

# Computing Standard-Deviation-to-Mean and Variance-to-Mean Ratios under Interval Uncertainty is NP-Hard

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A Practical Problem: ...

How This Problem is ...

A Standard Way to ...

Selecting the Parameter  $k$

Case of Interval ...

What is Known

What We Do in This Talk

Theorem 1

Theorem 2

Home Page

Title Page

«

»

◀

▶

Page 1 of 31

Go Back

Full Screen

Close

Quit

## 1. A Practical Problem: Checking Whether an Object Belongs to a Class

- In many practical situations, we want to check whether a new object belongs to a given class.
- In such situations, we usually have a sample of objects which are known to belong to this class.
- *Example:*
  - a biologist who is studying bats has observed several bats from a local species;
  - the question is
    - \* whether a newly observed bat belongs to the same species –
    - \* or whether the newly observed bat belongs to a different bat species.

## 2. How This Problem is Usually Solved

- *Problem:* checking whether an object belongs to a class.
- *To solve this problem:* we usually
  - measure one or more quantities for the objects from this class and for the new object, and
  - compare the resulting values.
- *Simplest case of a single quantity.* In this case, we have:
  - a collection of values  $x_1, \dots, x_n$  corresponding to objects from the known class, and
  - a value  $x$  corresponding to the new object.

### 3. A Standard Way to Decide Whether an Object Belongs to a Class

- *Problem* (reminder): to decide whether
  - a new object with the value  $x$
  - belongs to the class characterized by the values  $x_1, \dots, x_n$ .
- *Usual solution*: check whether the value  $x$  belongs to the “ $k$  sigma” interval  $[E - k \cdot \sigma, E + k \cdot \sigma]$ , where:
  - $E \stackrel{\text{def}}{=} \frac{1}{n} \cdot \sum_{i=1}^n x_i$  is the sample mean,
  - $\sigma = \sqrt{V}$ , where  $V \stackrel{\text{def}}{=} \frac{1}{n} \cdot \sum_{i=1}^n (x_i - E)^2$  is the sample variance, and
  - the parameter  $k$  is determined by the degree of confidence with which we want to make the decision.

## 4. Selecting the Parameter $k$

- *Problem* (reminder): to decide whether
  - a new object with the value  $x$
  - belongs to the class characterized by the values  $x_1, \dots, x_n$ .
- *Solution* (reminder): check whether the value  $x$  belongs to the “ $k$  sigma” interval  $[E - k \cdot \sigma, E + k \cdot \sigma]$
- Usually, we take:
  - $k = 2$  (corresponding to confidence 0.9),
  - $k = 3$  (corresponding to 0.999), or
  - $k = 6$  (corresponding to  $1 - 10^{-8}$ ).

A Practical Problem: ...

How This Problem is ...

A Standard Way to ...

Selecting the Parameter  $k$

Case of Interval ...

What is Known

What We Do in This Talk

Theorem 1

Theorem 2

Home Page

Title Page

◀◀

▶▶

◀

▶

Page 5 of 31

Go Back

Full Screen

Close

Quit

## 5. Formulation of the Problem

- *Problem*: checking whether an object belongs to a class.
- *Standard approach*: an object belongs to the class if the value  $x$  belongs to the “ $k$  sigma” interval

$[E - k \cdot \sigma, E + k \cdot \sigma]$ , where:

- $E \stackrel{\text{def}}{=} \frac{1}{n} \cdot \sum_{i=1}^n x_i$  is the sample mean, and
- $\sigma = \sqrt{V}$ , where  $V \stackrel{\text{def}}{=} \frac{1}{n} \cdot \sum_{i=1}^n (x_i - E)^2$  is the sample variance
- How confident are we about the decision
  - depends on the smallest values  $k^-$  for which  $x \geq E - k^- \cdot \sigma$ , and
  - on the smallest value  $k^+$  for which  $x \leq E + k^+ \cdot \sigma$ .

