p_1\|, \|p_2\|, \|p_3\|) \).
- \( p_{12} \leq p_{22} \leq p_{32} \) and \( 0 \leq p_{22} \).

If, in addition, we are working with a monotone decreasing energy function, we also require that \( p_{12} \leq 0 \). If this condition fails then our normalization implies that all the points lie in the same hemisphere. But then one can decrease the energy by reflecting one of the points across the hemisphere boundary. (This argument can fail for the one energy function we consider which is not monotone, namely \( G^\flat_5 \).) We call such configurations normalized.

Lemma 4.1 Let \( \Gamma \) be any of \( G_3, G_4, G^\flat_5, G_5, G_{10}, G^\#_{10}, G^\##_{10} \). Any normalized minimizer w.r.t. \( \Gamma \) has \( p_0 \in [0, 4] \) and \( \|p_k\| \leq 3/2 \) for \( k = 1, 2, 3 \).

Proof: The TBP has 6 bonds of length \( \sqrt{2} \), and 3 bonds of length \( \sqrt{3} \), and one bond of length 2. Hence,

\[
G_k(T) = 3(2^{k+1} + 1), \quad k = 1, 2, 3, \ldots
\]  

Suppose that \( \Gamma \) is one of the energies above, but not \( G_3 \) or \( G^\flat_5 \). If \( \|p_0\| \geq 4 \) then the distance from \( \hat{p}_0 \) to \( (0, 0, 1) \) is at most \( d = \sqrt{4/17} \). We check this by computing \( \Sigma^{-1}(4, 0) \), which would be the farthest point from \( (0, 0, 1) \) with the given constraints. We check by direct calculation that \( \Gamma(d) > \Gamma(T) \) in all cases. This single bond contributes too much to the energy all by itself. If the second condition fails then we have \( \|p_k\| > 3/2 \) for some \( k \) and also \( \|p_0\| > 3/2 \). The distance from \( \hat{p}_k \) and \( \hat{p}_0 \) from \( \hat{p}_4 \) is at most \( d' = 4/\sqrt{13} \). In all cases we compute that \( 2\Gamma(d') = \Gamma(T) \). For example \( G_4(T) = 99 \) whereas \( 2G_4(d') \approx 117.616 \).

This deals with all cases except \( G_3 \) and \( G^\flat_5 \). See [S0] for the more elaborate argument that deals with these two cases. \( \spadesuit \)
4.2 The TBP Configurations

The TBP has two kinds of points, the two at the poles and the three at the equator. When $\infty$ is a polar point, the points $p_0, p_1, p_2, p_3$ are, after suitable permutation,

$$(1,0), \quad (-1/2, -\sqrt{3}/2), \quad (0,0), \quad (-1/2, +\sqrt{3}/2).$$

We call this the polar configuration. When $\infty$ is an equatorial point, the points $p_0, p_1, p_2, p_3$ are, after suitable permutation,

$$(1,0), \quad (0, -1/\sqrt{3}), \quad (-1,0), \quad (0,1/\sqrt{3}).$$

We call this the equatorial configuration. We can visualize the two configurations together in relation to the regular 6-sided star. The black points are part of the polar configuration and the white points are part of the equatorial configuration. The grey point belongs to both configurations. The points represented by little squares are polar and the points represented by little disks are equatorial.

\[\text{Figure 4.1: Polar and equatorial versions of the TBP.}\]

For each $k = 0, \ldots, 4$ we introduce the quantity

$$\delta_k = \min_i \hat{p}_i \cdot \hat{p}_k.$$

We say that a normalized configuration is totally normalized if

$$\delta_4 \geq \delta_k, \quad k = 0, 1, 2, 3.$$ 

This condition is saying that the points in the configuration are bunched up around $\hat{p}_4 = (0,0,1)$ as much as possible. The polar TBP has $\delta_4 = -1$ and 3 values of $k$ for which $\delta_k = 0$. Hence, it is not totally normalized. The equatorial TBP is totally normalized.
4.3 Dyadic Blocks

For the moment we find it convenient to only require that $p_k \in [-2, 2]^2$ for $k = 1, 2, 3$. Later on, we will enforce the stronger condition given by Lemma 4.1. Define

$$\square = [0, 4] \times [-2, 2]^2 \times [-2, 2]^2 \times [-2, 2]^2. \tag{27}$$

Any minimizer of any of the energies we consider is isometric to one which is represented by a point in this cube. This cube is our universe.

In 1 dimension, the *dyadic subdivision* of a line segment is the union of the two segments obtained by cutting it in half. In 2 dimensions, the *dyadic subdivision* of a square is the union of the 4 quarters that result in cutting the the square in half along both directions. We say that a *dyadic segment* is any segment obtained from $[0, 4]$ by applying dyadic subdivision recursively. We say that a *dyadic square* is any square obtained from $[-2, 2]^2$ by applying dyadic subdivision recursively. We count $[0, 4]$ as a dyadic segment and $[-2, 2]^2$ as a dyadic square.

**Hat and Hull Notation:** We let $\langle X \rangle$ denote the convex hull of any Euclidean subset. Thus, we think of a dyadic square $Q$ as the set of its 4 vertices and we think of $\langle Q \rangle$ as the solid square having $Q$ as its vertex set. For any $S \subset \mathbb{R}^2$ we let $\hat{S} = \Sigma^{-1}(S)$, where $\Sigma$ is stereographic projection. In particular,

- $\hat{Q}$ is a set of 4 co-circular points on $S^2$.
- $\langle Q \rangle$ is a convex quadrilateral whose vertices are $\hat{Q}$.
- $\overline{\langle Q \rangle}$ is a “spherical patch” on $S^2$, bounded by 4 circular arcs.

