

# MEB and Anomaly Detection

Optimization for Data Science

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#### Abstract

The ABSTRACT is to be in fully-justified italicized text, at the top of the left-hand column, below the author and affiliation information. Use the word "Abstract" as the title, in 12-point Times, boldface type, centered relative to the column, initially capitalized. The abstract is to be in 10-point, single-spaced type. Leave two blank lines after the Abstract, then begin the main text. Abstract should be no longer than 300 words.

### 1 Introduction

# 2 Minimum Enclosing Balls

[Here is where I'd do the transition between the formulation of the problem in the introduction and how we intend to solve it using MEB]

#### 2.1 Description of MEB

We can describe the Minimum Enclosing Ball problem in the following way: given a set of points  $\mathcal{A} = a_1, ..., a_n$  in d dimensions, the objective is to find the smallest ball  $B^n(c, \rho)$  containing all points of  $\mathcal{A}$ , such that:

$$\mathcal{A} \subseteq B^n(c,\rho) \tag{1}$$

where

$$B^{n}(c,\rho) = \{ x \in \mathbb{R}^{n} \mid ||x - c|| \le \rho \}$$
 (2)

where  $c \in \mathbb{R}^n$  and  $\rho \in \mathbb{R}$  are the decision variables. We define c as the centre of the minimum enclosing ball and  $\rho$  as the radius. A sphere  $B^n(c,\rho) \subset \mathbb{R}^n$  is called an enclosing ball of  $\mathcal{A}$  if and only if (1) holds.

The problem of calculating an approximation to the minimum enclosing ball of  $\mathcal{A}$  can be denoted as MEB( $\mathcal{A}$ ), which can be computed by solving the following optimization problem:

$$(\mathcal{P}_1) \quad \min_{c,\rho} \rho$$
 subject to  $a_i - c \leq \rho, \quad i = 1, \dots, m$ 

We obtain a second optimization problem  $\mathcal{P}_2$  with smooth and convex quadratic constraints, by setting  $\gamma := \rho^2$  and squaring the constraints of  $\mathcal{P}_1$ .

$$(\mathcal{P}_2) \quad \min_{c,\gamma} \gamma$$
subject to  $(a_i)^T a_i - 2(a_i)^T c + c^T c \le \gamma, \quad i = 1, \dots, m$ 

We can derive the lagrangian dual from  $\mathcal{P}_2$  by adding Lagrange multiplier to each constraint.

## 2.2 References

List and number all bibliographical references in 9-point Times, single-spaced, at the end of your paper. When referenced in the text, enclose the citation number in square brackets, for example [1]. Where appropriate, include the name(s) of editors of referenced books.

# References

[1] Authors. The frobnicatable foo filter, 2014. Face and Gesture submission ID 324. Supplied as additional material fg324.pdf.