## 6. How to Compute the Parameters Describing Confidence?

- The inequality  $x \geq E - k^- \cdot \sigma$  is equivalent to  $k^- \cdot \sigma \geq E - x$  and  $k^- \geq \frac{E - x}{\sigma}$ .
- Thus, when  $x < E$ , the corresponding smallest value is equal to  $k^- = \frac{E - x}{\sigma}$ .
- Similarly, the inequality  $x \leq E + k^+ \cdot \sigma$  is equivalent to  $k^+ \cdot \sigma \geq x - E$  and  $k^+ \geq \frac{x - E}{\sigma}$ .
- Thus, when  $x > E$ , the corresponding smallest value is equal to  $k^+ = \frac{x - E}{\sigma}$ .
- So, to determine the parameter describing confidence, we must compute one of the ratios

$$k^- \stackrel{\text{def}}{=} \frac{E - x}{\sigma} \text{ or } k^+ \stackrel{\text{def}}{=} \frac{x - E}{\sigma}.$$

## 7. How to Compute the Parameters Describing Confidence (cont-d)

- *Reminder:* to determine the parameter describing confidence, we must compute one of the ratios

$$k^- \stackrel{\text{def}}{=} \frac{E - x}{\sigma} \text{ or } k^+ \stackrel{\text{def}}{=} \frac{x - E}{\sigma}.$$

- Often, reciprocal ratio are used:

$$r^- \stackrel{\text{def}}{=} \frac{1}{k^-} = \frac{\sigma}{E - x} \text{ and } r^+ \stackrel{\text{def}}{=} \frac{1}{k^+} = \frac{\sigma}{x - E}.$$

A Practical Problem: ...

How This Problem is ...

A Standard Way to ...

Selecting the Parameter  $k$

Case of Interval ...

What is Known

What We Do in This Talk

Theorem 1

Theorem 2

Home Page

Title Page

◀◀

▶▶

◀

▶

Page 8 of 31

Go Back

Full Screen

Close

Quit



## 8. Case of Interval Uncertainty

- *Simplifying assumption*: we know the exact values  $x_1, \dots, x_n$  of the corresponding quantity.
- *In practice*: these values come from measurement, and measurements are never absolutely accurate.
- *Specifics*: the measurement results  $\tilde{x}_1, \dots, \tilde{x}_n$  are, in general, different from the actual (unknown) values  $x_i$ .
- *Traditional engineering techniques* assume that we know the probabilities of different values of  $\Delta x_i \stackrel{\text{def}}{=} \tilde{x}_i - x_i$ .
- *In many practical situations*: we only know the upper bound  $\Delta_i$ :  $|\Delta x_i| \leq \Delta_i$ .
- *In this case*: we only know that the actual (unknown) value  $x_i$  belongs to the interval  $\mathbf{x}_i \stackrel{\text{def}}{=} [\tilde{x}_i - \Delta_i, \tilde{x}_i + \Delta_i]$ .

## 9. Case of Interval Uncertainty (cont-d)

- *Reminder:* we only know that the actual (unknown) value  $x_i$  belongs to the interval  $\mathbf{x}_i \stackrel{\text{def}}{=} [\tilde{x}_i - \Delta_i, \tilde{x}_i + \Delta_i]$ .
- Different possible values  $x_i \in \mathbf{x}_i$  lead, in general, to different values of the corresponding ratios  $r(x_1, \dots, x_n)$ .
- Thus, it is desirable to compute the *range* of possible values of this ratio:

$$\mathbf{r} = [\underline{r}, \bar{r}] \stackrel{\text{def}}{=} \{r(x_1, \dots, x_n) \mid x_1 \in \mathbf{x}_1, \dots, x_n \in \mathbf{x}_n\}.$$

- This problem is a particular case of the main problem of *interval computation*:
  - *given:* an algorithm  $f(x_1, \dots, x_n)$  and intervals  $\mathbf{x}_i$ ,
  - *compute:* the range

$$\mathbf{y} = [\underline{y}, \bar{y}] \stackrel{\text{def}}{=} \{f(x_1, \dots, x_n) \mid x_1 \in \mathbf{x}_1, \dots, x_n \in \mathbf{x}_n\}.$$

## 10. What is Known

- What was analyzed earlier:
  - the problem of computing the range of  $r$ , and
  - similar problems of computing ranges for the thresholds  $E - k \cdot \sigma$  and  $E + k \cdot \sigma$  for a given  $k$ .
- Feasible algorithms were described:
  - for computing the upper bounds for  $E - k \cdot \sigma$ , and
  - for computing the lower bounds for  $E + k \cdot \sigma$ ,  $\frac{\sigma}{E - x}$ , and  $\frac{\sigma}{x - E}$ .
- For other bounds, feasible algorithms are known under certain conditions on the intervals:
  - for computing the lower bounds for  $E - k \cdot \sigma$ , and
  - for computing the upper bounds for  $E + k \cdot \sigma$ ,  $\frac{\sigma}{E - x}$ , and  $\frac{\sigma}{x - E}$ .

