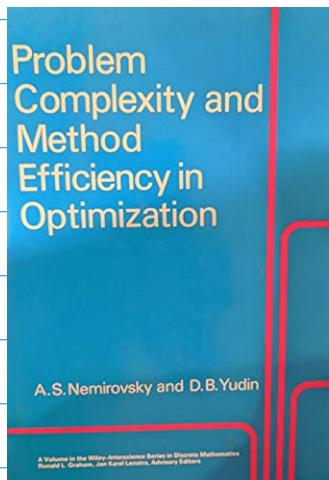


Complexity in continuous optimization (4/6)

February 13, 2025

Today: From blackbox complexity to computational complexity



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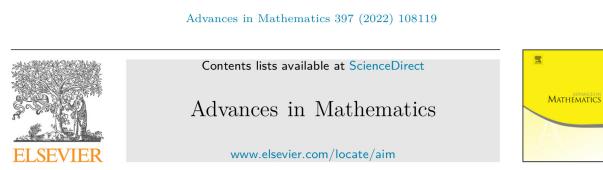
SOME NP-COMPLETE PROBLEMS IN QUADRATIC AND NONLINEAR PROGRAMMING

Katta G. MURTY*

Department of Industrial and Operations Engineering, The University of Michigan, 1205 Beal Avenue,
Ann Arbor, MI 48109-2117, USA

Santosh N. KABADI**

Faculty of Administration, University of New Brunswick, Fredericton, NB, Canada E3B 5A6



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Complexity aspects of local minima and related notions *

Amir Ali Ahmadi^{a,*}, Jeffrey Zhang^b

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FULL LENGTH PAPER

Series A



The computational complexity of finding stationary points
in non-convex optimization

Alexandros Hollender¹ · Manolis Zampetakis²

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① Motivation

Setup: minimize $f(x)$, $f: \mathbb{R}^n \rightarrow \mathbb{R}$ nonconvex (C^1, C^2, C^3)
 $x \in \mathbb{R}^n$

Blackbox complexity: Given an oracle (e.g. $x \mapsto (f(x), \nabla f(x))$)
 how many oracle calls do we need to
 reach an ε -approximate solution as a function
 of ε ?
 e.g. $\|(\nabla f(x))\| \leq \varepsilon$

$$\begin{cases} \|(\nabla f(x))\| \leq \varepsilon \\ \text{dist}(\nabla f(x)) \geq -\varepsilon \end{cases}$$

$$\|x - x^*\| \leq \varepsilon \quad \text{where } x^* \text{ is a global minimum of } f(x)$$

Computational complexity (for this Lecture)

For this nonconvex problem,

- ① How hard is it to check that a point is a local minimum or a stationary point?
- ② How hard is it to determine whether the problem has a local minimum or a stationary point of some kind?

x^* local minimum of f if $\exists \delta > 0$, $f(x^*) \leq f(u) \forall u \in \mathbb{R}^n$
 $\|x - x^*\| \leq \delta$

NB: Since f is nonconvex, it can have local but non-global minima

x^* first-order stationary point (also first-order critical point)
 if $\|\nabla f(x^*)\| = 0$

x^* second-order stationary point if $\begin{cases} \|\nabla f(x^*)\| = 0 \\ \lambda_{\min}(\nabla^2 f(x^*)) \geq 0 \end{cases}$

N.B.: For a convex function, the three concepts are the same (and they also coincide with the notion of global minimum)

For nonconvex functions,

x^* local minimum $\Rightarrow x^*$ 2nd-order stationary
 $\Rightarrow x^*$ 1st-order stationary

(2) Local minima: hard (Monty & Kabadi, 1987)

→ Decision problem: Subset sum problem

Given d_0, d_1, \dots, d_m integer values, is there a solution to

$$\sum_{j=1}^m d_j y_j = d_0 \quad \text{with } y_j \in \{0, 1\} \quad \forall j = 1..m$$

→ Known to be NP-complete

→ With the same d_0, d_1, \dots, d_m , consider the optimization problem

Nonconvex quadratic program (QP) ↑ Nonconvex quadratic

$$\left\{ \begin{array}{l} \text{minimize}_{y \in \mathbb{R}^n} \quad \left(\sum_{j=1}^m d_j y_j - d_0 \right)^2 + \sum_{j=1}^m y_j (1-y_j) \\ \text{s.t.} \quad 0 \leq y \leq 1 \quad (\Leftrightarrow 0 \leq y_j \leq 1 \quad \forall j=1..m) \end{array} \right.$$
linear constraints

The Subset sum problem has a solution if and only if the optimal value of (QP) is 0.