**Good Squares:** A dyadic square is *good* if it is contained in $[-3/2, 3/2]^2$ and has side length at most $1/2$. Note that a good dyadic square cannot cross the coordinate axes. The only dyadic square which crosses the coordinate axes is $[-2, 2]^2$, and this square is not good. Our computer program only does spherical geometry calculations on good squares.

**Dyadic Blocks:** We define a *dyadic block* to be a 4-tuple $(Q_0, Q_1, Q_2, Q_3)$, where $Q_0$ is a dyadic segment and $Q_i$ is a dyadic square for $j = 1, 2, 3$. We say that a block is *good* if each of its 3 component squares is good. By Lemma
4.1, any energy minimizer for $G_k$ is contained in a good block. Our algorithm quickly chops up the blocks in $\square$ so that only good ones are considered.

The product

$$\langle B \rangle = \langle Q_0 \rangle \times \langle Q_1 \rangle \times \langle Q_2 \rangle \times \langle Q_3 \rangle$$

is a rectangular solid in the configuration space $\square$. On the other hand, the product

$$B = Q_0 \times Q_1 \times Q_2 \times Q_3$$

is the collection of 128 vertices of $\langle B \rangle$. We call these the \textit{vertex configurations} of the block.

**Definition:** We say that a configuration $p_0, p_1, p_2, p_3$ is \textit{in} the block $B$ if $p_i \in \langle Q_i \rangle$ for $i = 0, 1, 2, 3$. In other words, the point in $\square$ representing our configuration is contained in $\langle B \rangle$. Sometimes we will say that this configuration is \textit{associated to} the block.

**Subdivision of Blocks:** There are 4 obvious subdivision operations we can perform on a block.

- The operation $S_0$ divides $B$ into the two blocks $(Q_{00}, Q_1, Q_2, Q_3)$ and $(Q_{01}, Q_1, Q_2, Q_3)$. Here $(Q_{00}, Q_{01})$ is the dyadic subdivision of $Q_0$.
- The operation $S_1$ divides $B$ into the 4 blocks $(Q_0, Q_{1ab}, Q_2, Q_3)$, where $(Q_{100}, Q_{101}, Q_{110}, Q_{111})$ is the dyadic subdivision of $Q_1$.

The operations $S_2$ and $S_3$ are similar to $S_1$. These subdivision operations will feed into a subdivision algorithm akin to the one discussed in §3.3.

4.4 \textbf{Spherical Geometry Estimates}

We introduce some geometric quantities in this section.

- Let $d(Q)$ be the diameter $\langle \hat{Q} \rangle$.
- Let $d_1(Q)$ be the length of the longest edge of $\langle \hat{Q} \rangle$.
- Let $D_Q \subset \mathbb{R}^2$ denote the disk containing $Q$ in its boundary and $d_2(Q)$ be the diameter of $\hat{D_Q}$.
When $Q$ is a dyadic segment, we define $\delta(Q) = \chi(2,d_2)$. When $Q$ is a good dyadic square, We define

$$
\delta(Q) = \max \left( \chi(1,d_1), \chi(2,d_2) \right), \quad \chi(D,d) = \frac{d^2}{4D} + \frac{d^4}{2D^3} \tag{30}
$$

We call $\delta(Q)$ the Hull approximation constant of $Q$.

**Lemma 4.2 (Hull Approximation)** Let $Q$ be a dyadic segment or a good dyadic square. Every point of the spherical patch $\hat{\langle Q \rangle}$ is within $\delta(Q)$ of a point of the convex quadrilateral $\langle \hat{Q} \rangle$.

**Proof:** (Sketch.) The basic idea behind many of our estimates is that the set $\langle \hat{Q} \rangle$, which is a spherical patch on $S^2$, is quite close to the convex hull of the vertices of this set, namely $\langle \hat{Q} \rangle$. See [S0, §8] for more details. ♠

Let $Q$ be a dyadic segment or a good dyadic square. Let $\delta$ be the hull approximation constant of $Q$. Let $\{q_i\}$ be the points of $Q$. We make all the same definitions for a second dyadic square $Q'$. We define

$$
(Q \cdot Q')_{\max} = \max_{i,j} (\hat{q}_i \cdot \hat{q}_j) + \delta + \delta' + \delta\delta'. \tag{31}
$$

$$
(Q \cdot \{\infty\})_{\max} = \max_i \hat{q}_i \cdot (0,0,1) \tag{32}
$$

We define the same quantities with respect to $\min$, except that in the first equation we subtract $\delta + \delta' + \delta\delta'$. Let $\Omega(Q)$ denote the union of line segments that connect points in $\langle \hat{Q} \rangle$ to a nearest point in $\langle \hat{Q} \rangle$. Note that $\Omega(Q)$ contains $\langle \hat{Q} \rangle$. We define $\Omega(Q')$ similarly.

**Lemma 4.3 (Dot Product Bounds)** For all $V, V' \in \Omega(Q) \times \Omega(Q')$,

$$
(Q \cdot Q')_{\min} \leq V \cdot V' \leq (Q \cdot Q')_{\max}
$$

**Proof:** (Sketch.) The dot product is bilinear, and so the restriction of the dot product to the convex polyhedral set $\langle \hat{Q} \rangle \times \langle \hat{Q'} \rangle$ takes on its extrema at vertices. The rest of the proof comes down to several applications of the Cauchy-Schwarz Inequality, the Hull Approximation Lemma, and the triangle inequality. See [S0, §8] for more details. ♠
Let \( B = (Q_0, Q_1, Q_2, Q_3) \) be a good block. As usual \( Q_4 = \{\infty\} \). For each index \( j \) we define

\[
(B, k)_{\min} = \min_{j \neq k} (Q_j \cdot Q_k)_{\min}, \quad (B, k)_{\max} = \min_{j \neq k} (Q_j \cdot Q_k)_{\max}.
\]

(33)

We say that \( B \) is disordered if there is some \( k \in \{0, 1, 2, 3\} \) such that

\[
(B, 4)_{\max} < (B, k)_{\min}
\]

(34)

The following result is an immediate consequence of Lemma 4.3 and the definition given in Equation 25.