## 11. What is Known (cont-d)

- *Reminder:* for some bounds, feasible algorithms are known only under conditions on intervals:
  - for computing the lower bounds for  $E - k \cdot \sigma$ , and
  - for computing the upper bounds for  $E + k \cdot \sigma$ ,  $\frac{\sigma}{E - x}$ , and  $\frac{\sigma}{x - E}$ .
- It was proven that for  $E \pm k \cdot \sigma$ , such conditions are necessary.
- Specifically, it was proven that the following problems are NP-hard:
  - computing the lower bound for  $E - k \cdot \sigma$  and
  - computing the upper bound for  $E + k \cdot \sigma$ .
- This means that, unless  $P=NP$ , these problems cannot be, in general, solved in polynomial (= feasible) time.

A Practical Problem: ...

How This Problem is ...

A Standard Way to ...

Selecting the Parameter  $k$

Case of Interval ...

What is Known

What We Do in This Talk

Theorem 1

Theorem 2

Home Page

Title Page

◀◀

▶▶

◀

▶

Page 12 of 31

Go Back

Full Screen

Close

Quit

## 12. What We Do in This Talk

- *It was known:* that computing bounds for the thresholds  $E - k \cdot \sigma$  and  $E + k \cdot \sigma$ .

- *Another problem:* computing the range  $\mathbf{r}$  of the ratio

$$r = \frac{\sigma}{E - x}.$$

- *It was not known:* whether computing  $\mathbf{r}$  is NP-hard.
- *In this talk:* we prove that the problem of computing  $\mathbf{r}$  is NP-hard.

- *A similar problem:* computing the range  $\mathbf{R}$  of a ratio

$$R \stackrel{\text{def}}{=} \frac{V}{E}.$$

- *Comment:* the ration  $R$  is used in clustering.
- *We prove:* that the problem of computing  $\mathbf{R}$  is also NP-hard.

A Practical Problem: ...

How This Problem is ...

A Standard Way to ...

Selecting the Parameter  $k$

Case of Interval ...

What is Known

What We Do in This Talk

Theorem 1

Theorem 2

Home Page

Title Page

◀◀

▶▶

◀

▶

Page 13 of 31

Go Back

Full Screen

Close

Quit

## 13. Discussion

- *Known fact:*
  - to prove that a problem is NP-hard,
  - it is sufficient to prove that a particular case of this problem is NP-hard.
- $\widehat{\text{We want:}}$  to prove that computing the upper bound of the ratios  $\frac{\sigma}{E - x}$  and  $\frac{\sigma}{x - E}$  is NP-hard.
- *It is sufficient to prove:* that computing the range of the ratio  $\frac{\sigma}{E}$  (corr. to  $x = 0$ ) is NP-hard.
- *It is sufficient to prove:* for the case when all the intervals  $[\underline{x}_i, \bar{x}_i]$  contain only non-negative values.
- *Comment:* this case is equivalent to  $\underline{x}_i \geq 0$  for all  $i$ .

## 14. Theorem 1

*The following problem is NP-hard:*

- given: a natural number  $n$  and  $n$  (rational-valued) intervals  $[\underline{x}_i, \bar{x}_i]$ ,
- compute: the upper endpoint  $\bar{r}$  of the range

$$\mathbf{r} = [\underline{r}, \bar{r}] = \{r(x_1, \dots, x_n) \mid x_1 \in [\underline{x}_1, \bar{x}_1], \dots, x_n \in [\underline{x}_n, \bar{x}_n]\}$$

of the ratio  $r = \frac{\sqrt{V}}{E}$ , where  $E = \frac{1}{n} \cdot \sum_{i=1}^n x_i$  and

$$V = \frac{1}{n} \cdot \sum_{i=1}^n (x_i - E)^2.$$

## 15. Theorem 2

The following problem is NP-hard:

