

Modern Family

Convex Minimization in \mathbb{R}^n

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Informatik 12:
Software and Tools for Computational Engineering (STCE)

RWTH Aachen University

Objective and Learning Outcomes

Least-Squares Objective

Optimality Conditions

Linear Model

- Optimality Conditions
- Implementation

Nonlinear Model

- Optimality Conditions
- Implementation

Summary and Next Steps

Objective and Learning Outcomes

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Summary and Next Steps

Objective

- ▶ Introduction to sample code approaching the Modern Family example as a general convex minimization problem.

Learning Outcomes

- ▶ You will understand
 - ▶ the formulation of the Modern Family example as a general convex minimization problem
 - ▶ its solution using the general-purpose optimization methods steepest descent and Newton's method.
- ▶ You will be able to
 - ▶ run the sample code
 - ▶ compare the results

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Summary and Next Steps

We state the Modern Family example for model

$$y = f(\mathbf{p}, \mathbf{x}) : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$$

as minimization of the error function

$$E(\mathbf{p}, X, \mathbf{y}) = \sum_{i=0}^{m-1} (f(\mathbf{p}, \mathbf{x}_i^T) - y_i)^2$$

Linear (in \mathbf{p})

$$y = \mathbf{p}^T \cdot \mathbf{x}$$

and nonlinear

$$y = (\mathbf{p}^T \cdot \mathbf{x})^2$$

models are considered.

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Necessary

$$\frac{dE}{d\mathbf{p}}(\mathbf{p}, X, \mathbf{y}) = 2 \cdot \sum_{i=0}^{m-1} \left((f(\mathbf{p}, \mathbf{x}_i^T) - y_i) \cdot \frac{df}{d\mathbf{p}}(\mathbf{p}, \mathbf{x}_i^T) \right) \rightarrow 0$$

Sufficient

$$\mathbf{v}^T \cdot \frac{d^2E}{d\mathbf{p}^2}(\mathbf{p}, X, \mathbf{y}) \cdot \mathbf{v} > 0 \quad \forall \mathbf{v} \neq \mathbf{0} \in R^n$$

where

$$\frac{d^2E}{d\mathbf{p}^2} = 2 \cdot \sum_{i=0}^{m-1} \left(\left(\frac{df}{d\mathbf{p}}(\mathbf{p}, \mathbf{x}_i^T) \right)^T \cdot \frac{df}{d\mathbf{p}}(\mathbf{p}, \mathbf{x}_i^T) + \frac{d^2f}{d\mathbf{p}^2}(\mathbf{p}, \mathbf{x}_i^T) \cdot (f(\mathbf{p}, \mathbf{x}_i^T) - y_i) \right)$$

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For $y = f(\mathbf{p}, \mathbf{x}) = \mathbf{p}^T \cdot \mathbf{x} = \mathbf{x}^T \cdot \mathbf{p}$

$$E(\mathbf{p}, X, \mathbf{y}) = \sum_{i=0}^{m-1} (f(\mathbf{p}, \mathbf{x}_i^T) - y_i)^2 = \sum_{i=0}^{m-1} (\mathbf{x}_i \cdot \mathbf{p} - y_i)^2$$

and hence

$$\frac{dE}{d\mathbf{p}}(\mathbf{p}, X, \mathbf{y}) = 2 \cdot \sum_{i=0}^{m-1} (\mathbf{x}_i \cdot \mathbf{p} - y_i) \cdot \mathbf{x}_i \in \mathbf{R}^{1 \times n}$$

and

$$\frac{d^2E}{d\mathbf{p}^2}(\mathbf{p}, X, \mathbf{y}) = 2 \cdot \sum_{i=0}^{m-1} \mathbf{x}_i^T \cdot \mathbf{x}_i \in \mathbf{R}^{n \times n}.$$

We present implementations for the solution of the convex unconstrained minimization problem

$$\min_{\mathbf{p} \in \mathbb{R}^n} E(\mathbf{p}, X, \mathbf{y})$$

using

- ▶ steepest descent
- ▶ Newton's method.

See source code.

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For $y = f(\mathbf{p}, \mathbf{x}) = (\mathbf{p}^T \cdot \mathbf{x})^2 = (\mathbf{x}^T \cdot \mathbf{p})^2$

$$E(\mathbf{p}, X, \mathbf{y}) = \sum_{i=0}^{m-1} (f(\mathbf{p}, \mathbf{x}_i^T) - y_i)^2 = \sum_{i=0}^{m-1} (\mathbf{x}_i \cdot \mathbf{p} - y_i)^2$$

and hence

$$\frac{dE}{d\mathbf{p}}(\mathbf{p}, X, \mathbf{y}) = 4 \cdot \sum_{i=0}^{m-1} ((\mathbf{x}_i \cdot \mathbf{p})^3 - y_i \cdot \mathbf{x}_i \cdot \mathbf{p}) \cdot \mathbf{x}_i$$

and

$$\frac{d^2E}{d\mathbf{p}^2}(\mathbf{p}, X, \mathbf{y}) = 4 \cdot \sum_{i=0}^{m-1} (3 \cdot (\mathbf{x}_i \cdot \mathbf{p})^2 - y_i) \cdot \mathbf{x}_i^T \cdot \mathbf{x}_i \cdot$$

We present implementations for the solution of the convex unconstrained minimization problem

$$\min_{\mathbf{p} \in \mathbb{R}^n} E(\mathbf{p}, X, \mathbf{y})$$

using

- ▶ steepest descent
- ▶ Newton's method with differentiation performed
 - ▶ symbolically
 - ▶ approximately (finite differences)
 - ▶ algorithmically (dco/c++)

See source code.

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Summary

- ▶ Introduction to sample code approaching the Modern Family example as a general convex minimization problem.
- ▶ Solution of Modern Family problem using steepest descent and Newton's method.

Next Steps

- ▶ Run the sample code.
- ▶ Compare the results.
- ▶ Continue the course to find out more ...