**Lemma 4.4** If \( B \) is disordered, then \( B \) contains no totally normalized configurations.

### 4.5 Irrelevant Blocks

Call a block *irrelevant* if no configuration in the interior of the block is totally normalized. Call a block *relevant* if it is not irrelevant. Every relevant configuration in the boundary of an irrelevant block is also in the boundary of a relevant block. So, to prove the Big and Small Theorems, we can ignore the irrelevant blocks. Let \( \overline{Q}_{jk} \) and \( \underline{Q}_{jk} \) denote the maximum and minimum \( k \)th coordinate of a point in \( Q_j \). Call \( B \) *certifiably irrelevant* if \( B \) satisfies at least one of the following conditions.

1. \( \min(|\overline{Q}_{k1}|, |\underline{Q}_{k1}|) \geq \overline{Q}_{01} \) for some \( k = 1, 2, 3 \).
2. \( \min(|\overline{Q}_{k2}|, |\underline{Q}_{k2}|) \geq \overline{Q}_{01} \) for some \( k = 1, 2, 3 \).
3. \( \overline{Q}_{12} \geq 0 \), provided we have a monotone decreasing energy.
4. \( \overline{Q}_{22} \leq 0 \).
5. \( \overline{Q}_{32} \leq \underline{Q}_{22} \).
6. \( \overline{Q}_{22} \leq \underline{Q}_{12} \).
7. \( B \) is disordered.

The reason for our terminology is that the above conditions give a computational test for showing that \( B \) is irrelevant.
Lemma 4.5 If $B$ is good and certifiably irrelevant, then $B$ is irrelevant.

Proof: Conditions 1 and 2 each imply that there is some index $k \in \{1, 2, 3\}$ such that all points in the interior of $B_k$ are farther from the origin than all points in $B_0$. Condition 3 implies that all points in the interior of $B_1$ lie above the $x$-axis. This violates our normalization when the energy function is monotone decreasing. Condition 4 implies that all points in the interior of $B_2$ lie below the $x$-axis. Condition 5 implies that all points in the interior of $B_3$ lie below all points in the interior of $B_2$. Condition 6 implies that all points in the interior of $B_2$ lie below all points in the interior of $B_1$. Lemma 4.4 takes care of Condition 7. ♠

4.6 The Energy Theorem

We think of the energy potential $G = G_k$ as being a function on $(R^2 \times \infty)^2$, via the identification $p \leftrightarrow \hat{p}$. We take $k \geq 1$ to be an integer.

Let $Q$ denote the union of the following sets.

- The set of dyadic squares in $[-2, 2]^2$.
- The set of dyadic segments in $[0, 4]$.
- The point $\{\infty\}$.

When $Q = \{\infty\}$ the corresponding hull approximation constant is 0, and the corresponding hull diameter are 0.

Now we are going to define a function $\epsilon : Q \times Q \rightarrow [0, \infty)$. First of all, for notational convenience we set $\epsilon(Q, Q) = 0$ for all $Q$. When $Q, Q' \in Q$ are unequal, we define

\[
\epsilon(Q, Q') = \frac{1}{2} k(k - 1) T^{k-2} d^2 + 2k T^{k-1} \delta
\] (35)

Here

- $d$ is the diameter of $\hat{Q}$.
- $\delta = \delta(Q)$ is the hull approximation constant for $Q$.
- $T = T(Q, Q') = 2 + 2(Q \cdot Q')_{\text{max}}$. 
This is a rational function in the coordinates of $Q$ and $Q'$. The quantities $d^2$ and $\delta$ are essentially quadratic in the side-lengths of $Q$ and $Q'$. Note that we have $\epsilon(\{\infty\}, Q') = 0$ but $\epsilon(Q, \{\infty\})$ is nonzero when $Q \neq \{\infty\}$.

Let $B = (Q_0, Q_1, Q_2, Q_3)$. For notational convenience we set $Q_4 = \{\infty\}$. We define

$$\text{ERR}_k(B) = \sum_{i=0}^{3} \sum_{j=0}^{4} \epsilon(Q_i, Q_j).$$  \hspace{1cm} (36)

This is a rational function of the vertices of $B$.

**Theorem 4.6 (Energy)**

$$\min_{v \in \langle B \rangle} \mathcal{E}_k(v) > \min_{v \in B} \mathcal{E}_k(v) - \text{ERR}_k(B).$$

**Remark:** As the form of Equation 36 suggests, a similar result holds for $N$ point configurations. Our argument in §5 readily covers this more general case.

Theorem 4.6 suffices to deal with $G_3, G_4, G_6$, but we need a more general result to deal with $G_5^\circ, G_{10}^\#$ and $G_{10}^{##}$. Suppose we have some energy of the form

$$F = \sum_{k=1}^{N} a_k G_k$$  \hspace{1cm} (37)

where $a_1, ..., a_N$ is some sequence of numbers, not necessarily positive.

Suppressing the dependence on $F$, we define

$$\epsilon(Q_i, Q_j) = \sum |a_k| \epsilon_k(Q_i, Q_j).$$  \hspace{1cm} (38)

where $\epsilon_k(Q_i, Q_j)$ is the above quantity computed with respect to $G_k$. We then define $\text{ERR}$ exactly as in Equation 36. With this definition, we have

**Corollary 4.7** Suppose that $F$ is as in Equation 37. Suppose that $B$ is a block such that

$$\min_{v \in B} \mathcal{E}_F(v) - \text{ERR}_F(B) > \mathcal{E}_F(\text{TBP}).$$  \hspace{1cm} (39)

Then all configurations in $B$ have higher energy than the TBP.
We can write
\[ \text{ERR}(B) = \sum_{i=0}^{3} \text{ERR}_i(B), \quad \text{ERR}_i(B) = \sum_{j=0}^{4} \epsilon(Q_i, Q_j). \quad (40) \]

We define the *subdivision recommendation* to be the index \( i \in \{0, 1, 2, 3\} \) for which \( \text{ERR}_i(B) \) is maximal. A tie never arises in practice, but in the event of a tie we would pick the smaller index.