- given: a natural number  $n$  and  $n$  (rational-valued) intervals  $[\underline{x}_i, \bar{x}_i]$ ,
- compute: the upper endpoint  $\bar{r}$  of the range

$$\mathbf{r} = [\underline{r}, \bar{r}] = \{r(x_1, \dots, x_n) \mid x_1 \in [\underline{x}_1, \bar{x}_1], \dots, x_n \in [\underline{x}_n, \bar{x}_n]\}$$

of the ratio  $r = \frac{V}{E}$ , where  $E = \frac{1}{n} \cdot \sum_{i=1}^n x_i$  and

$$V = \frac{1}{n} \cdot \sum_{i=1}^n (x_i - E)^2.$$



## 16. Proof of Theorem 1: Eliminating Square Root

- The expression for the ratio  $r = \frac{\sqrt{V}}{E}$  uses a square root – to compute  $\sigma = \sqrt{V}$ .
- In optimization, we usually use derivatives.
- The square root function  $f(x) = \sqrt{x}$  has infinite derivative when  $x = 0$ .
- Thus, it is desirable to avoid square roots.
- To avoid the square root problem, we can use the facts that
  - $r = \sqrt{R}$ , where  $R \stackrel{\text{def}}{=} \frac{V}{E^2}$ , and
  - the function  $\sqrt{x}$  is strictly increasing.

A Practical Problem: ...

How This Problem is ...

A Standard Way to ...

Selecting the Parameter  $k$

Case of Interval ...

What is Known

What We Do in This Talk

Theorem 1

Theorem 2

Home Page

Title Page

◀◀

▶▶

◀

▶

Page 17 of 31

Go Back

Full Screen

Close

Quit

## 17. Eliminating Square Root (cont-d)

- Since the function  $\sqrt{x}$  is strictly increasing:
  - the smallest possible value  $\underline{r}$  of  $r$  is equal to the square root of the smallest possible value of  $R$ :

$$\underline{r} = \sqrt{\underline{R}};$$

- the largest possible value  $\bar{r}$  of  $r$  is equal to the square root of the largest possible value of  $R$ :

$$\bar{r} = \sqrt{\bar{R}}.$$

- So:
  - the problem of computing the range of the ratio  $r$  is feasibly equivalent to
  - the problem of computing the range  $[\underline{R}, \bar{R}]$  of the new ratio  $R$ .
- Thus, computing  $\mathbf{r}$  is NP-hard  $\Leftrightarrow$  computing  $\mathbf{R}$  is NP-hard.

## 18. Proof: Part 2

- A problem is NP-hard if every problem from a certain class NP can be reduced to it.
- We will show that a known NP-hard problem  $\mathcal{P}$  can be reduced to our problem  $\mathcal{P}_0$ . Then:
  - every problem from the class NP can be reduced to  $\mathcal{P}$ , and
  - $\mathcal{P}$  can be reduced to  $\mathcal{P}_0$ ,
  - hence every problem from the class NP can also be reduced to  $\mathcal{P}_0$ ;
  - thus, our problem  $\mathcal{P}_0$  is indeed NP-hard.
- As  $\mathcal{P}$ , we choose a *subset sum* problem:
  - given  $n$  positive integers  $s_1, \dots, s_n$ ,
  - check whether there exists signs  $\eta_i \in \{-1, 1\}$  for which  $\sum_{i=1}^n \eta_i \cdot s_i = 0$ .

## 19. Part 2 (cont-d)

- *Reminder:* we reduce the problem of computing the range  $\mathbf{R} = [\underline{R}, \overline{R}]$  to the *subset sum* problem:
  - given  $n$  positive integers  $s_1, \dots, s_n$ ,
  - check whether there exists signs  $\eta_i \in \{-1, 1\}$  for which  $\sum_{i=1}^n \eta_i \cdot s_i = 0$ .
- Specifically, we will prove that for an appropriately chosen integer  $N$ :
  - such signs exist if and only if
  - for the intervals  $\mathbf{x}_i = [N - s_i, N + s_i]$ , the upper endpoint  $\overline{R}$  is greater than or equal to

$$R_0 \stackrel{\text{def}}{=} \frac{M_0}{N^2}, \text{ where } M_0 \stackrel{\text{def}}{=} \frac{1}{n} \cdot \sum_{i=1}^n s_i^2.$$

A Practical Problem: ...