\Rightarrow Finding the optimal value of (QP) is NP-hard.

\hookrightarrow From this example, the authors conclude that in general computing a global minimum of a nonconvex problem is hard.

Other questions in the paper: Answer through completeness.

Completeness

Given $D \in \mathbb{Z}^{n \times n}$. Checking $D \succeq 0$ ($\Leftrightarrow x^T D x \geq 0 \quad \forall x \in \mathbb{R}^n$) corresponds to the decision problem:
 Does there exist $x \in \mathbb{R}^n$ such that $x^T D x < 0$?

\Rightarrow Can answer in polynomial time

(e.g. Gaussian pivots, $O(n^3)$ ops)

\Rightarrow Implies that checking $\nabla^2 f(x) \succeq 0$ for some f and x can be done in polynomial time

(hence checking that a point is 2nd-order stationary point can be done in polynomial time)

- $D \in \mathbb{R}^{n \times n}$ is copositive if $x^T D x \geq 0 \quad \forall x \geq 0$

Decision pb.: Does there exist x such that
 $x \geq 0$ and $x^T D x < 0$?

Unless D is $\succeq 0$, NP-complete

$$\Rightarrow \begin{array}{l} \text{minimize}_{x \in \mathbb{R}^n} x^T D x \text{ s.t. } x \geq 0 \end{array} \quad \text{NP-hard}$$

\hookrightarrow Can use copositive matrices to reduce unconstrained nonconvex optimization problems to nonconvex QPs

* Consider

$$(P) \quad \begin{array}{l} \text{minimize}_{z \in \mathbb{R}^n} h(z) = \begin{bmatrix} z_1^2 \\ 1 \\ z_m^2 \end{bmatrix}^T D \begin{bmatrix} z_1^2 \\ \vdots \\ z_m^2 \end{bmatrix} \end{array} \quad \text{w.h. } D \in \mathbb{R}^{m \times m}, D = D^T$$

\rightarrow Quartic polynomial (polynomial of degree 4 in the entries of z)

with the change of variables $x = z^2$ ($x_i = z_i^2 \quad i=1..n$),
 the problem (P) is reformulated as

$$\begin{array}{ll} \text{minimize}_{x \in \mathbb{R}^n} & x^T D x \quad \text{s.t. } x \geq 0 \end{array}] \quad \begin{array}{l} \text{Back to} \\ \text{the} \\ \text{positivity} \\ \text{problem} \end{array}$$

→ From here, conclude that unconstrained nonconvex problems are NP-hard

Takeaway

- For general nonconvex optimization problems, checking that a point is a local minimum is NP-hard
- Tool: Matrix positivity problems

(3) Polynomial problems

(nonconvex)

→ Idea: What if we restrict ourselves to polynomial functions?

Motivation: Taylor expansion

$$f(x) \Rightarrow f(x+s) \approx f(x) + \nabla f(x)^T s \quad \begin{array}{l} 1^{\text{st}} \text{ order} \\ \text{polynomial} \end{array}$$

$$f(x) \Rightarrow f(x+s) \approx f(x) + \nabla f(x)^T s + \frac{1}{2} s^T \nabla^2 f(x) s \quad \begin{array}{l} 2^{\text{nd}} \text{ order} \\ \text{polynomial} \end{array}$$

$$f \in \mathcal{C}^p \Rightarrow f(x+s) \approx f(x) + \nabla f(x)^T s + \dots + \underbrace{\nabla^p f(x)}_{\text{pth order derivative}} [s, s, \dots, s] \underbrace{s}_{\text{p times}}$$