### 4.7 Estimates on the Hessian

The Energy Theorem is too crude to deal with configurations very near the TBP. Since we are considering totally normalized configurations we only have to worry about a neighborhood of the equatorial TBP. Let \( B_0 \subset \Box \) denote the cube of side-length \( 2^{-17} \) centered at the configuration representing the equatorial TBP. We consider our energy functions on the configuration space \( \Box \). We set
\[ \Gamma_k = G_k \circ \Sigma^{-1}. \quad (41) \]
Here \( \Sigma \) is stereographic projection. The fact that \( \Sigma^{-1} \) is a rational function means that all the functions of interest to us are rational functions on \( \Box \). In [S0, §11] we prove the following result.

**Lemma 4.8 (Hessian)** For each \( \Gamma = \Gamma_3, \Gamma_4, \Gamma_5, \Gamma_6, \Gamma_7, \Gamma_8, \Gamma_9, \Gamma_{10}, \Gamma_{11} \), the Hessian of \( \Gamma \) is positive definite at all points of \( B_0 \).

The basic idea is to get global upper bounds on the 8th partial derivatives of \( \Gamma \) which hold on all of \( B_0 \) using crude combinatorial information about the function. The bounds we get are on the order of \( 2^{90} \). We then essentially use Taylor’s Theorem with Remainder, evaluating at the point representing the equatorial TBP, to bound the lower derivatives. There is nothing sacred about the cutoff at the 8 derivative; this is just what worked. Also, it seems possible that we could replace \( B_0 \) by a considerably larger region. However, the proof for a much larger region seems much more difficult.

### 4.8 The Main Algorithm

We fix some energy function \( F \) that appears in either the Big Theorem or the Small Theorem. Our divide and conquer algorithm works schematically like
the one described in §3.3, except that we don’t use the Positivity Criterion from Lemma 3.1 and we determine the subdivision order differently. Rather, we perform the following test to each block \( B = (Q_0, Q_1, Q_2, Q_3) \).

1. If \( B \) lies entirely in the region \textbf{SMALL} and we are working with the function \( G_{10}^2 \) we pass \( B \). This is the extra step for the Small Theorem.

2. If some component square \( Q_i \) of \( B \) has side length more than \( 1/2 \) we fail \( B \) and recommend that \( B \) be subdivided along the first such index. This step guarantees that we only pass good blocks.

3. If \( B \) is certifiably irrelevant, we pass \( B \).

4. If we compute that \( Q_i \nsubseteq [-3/2, 3/2]^2 \) for some \( i = 1, 2, 3 \), we pass \( B \). Given Step 2, we know that \( Q_i \) is disjoint from \( (-3/2, 3/2)^2 \).

5. If \( B \subset B_0 \), the small cube from the Hessian Lemma, we pass \( B \).

6. If \( B \) satisfies Corollary 4.7, we pass \( B \). Otherwise, we fail \( B \) and subdivide the block according to the subdivision recommendation.

In [S0, §10] we explain how we implement our calculations using interval arithmetic so as to avoid round-off error. We also detail the results of the calculations. In short, everything runs to completion successfully and the whole calculation takes about 12 hours on a modern MacBook Pro. This gives a rigorous computer-assisted proof of the Big Theorem and the Small Theorem. The reader can see the algorithm run on our Java program [S3].

5 Proof of the Energy Theorem

5.1 A Polynomial Inequality

In this chapter we prove Theorem 4.6. For the (dis)interested reader, we remark that the discussion of symmetrization in the next chapter does not depend on the work here. Theorem 4.6 builds on the case \( M = 4 \) of the inequality below. My motivation was to get an expression that varied quadratically regardless of the exponent \( k \). I am grateful to the referee for pointing out a very efficient proof of the following preliminary lemma.
Lemma 5.1  Suppose \(a, x \in [0, 1]\) and \(k \geq 2\). Then \(f(x) \leq g(x)\), where

\[
f(x) = (ax^k + 1 - a) - (ax + 1 - a)^k; \quad g(x) = \frac{1}{8}k(k - 1)(1 - x)^2. \tag{42}
\]

Proof: Since \(f(1) = g(1) = f'(1) = g'(1) = 0\) the Cauchy Mean Value Theorem (applied twice) tells us that for any \(x \in (0, 1)\) there are values \(y < z \in [x, 1]\) such that

\[
\frac{f(x)}{g(x)} = \frac{f'(y)}{g'(y)} = \frac{f''(z)}{g''(z)} = 4az^{k-2}\left[1 - a\left(1 - a\right)^{k-2}\right] \leq 4a(1-a) \leq 1. \tag{43}
\]

This completes the proof. ♠

Here is the inequality mentioned above.

Lemma 5.2  Let \(M \geq 2\) and \(k = 1, 2, 3, \ldots\). Suppose

- \(0 \leq x_1 \leq \ldots \leq x_M\)
- \(\sum_{i=1}^{M} \lambda_i = 1\) and \(\lambda_i \geq 0\) for all \(i\).