How This Problem is ...

A Standard Way to ...

Selecting the Parameter  $k$

Case of Interval ...

What is Known

What We Do in This Talk

Theorem 1

Theorem 2

Home Page

Title Page

◀◀

▶▶

◀

▶

Page 20 of 31

Go Back

Full Screen

Close

Quit

## 20. Part 3

- *Lemma.* The ratio  $R = \frac{V}{E^2}$  attains its maximum on the box

$$[\underline{x}_1, \bar{x}_1] \times \dots \times [\underline{x}_n, \bar{x}_n]$$

when each of the variables  $x_i$  is equal to one of the endpoints  $\underline{x}_i$  or  $\bar{x}_i$ .

- We will prove this statement by contradiction.
- Let us assume that for some  $i$ , the function  $R$  attains its maximum at an internal point  $x_i \in (\underline{x}_i, \bar{x}_i)$ .
- In this case, according to calculus, at this point,

- the partial derivative  $\frac{\partial R}{\partial x_i}$  should be equal to 0, and
- the second derivative  $\frac{\partial^2 R}{\partial x_i^2}$  should be non-positive.

## 21. Part 3 (cont-d)

- Here,

$$\frac{\partial E}{\partial x_i} = \frac{\partial}{\partial x_i} \left( \frac{1}{n} \cdot \sum_{j=1}^n x_j \right) = \frac{1}{n}.$$

- Since  $V = M - E^2$ , where  $M \stackrel{\text{def}}{=} \frac{1}{n} \cdot \sum_{j=1}^n x_j^2$ , we have

$$\frac{\partial V}{\partial x_i} = \frac{\partial M}{\partial x_i} - \frac{\partial E^2}{\partial x_i}.$$

- Here,

$$\frac{\partial E^2}{\partial x_i} = 2 \cdot E \cdot \frac{\partial E}{\partial x_i} = 2 \cdot E \cdot \frac{1}{n},$$

and

$$\frac{\partial M}{\partial x_i} = \frac{\partial}{\partial x_i} \left( \frac{1}{n} \cdot \sum_{j=1}^n x_j^2 \right) = \frac{1}{n} \cdot 2x_i.$$

A Practical Problem: ...

How This Problem is ...

A Standard Way to ...

Selecting the Parameter  $k$

Case of Interval ...

What is Known

What We Do in This Talk

Theorem 1

Theorem 2

Home Page

Title Page

◀◀

▶▶

◀

▶

Page 22 of 31

Go Back

Full Screen

Close

Quit

## 22. Computing the First Derivative

- So, for  $V = M - E^2$ , we have  $\frac{\partial V}{\partial x_i} = \frac{1}{n} \cdot 2x_i - 2 \cdot E \cdot \frac{1}{n}$ .
- Thus,

$$\begin{aligned}\frac{\partial R}{\partial x_i} &= \frac{\partial}{\partial x_i} \left( \frac{V}{E^2} \right) = \frac{\frac{\partial V}{\partial x_i} \cdot E^2 - V \cdot \frac{\partial E^2}{\partial x_i}}{E^4} = \\ &= \frac{\left( \frac{1}{n} \cdot 2x_i - 2 \cdot E \cdot \frac{1}{n} \right) \cdot E^2 - V \cdot 2 \cdot E \cdot \frac{1}{n}}{E^4} = 2 \cdot \frac{x_i \cdot E - E^2 - V}{n \cdot E^3}.\end{aligned}$$

- So, when  $\frac{\partial R}{\partial x_i} = 0$ , we get

$$x_i = \frac{E^2 + V}{E} = \frac{M}{E}.$$

[A Practical Problem: ...](#)[How This Problem is ...](#)[A Standard Way to ...](#)[Selecting the Parameter  \$k\$](#) [Case of Interval ...](#)[What is Known](#)[What We Do in This Talk](#)[Theorem 1](#)[Theorem 2](#)[Home Page](#)[Title Page](#)[<<](#)[>>](#)[<](#)[>](#)[Page 23 of 31](#)[Go Back](#)[Full Screen](#)[Close](#)[Quit](#)