Decision problems

- ① Given a nonconvex polynomial f and a point $x \in \mathbb{R}^n$,
 is x a { local minimum of f ?
 second-order stationary point?
 first-order stationary point? }
- ② Given a nonconvex polynomial f ,
 does f possess { a local minimum?
 a 2nd order - stationary point?
 a 1st order - stationary point? }

Answers (Ahmadi & Zhang 2022)

- For ① (f, x)
- x 1st order stationary point $\in P$
 - x 2nd order $\overline{\quad}$ $\in P$
 - x local minimum
 $\in NP$ -hard if degree of $f \geq 4$
 (Nurty & Kabadi)
 - $\in P$ if degree ≤ 2
 - (Ahmadi & Zhang) $\in P$ if degree = 3

For (2) f, . . . f has a 1st order stationary point
 $\in P$ if degree of f ≤ 2
 $\in \text{NP-hard}$ otherwise

f has a 2nd order stationary point
 $\stackrel{\text{local minimum}}{\in P}$ if degree ≤ 2
 $\in \text{NP-hard}$ if degree ≥ 3
 $\in \text{"SDP"}$ if degree = 3
 (Can find the answer by solving
 a finite number of SDPs)

Proof sketches

(a) If f: $\mathbb{R}^m \rightarrow \mathbb{R}$ is a polynomial of degree d ≥ 3 , then it is
 strongly NP-hard to decide whether f has a first-order
 stationary point

→ Reduction to MAXCUT

MAXCUT: G undirected, unweighted graph with m vertices
Goal: Find a cut of size k

Let A be the adjacency matrix of G.

$$A \in \mathbb{R}^{m \times m}$$

G has a cut of size k \iff

$$\begin{cases} q_0(x) = \frac{1}{n} \sum_{i=1}^m \sum_{j=1}^m A_{ij} (1 - x_i x_j) - k = 0 \\ q_i(x) = x_i^2 - 1 = 0 \quad i = 1 \dots m \end{cases}$$

has a solution

Define $P(x_1, y_1, z) = \sum_{i=0}^m y_i q_i(x) + z^d$

P has degree d and $2m+2$ variables.

P has a 1st order critical point

$$\Leftrightarrow \begin{cases} q_0(x) = 0 \\ q_i(u) = 0 \quad i=1..m \end{cases} \text{ has a solution}$$

$$D_P(\bar{x}, \bar{y}, \bar{z}) = 0 \Leftrightarrow \left\{ \begin{array}{l} \frac{\partial P}{\partial x_1}(\bar{x}, \bar{y}, \bar{z}) = 0 \\ \vdots \\ \frac{\partial P}{\partial x_m}(\bar{x}, \bar{y}, \bar{z}) = 0 \\ \frac{\partial P}{\partial y_0}(\bar{x}, \bar{y}, \bar{z}) = q_0(\bar{x}) = 0 \\ \vdots \\ \frac{\partial P}{\partial y_i}(\bar{x}, \bar{y}, \bar{z}) = q_i(\bar{x}) = 0 \\ \vdots \\ \frac{\partial P}{\partial z}(\bar{x}, \bar{y}, \bar{z}) = 0 \end{array} \right.$$

- (b) If f is a polynomial with degree $d \geq 4$, it is strongly NP-hard whether f has a 2nd order stationary point.

→ Again reduction to MAXCUT
(with the same notation as before)

Polynomial

$$p(x, y, z, w) = \sum_{i=0}^m \left(y_i q_i(x) - z_i q_i(x) \right) + w^d$$

degree d

2nd order condition $\nabla^2 p(x, y, z, w) \succeq 0$

$$\frac{\partial^2}{\partial y^2} p(x, y, z, w)$$

only holds if

$$2 \begin{bmatrix} 0 & q_0(x) \\ 0 & q_m(x) \end{bmatrix} = \nabla_{yy}^2 p(x, y, z, w) \succeq 0$$

$$2 \begin{bmatrix} -q_0(x) & 0 \\ 0 & -q_m(x) \end{bmatrix} = \nabla_{zz}^2 p(x, y, z, w) \succeq 0$$

p has a 2nd order stationary point $\Rightarrow \begin{cases} q_0(x) = 0 \\ q_i(x) = 0 \quad i=1..m \end{cases}$
has a solution.