Then

\[
0 \leq \sum_{i=1}^{M} \lambda_i x_i^k - \left(\sum_{i=1}^{M} \lambda_i x_i\right)^k \leq \frac{1}{8}k(k - 1)x_M^{k-2} (x_M - x_1)^2. \tag{44}
\]

The lower bound is a trivial consequence of convexity, and both bounds are trivial when \(k = 1\). So, we take \(k = 2, 3, 4, \ldots\) and prove the upper bound. Suppose first that \(M \geq 3\). We have one degree of freedom when we keep \(\sum \lambda_i x_i\) constant and try to vary \(\{\lambda_j\}\) so as to maximize the left hand side of the inequality. The right hand side does not change when we do this, and the left hand side varies linearly. Hence, the left hand size is maximized when \(\lambda_i = 0\) for some \(i\). But then any counterexample to the lemma for \(M \geq 3\) gives rise to a counter example for \(M - 1\). Hence, it suffices to prove the inequality when \(M = 2\).

In the case \(M = 2\), we set \(a = \lambda_1\). Both sides of the inequality in Lemma 5.2 are homogeneous of degree \(k\), so it suffices to consider the case when \(x_2 = 1\). We set \(x = x_1\). Our inequality then becomes exactly the one treated in Lemma 5.1. ♠
5.2 The Local Energy Lemma

Let \( Q = \{ q_1, q_2, q_3, q_4 \} \) be the vertex set of \( Q \in \mathcal{Q} \). We allow for the degenerate case that \( Q \) is a line segment or \( \{ \infty \} \). In this case we just list the vertices multiple times, for notational convenience.

Note that every point in the convex quadrilateral \( \langle \hat{Q} \rangle \) is a convex average of the vertices. For each \( z \in \langle Q \rangle \), there is a point \( z^* \in \langle \hat{Q} \rangle \) which is as close as possible to \( \hat{z} \in \langle \hat{Q} \rangle \). There are constants \( \lambda_i(z) \) such that

\[
z^* = \sum_{i=1}^{4} \lambda_i(z) \hat{q}_i, \quad \sum_{i=1}^{4} \lambda_i(z) = 1.
\]  

(45)

We think of the 4 functions \( \{ \lambda_i \} \) as a partition of unity on \( \langle Q \rangle \). The choices above might not be unique, but we make such choices once and for all for each \( Q \). We call the assignment \( Q \to \{ \lambda_i \} \) the stereographic weighting system.

Lemma 5.3 (Local Energy) Let \( \epsilon \) be the function defined in the Theorem 4.6. Let \( Q, Q' \) be distinct members of \( \mathcal{Q} \). Given any \( z \in Q \) and \( z' \in Q' \),

\[
\left| G(z, z') - \sum_{i=1}^{4} \lambda_i(z)G(q_i, z') \right| \leq \epsilon(Q, Q').
\]  

(46)

Proof: For notational convenience, we set \( w = z' \). Let

\[
X = (2 + 2z^* \cdot \hat{w})^k.
\]  

(47)

The Local Energy Lemma follows from the triangle inequality and the following two inequalities

\[
\left| \sum_{i=1}^{4} \lambda_i G(q_i, w) - X \right| \leq \frac{1}{2} k(k-1)T^{k-2}d^2
\]  

(48)

\[
|X - G(z, w)| \leq 2kT^{k-1}\delta.
\]  

(49)

We will establish these inequalities in turn.

Let \( q_1, q_2, q_3, q_4 \) be the vertices of \( Q \). Let \( \lambda_i = \lambda_i(z) \). We set

\[
x_i = 4 - \| \hat{q}_i - \hat{w} \|^2 = 2 + 2\hat{q}_i \cdot \hat{w}, \quad i = 1, 2, 3, 4.
\]  

(50)

25
Note that \( x_i \geq 0 \) for all \( i \). We order so that \( x_1 \leq x_2 \leq x_3 \leq x_4 \). We have
\[
\sum_{i=1}^{4} \lambda_i(z)G(q_i, w) = \sum_{i=1}^{4} \lambda_i x_i^k, \tag{51}
\]
\[
X = (2 + 2 \hat{z} \cdot \hat{w})^k = \left( \sum_{i=1}^{4} \lambda_i(2 + \hat{q}_i \cdot \hat{w}) \right)^k = \left( \sum_{i=1}^{4} \lambda_i x_i \right)^k. \tag{52}
\]
\[
\left| \sum_{i=1}^{4} \lambda_i G(q_i, w) - X \right| = \left| \sum_{i=1}^{4} \lambda_i x_i^k - \left( \sum_{i=1}^{4} \lambda_i x_i \right)^k \right| \leq \frac{1}{8} k(k-1)x_4^{k-2}(x_4 - x_1)^2. \tag{53}
\]
By Equation 51, Equation 52, and the case \( M = 4 \) of Lemma 5.2,
\[
\left| \sum_{i=1}^{4} \lambda_i G(q_i, w) - X \right| \leq \frac{1}{8} k(k-1)x_4^{k-2}(x_4 - x_1)^2. \tag{54}
\]
By Lemma 4.3, we have
\[
x_4 = 2 + 2(\hat{q}_4 \cdot \hat{w}) \leq 2 + 2(Q \cdot Q')_{\text{max}} = T. \tag{55}
\]
Since \( d \) is the diameter of \( \langle \hat{Q} \rangle \) and \( \hat{w} \) is a unit vector,
\[
x_4 - x_1 = 2\hat{w} \cdot (\hat{q}_4 - \hat{q}_1) \leq 2\|\hat{w}\|\|\hat{q}_4 - \hat{q}_1\| = 2\|\hat{q}_4 - \hat{q}_1\| \leq 2d. \tag{56}
\]
Plugging Equations 54 and 55 into Equation 53, we get Equation 48.