## 23. Computing the Second Derivative

- We differentiate  $\frac{\partial R}{\partial x_i}$  with respect to  $x_i$ .
- As a result, we get the following expression for the second derivative:

$$\frac{\partial^2 R}{\partial x_i^2} = 2 \cdot \frac{3 \cdot V + (n + 3) \cdot E^2 - 4 \cdot x_i \cdot E}{n^2 \cdot E^4}.$$

- The denominator is positive.
- We assumed that the second derivative is non-positive.
- We thus conclude that

$$3 \cdot V + (n + 3) \cdot E^2 - 4 \cdot x_i \cdot E =$$

$$3 \cdot M + n \cdot E^2 - 4 \cdot x_i \cdot E \leq 0.$$

A Practical Problem: ...

How This Problem is ...

A Standard Way to ...

Selecting the Parameter  $k$

Case of Interval ...

What is Known

What We Do in This Talk

Theorem 1

Theorem 2

Home Page

Title Page



Page 24 of 31

Go Back

Full Screen

Close

Quit



## 24. Proof of the Lemma

- *Reminder:*  $3 \cdot M + n \cdot E^2 - 4 \cdot x_i \cdot E \leq 0$ .
- By definition,  $E = \frac{1}{n} \cdot \sum_{j=1}^n x_j = \frac{1}{n} \cdot x_i + \frac{1}{n} \cdot E_i$ , where we denoted  $E_i \stackrel{\text{def}}{=} \sum_{j \neq i} x_j$ .
- Similarly,  $M = \frac{1}{n} \cdot x_i^2 + \frac{1}{n} \cdot M_i$  where  $M_i \stackrel{\text{def}}{=} \sum_{j \neq i} x_j^2$ .
- So, we get  $3 \cdot M_i + E_i^2 - 2 \cdot x_i \cdot E_i \leq 0$ .
- Due to  $x_i = \frac{M}{E}$ , we have  $x_i \cdot E = M$ , hence

$$x_i \cdot \left( \frac{1}{n} \cdot x_i + \frac{1}{n} \cdot E_i \right) = \frac{1}{n} \cdot x_i^2 + \frac{1}{n} \cdot M_i.$$

- We also get  $x_i \cdot E_i = M_i$ , so we conclude that

$$3 \cdot M_i + E_i^2 - 2 \cdot x_i \cdot E_i = M_i + E_i^2 \leq 0.$$

## 25. Proof of the Lemma

- *Reminder:* we proved that  $M_i + E_i^2 \leq 0$ .
- However, for large enough  $N$  – specifically, for  $N > \max_i s_i$  – we have  $x_j > 0$ .
- Hence  $M_i = \frac{1}{n} \cdot \sum_{j \neq i} x_j^2 > 0$  and thus,

$$M_i + E_i^2 > 0.$$

- This shows that the maximum of  $R$  cannot be attained at an internal point of the interval  $(\underline{x}_i, \bar{x}_i)$ .
- Thus, this maximum can only be attained when  $x_i = \underline{x}_i$  or  $x_i = \bar{x}_i$ .
- The Lemma is proven.

## 26. Proof of Theorem 1

- *Statement:*  $\bar{R} \geq R_0 = \frac{M_0}{N^2} \Leftrightarrow$  there exist signs  $\eta_i \in \{-1, 1\}$  for which  $\sum_{i=1}^n \eta_i \cdot s_i = 0$ .

$\Leftrightarrow$  If such signs exist, then we take  $x_i = N + \eta_i \cdot s_i$ ; then:

- $E = N$ ,  $x_i - E = \pm s_i$ , and
  - $V = \frac{1}{n} \cdot \sum_{i=1}^n (x_i - E)^2 = \frac{1}{n} \cdot \sum_{i=1}^n s_i^2 = M_0$ , and
  - $R = \frac{V}{E^2} = \frac{M_0}{N^2} = R_0$ .
- The largest possible value  $\bar{R}$  must therefore be larger than or equal to this value.