(\Leftarrow) If $q_0(w) = \dots = q_m(w) = 0$,

then $(x, 0_{\mathbb{R}^{m+1}}, 0_{\mathbb{R}^{m+1}}, 0)$ is a
second-order stationary point for p .

c) For polynomial of degree 4, it is strongly NP-hard
to decide whether it has a local minimum

→ Reduction to the stable set problem

Given a graph with m vertices
Find a set of pairwise non-adjacent
vertices

G has a stable of size n

(\Rightarrow the polynomial $PA_{n,n-0.5}$ has no local minima
where A is the adjacency matrix of G)

and

$$H \succ 0, \quad PA_{n,n}P(x) = \begin{bmatrix} x_1^2 \\ 1 \\ x_m^2 \end{bmatrix}^T (PA + I - J) \begin{bmatrix} x_1^2 \\ 1 \\ x_m^2 \end{bmatrix}$$

Matrix of all ones

\rightarrow Haderer results for quartic polynomials

\rightarrow Cubic polynomials are a special case

④ Cubic polynomials

$p: \mathbb{R}^n \rightarrow \mathbb{R}$ polynomial of degree 3

p is C^3 and we know that

$$\bar{x} \text{ local minimum of } p \Rightarrow \underbrace{\begin{cases} \|\nabla p(\bar{x})\| = 0 \\ \nabla^2 p(\bar{x}) \succeq 0 \end{cases}}_{\in P}$$

Existing result (applies to every C^3 function)

\bar{x} local minimum of $p \Rightarrow$

third-order stationary point

$$\left. \begin{array}{l} \|\nabla p(\bar{x})\| = 0 \\ \nabla^2 p(\bar{x}) \succeq 0 \\ \nabla^3 p(\bar{x}) [d, d, d] = 0 \end{array} \right\} \text{if } d \text{ such that } \nabla^3 p(\bar{x}) d = 0$$

① Third-order stationarity does not characterize local minima

The 3rd order derivative vanishes in the null space of the Hessian

$$\text{Ex) } n=2, p(x) = x_2^2 - x_1^2 x_2$$

$$\bar{x} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \nabla p(\bar{x}) = \begin{pmatrix} -2x_1 x_2 \\ 2x_2 - x_1^2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\nabla^2 p(\bar{x}) = \begin{pmatrix} -2x_2 & -2x_1 \\ -2x_1 & 2 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 2 \end{pmatrix}$$

$$p\left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}\right) = 0$$

$$p\left(\begin{pmatrix} 1 \\ 1/2 \end{pmatrix}\right) = \frac{1}{4} - \frac{1}{2} = -\frac{1}{4}$$

$$\underbrace{\nabla^3 p(\bar{x})(d, d, d)}_{\nabla^3 p(\bar{x}) = 0} \quad \forall d, \underbrace{\nabla^1 p(\bar{x})d = 0}_{d = \begin{pmatrix} \alpha \\ 0 \end{pmatrix}}$$

Key result

\bar{x} is a Local minimum of p

$$\Leftrightarrow \left\{ \begin{array}{l} \|\nabla p(\bar{x})\| = 0 \\ \nabla^2 p(\bar{x}) \succeq 0 \\ \nabla p_3(\bar{x}) = 0 \quad \forall d \text{ such that } \nabla^3 p(\bar{x})d = 0 \end{array} \right.$$

p_3 is the homogeneous part of degree 3 of p

$$\text{Ex) } n=3, p(u) = x_1^3 + x_2^2 x_3 + x_1 x_2 + 2x_3$$

$$p_3(x) = x_1^3 + x_2^2 x_3$$

As a result, can check that a point is a local minimum of p in polynomial time

(in particular because you can compute a rational basis of the null space of $\nabla^2 p(x)$ when p has integer coefficients in polynomial time)

Procedure

- Check $\nabla_p(x) = 0, \nabla^2 p(x) \geq 0$
- Compute (v_1, \dots, v_l) basis for null($\nabla^2 p(x)$)
- Check if $g(\lambda) = \Delta_{P_3} \left(\sum_{i=1}^l \lambda_i v_i \right)$ has zero coefficients.