Now we establish Equation 49. Let \( \gamma \) denote the unit speed line segment connecting \( \hat{z} \) to \( z^* \). Note that the length \( L \) of \( \gamma \) is at most \( \delta \), by the Hull Approximation Lemma. Define
\[
f(t) = \left( 2 + 2\hat{w} \cdot \gamma(t) \right)^k. \tag{57}
\]
We have \( f(0) = X \). Since \( \hat{w} \) and \( \gamma(1) = \hat{z} \) are unit vectors, \( f(L) = G(z, w) \). Hence
\[
X - G(z, w) = f(0) - f(L), \quad L \leq \delta. \tag{58}
\]
By the Chain Rule,
\[
f'(t) = (2\hat{w} \cdot \gamma'(t)) \times k \left( 2 + 2\hat{w} \cdot \gamma(t) \right)^{k-1}. \tag{59}
\]
Note that \( |2\hat{w} \cdot \gamma'(t)| \leq 2 \) because both of these vectors are unit vectors. \( \gamma \) parametrizes one of the connectors from Lemma 4.3, so
\[
|f'(t)| \leq 2k \left( 2 + 2\hat{w} \cdot \gamma(t) \right)^{k-1} \leq 2k \left( 2 + 2(Q \cdot Q')_{\text{max}} \right)^{k-1} = 2kT^{k-1}. \tag{60}
\]
Equation 49 now follows from Equation 57, Equation 59, and integration. ♠
5.3 From Local to Global

Let $\epsilon$ be the function from the Energy Theorem. Let $B = (Q_0, \ldots, Q_N)$ be a list of $N + 1$ elements of $Q$. We care about the case $N = 4$ and $Q_4 = \{\infty\}$, but the added generality makes things clearer. Let $q_i, q_i, q_i, q_i, q_i, q_i$ be the vertices of $Q_i$. The vertices of $\langle B \rangle$ are indexed by a multi-index $I = (i_0, \ldots, i_n) \in \{1, 2, 3, 4\}^{N+1}$.

Given such a multi-index, which amounts to a choice of vertex of $\langle B \rangle$, we define the energy of the corresponding vertex configuration:

$$E(I) = E(q_{i_0, i_0}, \ldots, q_{i_N, i_N})$$ (60)

We will prove the following sharper result.

**Theorem 5.4** Let $z_0, \ldots, z_N \in \langle B \rangle$. Then

$$\left| E(z_0, \ldots, z_N) - \sum_I \lambda_{i_0}(z_0) \ldots \lambda_{i_N}(z_N) E(I) \right| \leq \sum_{i=0}^{N} \sum_{j=0}^{N} \epsilon(Q_i, Q_j).$$ (61)

The sum is taken over all multi-indices.

**Lemma 5.5** Theorem 5.4 implies Theorem 4.6.

**Proof:** Notice that

$$\sum_I \lambda_{i_0}(z_0) \ldots \lambda_{i_N}(z_N) = \prod_{j=0}^{N} \left( \sum_{a=1}^{4} \lambda_a(z_j) \right) = 1.$$ (62)

Therefore

$$\min_{v \in B} E(v) \leq \sum_I \lambda_{i_0}(z_0) \ldots \lambda_{i_N}(z_N) E(I) \leq \max_{v \in B} E(v),$$ (63)

because the sum in the middle is the convex average of vertex energies.

We will deal with the min case of Theorem 4.6. The max case has the same treatment. Choose some $(z_1, \ldots, z_N) \in B$ which minimizes $E$. We have

$$0 \leq \min_{v \in B} E(v) - \min_{v \in \langle B \rangle} E(v) = \min_{v \in B} E(v) - E(z_0, \ldots, z_N) \leq \sum_I \lambda_{i_0}(z_0) \ldots \lambda_{i_N}(z_N) E(I) - E(z_0, \ldots, z_N) \leq \sum_{i=0}^{N} \sum_{j=0}^{N} \epsilon(Q_i, Q_j).$$ (64)

The last expression is ERR when $N = 4$ and $Q_4 = \infty$. ♠

We now prove Theorem 5.4.
5.3.1 A Warmup Case

Consider the case when $N = 1$. Setting $\epsilon_{ij} = \epsilon(Q_i, Q_j)$, the Local Energy Lemma gives us

$$G(z_0, z_1) \geq \sum_{\alpha=1}^{4} \lambda_{\alpha}(z_0)G(q_{0\alpha}, z_1) - \epsilon_{01}. \quad (65)$$

$$G(q_{0\alpha}, z_1) \geq \sum_{\beta=1}^{4} \lambda_{\beta}(z_1)G(q_{1\beta}(z_1), q_{0\alpha}) - \epsilon_{10}. \quad (66)$$

Plugging the second equation into the first and using $\sum \lambda_{\alpha}(z_0) = 1$, we have

$$G(z_0, z_1) \geq \left( \sum_{\alpha, \beta} \lambda_{\alpha}(z_0)\lambda_{\beta}(z_1)G(q_{0\alpha}, q_{1\beta}) \right) - (\epsilon_{01} + \epsilon_{10}). \quad (67)$$

Similarly,

$$G(z_0, z_1) \leq \left( \sum_{\alpha, \beta} \lambda_{\alpha}(z_0)\lambda_{\beta}(z_1)G(q_{0\alpha}, q_{1\beta}) \right) + (\epsilon_{01} + \epsilon_{10}). \quad (68)$$

Equations 67 and 68 are equivalent to Equation 61 when $N = 1$.

5.3.2 The General Case

Now assume that $N \geq 2$. We rewrite Equation 67 as follows:

$$G(z_0, z_1) \geq \sum_{A} \lambda_{A_0}(z_0)\lambda_{A_1}(z_1)G(q_{0A_0}, q_{1A_1}) - (\epsilon_{01} + \epsilon_{10}). \quad (69)$$

The sum is taken over multi-indices $A$ of length 2.