A Practical Problem: ...

How This Problem is ...

A Standard Way to ...

Selecting the Parameter  $k$

Case of Interval ...

What is Known

What We Do in This Talk

Theorem 1

Theorem 2

Home Page

Title Page



Page 27 of 31

Go Back

Full Screen

Close

Quit

## 27. Proof of Theorem 1 (cont-d)

- Vice versa, assume that  $\overline{R} \geq R_0$ .
- Let  $x_i$  be the values for which the ratio  $R$  attains its maximum value  $\overline{R}$ .
- Due to the Lemma, this maximum is attained when  $x_i = N + t_i$  with  $t_i = \eta_i \cdot s_i$  and  $\eta_i \in \{-1, 1\}$ ; then:

- $E = N + e$ , where  $e \stackrel{\text{def}}{=} \frac{1}{n} \cdot \sum_{i=1}^n t_i$ , and

- $V(x_1, \dots, x_n) = V(t_1, \dots, t_n) = \frac{1}{n} \cdot \sum_{i=1}^n t_i^2 - e^2$ .

- Since  $t_i = \pm s_i$ , we have  $t_i^2 = s_i^2$  and thus,  $\frac{1}{n} \cdot \sum_{i=1}^n t_i^2 =$

$$\frac{1}{n} \cdot \sum_{i=1}^n s_i^2 = M_0 \text{ and } V = M_0 - e^2; \text{ thus, } \overline{R} = \frac{M_0 - e^2}{(N + e)^2}.$$

## 28. Proof of Theorem 1 (cont-d)

- *Reminder:*  $\overline{R} = \frac{M_0 - e^2}{(N + e)^2} \geq R_0$ .
- Multiplying both sides by the denominator, we get

$$e^2 \cdot (N^2 + M_0) + 2 \cdot M_0 \cdot N \cdot e \leq 0.$$

- If  $e > 0$ , then the left-hand side is positive and cannot be  $\leq 0$ , so  $e \leq 0$ .
- If  $e < 0$ , then this inequality leads to

$$|e|^2 \cdot (N^2 + M_0) - 2 \cdot M_0 \cdot N \cdot |e| \leq 0$$

and

$$|e| \leq \frac{2 \cdot M_0 \cdot N}{N^2 + M_0}.$$

A Practical Problem: ...

How This Problem is ...

A Standard Way to ...

Selecting the Parameter  $k$

Case of Interval ...

What is Known

What We Do in This Talk

Theorem 1

Theorem 2

Home Page

Title Page

◀◀

▶▶

◀

▶

Page 29 of 31

Go Back

Full Screen

Close

Quit

## 29. Proof of Theorem 1 (final part)

- *Reminder:*  $|e| \leq \frac{2 \cdot M_0 \cdot N}{N^2 + M_0}$ .
- Since  $\frac{2 \cdot M_0 \cdot N}{N^2 + M_0} \rightarrow 0$  as  $N \rightarrow \infty$ , for sufficiently large  $N$ , we get  $|e| \geq \frac{1}{n} > \frac{2 \cdot M_0 \cdot N}{N^2 + M_0}$ .
- However, by definition, all the values  $s_i$ , and all the values  $t_i = \pm s_i$ , and the sum  $n \cdot e = \sum_{i=1}^n t_i$  are integers.
- So  $|n \cdot e| \geq 1$  and  $|e| \geq \frac{1}{n}$ .
- Thus, the inequality  $|e| \leq \frac{2 \cdot M_0 \cdot N}{N^2 + M_0}$  is impossible.
- This shows that  $e$  cannot be negative, hence  $e = 0$ , and thus,  $n \cdot e = \sum_{i=1}^n \eta_i \cdot s_i = 0$ . The theorem is proven.

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<a href="#">A Standard Way to . . .</a>
<a href="#">Selecting the Parameter <math>k</math></a>
<a href="#">Case of Interval . . .</a>
<a href="#">What is Known</a>
<a href="#">What We Do in This Talk</a>
<a href="#">Theorem 1</a>
<a href="#">Theorem 2</a>

[Home Page](#)

[Title Page](#)



Page 31 of 31

[Go Back](#)

[Full Screen](#)

[Close](#)

[Quit](#)