For the example

$$P_3(x) = -x_1^2 x_2$$

$$\nabla P_3(x) = \begin{bmatrix} -2x_1 x_2 \\ -x_1^2 \end{bmatrix} =$$

$$\text{null } (\nabla^2 P_3(0)) = \text{vect}(\{0\})$$

$$\nabla P_3(0) = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$$

→ checking 1st order stationarity / 2nd order stationarity / local optimality is polynomial for cubic polynomials

⇒ Deciding whether a cubic polynomial has a 1st order stationary point is NP-hard

\Rightarrow Deciding whether the polynomial has a 2nd order stationary point or a local minimum can be done via solving SDPs.

For second-order stationary

If the problem is in NP , then so is the SDP feasibility problem. **SDP Feasibility Problem:** Given A_0, \dots, A_m and symmetric matrices $m \times n$ with rational entries, does there exist $x \in \mathbb{R}^m$ such that

$$A_0 + \sum_{i=1}^m x_i A_i \succeq 0$$

One form of SDP:

SDP in standard form

$$\begin{aligned} & \text{minimize} && \text{trace}(CX) \\ & \text{s.t.} && \text{trace}(A_i X) = b_i \quad i=1..m \\ & && X \succeq 0 \end{aligned}$$

C, A_i symmetric matrices

$$b \in \mathbb{R}^m$$

Dual of SDP:

$$\begin{aligned} & \text{maximize} && b^T y \\ & \text{s.t.} && -C + \sum_{i=1}^m y_i A_i \succeq 0 \\ & && S \succeq 0 \end{aligned}$$

\rightarrow Approach: Reformulate the decision problem using a semi-definite program
 (using sum-of-squares polynomials and sum-of-squares matrices)

\Rightarrow If has a 2nd order stationary point
 \Rightarrow A certain SDP has a finite optimal value,
 and that value is the value of p at
 every 2nd order stationary point
 (the set of 2nd order stationary points is
 a convex set)

thj For every twice polynomial $p: \mathbb{R}^m \rightarrow \mathbb{R}$, p can be written as

$$p(x) = c + b^T x + \frac{1}{2} x^T Q x + \frac{1}{6} \sum_{i=1}^m x^T x_i H_i x$$

 for $b \in \mathbb{R}^m$, $Q \in \mathbb{R}^{m \times m}$, $H_i \in \mathbb{R}^{m \times m}$ & i

then the set of second-order stationary points of p is

$$\{x \in \mathbb{R}^m \mid \exists Y \in \mathbb{R}^{m \times m}, Y = Y^T, \exists z \in \mathbb{R} \text{ such that}$$

$$\frac{1}{2} \text{trace}(QY) + b^T x + \frac{3}{2} z = 0$$

$$\frac{1}{2} \text{trace}(H_i Y) + e_i^T Q x + b_i = 0$$

$$\begin{bmatrix} \sum y_i H_i + Q & \sum \text{trace}(H_i Y) e_i + d y \\ \sum \text{trace}(H_i Y) e_i + d y & z \end{bmatrix} = 0$$

$$\begin{bmatrix} Y & z \\ x^T & 1 \end{bmatrix} \succeq 0 \quad \hookrightarrow$$

Summary : → For checking / finding stationary points / local minima,
we have hardness results

⇒ Case of whole polynomials not fully
understood

→ Partly explains the success of blackbox complexity

→ Recent results on complexity of finding
approximate 1st order stationary points

$$\|\nabla f(x)\| \leq \varepsilon \quad \text{where } f \text{ nonconvex } C^{1,1} \\ f: \mathbb{R}^m \rightarrow \mathbb{R} \quad m \geq 2$$

⇒ PLS-complete
Polynomial Local Search