We also observe that

$$\sum_{I'} \lambda_{i_2}(z_2)...\lambda_{i_N}(z_N) = 1. \quad (70)$$

The sum is taken over all multi-indices $I' = (i_2, ..., i_N)$. Therefore, if $A$ is held fixed, we have

$$\lambda_{A_0}(z_0)\lambda_{A_1}(z_1) = \sum_{I} \lambda_{I_0}(z_0)...\lambda_{I_N}(z_N). \quad (71)$$
The sum is taken over all multi-indices of length \( N + 1 \) which have \( I_0 = A_0 \) and \( I_1 = A_1 \). Combining these equations, we have

\[
G(z_0, z_1) \geq \sum_I \lambda_{I_0}(z_0) \ldots \lambda_{I_N}(z_N) G(q_{0I_0}, q_{1I_1}) = -(\epsilon_{01} + \epsilon_{10}). \tag{72}
\]

The same argument works for other pairs of indices, giving

\[
G(z_i, z_j) \geq \sum_I \lambda_{I_0}(z_0) \ldots \lambda_{I_N}(z_N) G(q_{iI_i}, q_{jI_j}) = -(\epsilon_{ij} + \epsilon_{ji}). \tag{73}
\]

Now we interchange the order of summation and observe that

\[
\sum_{i<j} \left( \sum_I \lambda_{I_0}(z_0) \ldots \lambda_{I_N}(z_N) G(q_{iI_i}, q_{jI_j}) \right) = \sum_I \sum_{i<j} \lambda_{I_0}(z_0) \ldots \lambda_{I_N}(z_N) G(q_{iI_i}, q_{jI_j}) = \sum_I \lambda_{I_0}(z_0) \ldots \lambda_{I_N}(z_N) \left( \sum_{i<j} G(q_{iI_i}, q_{jI_j}) \right) = \sum_I \lambda_{I_0}(z_0) \ldots \lambda_{I_N}(z_N) E(I). \tag{74}
\]

When we sum Equation 73 over all \( i < j \), we get

\[
E(z_0, \ldots, z_N) \geq \sum_I \lambda_{i_0}(z_0) \ldots \lambda_{i_N}(z_N) E(I) - \sum_{i=0}^{N} \sum_{j=0}^{N} e(Q_i, Q_j). \tag{75}
\]

Similarly,

\[
E(z_0, \ldots, z_N) \leq \sum_I \lambda_{i_0}(z_0) \ldots \lambda_{i_N}(z_N) E(I) + \sum_{i=0}^{N} \sum_{j=0}^{N} e(Q_i, Q_j). \tag{76}
\]

These two equations together are equivalent to Theorem 5.4. This completes the proof.
6 Symmetrization

6.1 A Three Step Process

In this chapter we sketch the ideas behind Lemma 2.4, the Symmetrization Lemma. Our symmetrization is a retraction from the small domain $\Omega$ to $K_4$, the set of configurations with 4-fold Klein symmetry. Here we describe it as a 3 step process. We start with the configuration $X$ having points $(p_0, p_1, p_2, p_3) \in \Omega$.

1. (Rotating) We let $(p'_0, p'_1, p'_2, p'_3)$ be the configuration which is obtained by rotating $X$ about the origin so that $p'_0$ and $p'_2$ lie on the same horizontal line, with $p'_0$ lying on the right. This slight rotation does not change the energy of the configuration, but it does slightly change the domain. While $X \in \Omega$, the new configuration $X'$ lies in a slightly modified domain $\Omega'$ which we describe in [S0, §13].

2. (Horizontal Sliding) Given a configuration $X' = (p'_0, p'_1, p'_2, p'_3) \in \Omega'$, there is a unique configuration $X'' = (p''_0, p''_1, p''_2, p''_3)$, invariant under reflection in the $y$-axis, such that $p'_j$ and $p''_j$ lie on the same horizontal line for $j = 0, 1, 2, 3$ and $\|p''_0 - p''_2\| = \|p'_0 - p'_2\|$. There is a slightly different domain $\Omega''$ which contains $X''$. Again, we describe $\Omega''$ precisely in [S0, §13]. The domain $\Omega''$ is 4-dimensional.

3. (Vertical Sliding) Given a configuration $X'' = (p''_0, p''_1, p''_2, p''_3) \in \Omega''$ there is a unique configuration $X* = (p*_0, p*_1, p*_2, p*_3) \in K_4$ such that $p''_j$ and $p*_j$ lie on the same vertical line for $j = 0, 1, 2, 3$.

We prove the Symmetrization Lemma in two steps. First, we prove the following result.

**Lemma 6.1** $R_s(X'') \leq R_s(X')$ for all $X' \in \Omega'$ and $s \geq 2$, with equality iff $X' = X''$.

Second, we prove the following result.

**Lemma 6.2** $R_s(X*) \leq R_s(X'')$ for all $X'' \in \Omega'$ and $s \in [12, 15 + 1/2]$, with equality iff $X'' = X*$.
Remarks:
(1) Horizontal and Vertical Sliding are really the same operation. The one operation is just the other one “turned sideways”, so to speak. However, it is useful to separate them out into two steps, because the domains involved are different and the resulting estimates we can get are also different.
(2) I tested these inequalities extensively on random inputs before attempting a proof. The inequalities seem to be quite robust, and work for a much wider domain than just the domains \( \Omega' \) and \( \Omega'' \) derived from \( \Omega \). However, the proof strategy, described in the next section, requires the small domains to work. I don’t know a robust proof.
(3) It seems that variants of Lemmas 6.1 and 6.2 might be useful in proving 5-point configurations for \( s > 15 + \frac{25}{512} \), but I could not make any headway on this.

6.2 Horizontal Symmetrization Proof Strategy

There are 10 bonds (i.e. distances between pairs of points) in a 5 point configuration \( \hat{p}_0, \ldots, \hat{p}_4 \). Ultimately, when we perform the symmetrization operations we want to see that a sum of 10 terms decreases. Doing this directly seems extremely difficult, given the complexity of the expressions involved. What makes our technique work is that we found a way to break the 10-term sum into smaller pieces, all of which decrease separately under the operation. This stronger kind of monotonicity seems to require the very small domains we use. Here are the pieces:

\[
A_{1,s} = R_s(\hat{p}_1, \hat{p}_0) + R_s(\hat{p}_1, \hat{p}_2).
\]

\[
A_{2,s} = R_s(\hat{p}_3, \hat{p}_0) + R_s(\hat{p}_3, \hat{p}_2).
\]

\[
B_{13,s} = R_s(\hat{p}_1, \hat{p}_3) + R_s(\hat{p}_1, \hat{p}_4) + R_s(\hat{p}_3, \hat{p}_4).
\]

\[
B_{02,s} = R_s(\hat{p}_0, \hat{p}_2) + R_s(\hat{p}_0, \hat{p}_4) + R_s(\hat{p}_2, \hat{p}_4).
\]

To prove Lemma 6.1 we establish the following stronger results. These results are meant to hold for all \( s \geq 2 \) and all \( X' \in \Omega' \).

1. \( A_{1,s}(X'') \leq A_{1,s}(X') \).
2. $A_{3,s}(X'') \leq A_{3,s}(X')$.

3. $B_{02,s}(X'') \leq B_{02,s}(X')$ with equality iff $X' = X''$.

4. $B_{13,s}(X'') \leq B_{13,s}(X')$ with equality iff $X' = X''$.

Inequalities 3 and 4 must be proved for each exponent $s$, but they are pretty easy. Each inequality involves a pair of points in the plane. Our proof is more or less geometric. See [S0, §14].

Inequalities 1 and 2 seem much harder, and I don’t know a geometric proof. However, we first note the following standard result.

**Lemma 6.3 (Monotonicity)** Let $\lambda_1, \ldots, \lambda_n > 0$. If $\beta/\alpha > 1$. Then

$$
\sum_{i=1}^{N} \lambda_i^\beta - N \geq \frac{\beta}{\alpha} \left( \sum_{i=1}^{N} \lambda_i^\alpha - N \right).
$$

See [S0, §12.1] for a proof. It follows from the Monotonicity Lemma that the truth of Inequality 1 at $s = 2$ implies the truth of Inequality 1 for $s > 2$. The same goes for Inequality 2. We deal with the case $s = 2$ with a direct calculation involving the positivity certificate from Lemma 3.1.

Consider Inequality 1. Since the $x$-coordinates of $p_0$ and $p_2$ are the same, we just have a 5-dimensional space of possibilities. We find 4 maps $f_1, f_2, f_3, f_4$ from the unit cube $[0, 1]^5$ into the configuration space of triples of points such that the union $\bigcup f_j([0, 1]^5)$ covers the projection of the domain $\Omega'$ onto the first 6 coordinates. That is, every triple $(p_0, p_1, p_2)$ that arises in a configuration of $\Omega'$ lies in the images of one of the cubes.

We then define

$$
\phi_j = (A_{s}^{\text{bottom}}(X') - A_{s}^{\text{bottom}}(X'')) \circ f_j.
$$

The function $\phi_j$ is a rational function on the unit cube for each $j = 1, 2, 3, 4$. The numerator and denominators of $\phi_j$ are enormous. They have several thousand terms and degree about 40. Nonetheless, we show that both the numerator and denominator satisfy the Positivity Criterion in Lemma 3.1! This establishes Inequality 1. The proof for Inequality 2 is similar. See [S0, §13,14] for more details.
6.3 Vertical Symmetrization in Brief

It seems that Inequalities 1-4 also work in the case of Lemma 6.2, except that for Inequalities 1 and 2 we need to take $s$ fairly large. I found failures for exponents as high as $s = 9$. In principle, we could prove Lemma 6.2 using a strategy similar to the one used for Lemma 6.1, except that we would pick (say) $s = 12$ for Inequalities 1 and 2. The problem is that the high exponent leads to enormous polynomials. When trying to understand the failure of Inequalities 1 and 2 for small exponents, I noticed the following phenomenon: When Inequality 1 or 2 fails, Inequalities 3 and 4 hold by a wide margin. The proof of Lemma 6.2, given in [S0, §13-15], makes calculations for Inequalities 1 and 2 at the exponent $s = 2$ similar to the ones described for Lemma 6.1, and then uses $L_p$ estimates to exploit the phenomenon just mentioned. The proof is considerably more delicate.

Remark: Lemma 6.2 has a painful proof, and one would like to do without this lemma. To this end, we mention that Lemma 6.1 implies that any minimizer for $R_s$ with $s \in [13, 15 + \frac{25}{512}]$ has to have 3 points on an equator and the remaining two points symmetrically placed with respect to reflection in this equator. It seems that there should be many endgames from here which avoid Lemma 6.2.

7 References


[BHS], S. V. Bondarenko, D. P. Hardin, E.B. Saff, *Mesh Ratios for Best
Packings and Limits of Minimal Energy Configurations,


[HS], Xiaorong Hou and Junwei Shao, Spherical Distribution of 5 Points with Maximal Distance Sum, Discrete and Computational Geometry, Volume 46, Issue 1 (2011) pp 156-174


[MKS], T. W. Melnyk, O. Knop, W.R. Smith, Extremal arrangements of point and unit charges on the sphere: equilibrium configurations revisited, Canadian Journal of Chemistry 55.10 (1977) pp 1745-1761


veys and Monographs 219, 2017

[S3] R. E. Schwartz, Java Program (available by download), http://www.math.brown.edu/res/Java/TBP.tar

[SK] E. B. Saff and A. B. J. Kuijlaars, Distributing many points on a Sphere, Math. Intelligencer, Volume 19, Number 1, December 1997 pp 5-11


[Y], V. A. Yudin, Minimum potential energy of a point system of charges (Russian) Diskret. Mat. 4 (1992), 115-121, translation in Discrete Math Appl. 3 (1993) 75-81