

# CONSTRUCTING A FOOLING FUNCTION FOR THE TRAPEZOIDAL RULE

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## 1. FOOLING FUNCTION

Choose a large  $N = 2^n$  and define

$$x_i = \frac{i}{N}, \quad i = 0, \dots, N,$$

These nodes are used to define a *periodic* piecewise linear function whose values at the nodes need to be solved. These function values, the derivative values, and the changes in the derivative values are defined as

$$\begin{aligned} f_i &= f(x_i), & i &= 0, \dots, N, & f_N &= f_0 \\ d_i &= f_{i+1} - f_i, & i &= 0, \dots, N-1, & d_N &= d_0, \\ s_i &= d_{i+1} - d_i, & i &= 0, \dots, N-1. \end{aligned}$$

Based on these definitions, it follows that

$$\begin{aligned} d_i &= d_0 + (d_1 - d_0) + (d_2 - d_1) + \dots + (d_i - d_{i-1}) \\ &= d_0 + s_0 + s_1 + \dots + s_{i-1}, & i &= 0, \dots, N, \\ 0 &= s_0 + s_1 + \dots + s_{N-1}, \\ f_i &= f_0 + (f_1 - f_0) + (f_2 - f_1) + \dots + (f_i - f_{i-1}) \\ &= f_0 + d_0 + d_1 + \dots + d_{i-1} \\ &= f_0 + id_0 + (i-1)s_0 + (i-2)s_1 + \dots + s_{i-2} \\ &= f_0 + id_0 + \sum_{j=0}^{i-1} (i-j-1)s_j, & i &= 1, \dots, N, \\ 0 &= d_0 + d_1 + \dots + d_{N-1} \\ &= Nd_0 + (N-1)s_0 + (N-2)s_1 + \dots + s_{N-2}. \end{aligned}$$

Since  $f$  is piecewise linear, the integral of  $f$  is given by the trapezoidal rule:

$$\begin{aligned} I(f) &:= \int_0^1 f(x) dx = \frac{1}{2N}(f_0 + 2f_1 + 2f_2 + \dots + 2f_{N-1} + f_N) \\ &= \frac{1}{N}(f_0 + f_1 + f_2 + \dots + f_{N-1}) \\ &= \frac{1}{N} \sum_{i=0}^{N-1} f_i \\ &= f_0 + d_0 \frac{1}{N} \sum_{i=0}^{N-1} i + \frac{1}{N} \sum_{i=1}^{N-1} \sum_{j=0}^{i-1} (i-j-1)s_j \end{aligned}$$

$$\begin{aligned}
&= f_0 + \frac{d_0(N-1)}{2} + \frac{1}{N} \sum_{j=0}^{N-2} s_j \sum_{i=j+1}^{N-1} (i-j-1) \\
&= f_0 + \frac{d_0(N-1)}{2} + \frac{1}{2N} \sum_{j=0}^{N-2} (N-1-j)(N-2-j)s_j
\end{aligned}$$

## 2. TRAPEZOIDAL RULE

Suppose that there is a trapezoidal rule with  $M = 2^m < N$  trapezoids. Then only every  $L^{\text{th}}$  function value is used, where  $L = 2^l = N/M$ ,  $l = n - m$ :

$$\begin{aligned}
T_M(f) &:= \frac{1}{M}(f_0 + f_L + f_{2L} + \cdots + f_{N-L}) = \frac{1}{M} \sum_{i=0}^{M-1} f_{iL} \\
&= f_0 + d_0 L \frac{1}{M} \sum_{i=0}^{M-1} i + \frac{1}{M} \sum_{i=1}^{M-1} \sum_{j=0}^{iL-1} (iL - j - 1) s_j \\
&= f_0 + \frac{d_0 L(M-1)}{2} + \frac{1}{M} \sum_{j=0}^{N-L-1} s_j \sum_{i=\lceil (j+1)/L \rceil}^{M-1} (iL - j - 1) \\
&= f_0 + \frac{d_0(N-L)}{2} + \\
&\quad \frac{1}{2M} \sum_{j=0}^{N-L-1} \left( M - \left\lceil \frac{j+1}{L} \right\rceil \right) \left( N - L + L \left\lceil \frac{j+1}{L} \right\rceil - 2j - 2 \right) s_j.
\end{aligned}$$

We will also consider a trapezoidal rule with  $M/2$  trapezoids, which then uses every  $2L^{\text{th}}$  function value:

$$\begin{aligned}
T_{M/2}(f) &:= \frac{1}{M}(f_0 + f_{2L} + f_{4L} + \cdots + f_{N-2L}) = \frac{2}{M} \sum_{i=0}^{M/2-1} f_{2iL} \\
&= f_0 + \frac{d_0(N-2L)}{2} + \\
&\quad \frac{1}{M} \sum_{j=0}^{N-2L-1} \left( \frac{M}{2} - \left\lceil \frac{j+1}{2L} \right\rceil \right) \left( N - 2L + 2L \left\lceil \frac{j+1}{2L} \right\rceil - 2j - 2 \right) s_j.
\end{aligned}$$

The error of this trapezoidal rule is

$$I(f) - T_M(f) = \frac{1}{N} [(1-L)f_0 + f_1 + \cdots + f_{L-1} + (1-L)f_L + \cdots + f_{N-1}]$$

## 3. CONSTRAINED OPTIMIZATION

Given  $m, n, p \in \mathbb{N}$  with  $p \leq m \leq n$ , and given

**A**  $n \times n$  symmetric, positive definite

**B**  $n \times m$

**C**  $p \times m$

**d**  $m \times 1$ ,

we want to find  $\mathbf{x} \in \mathbb{R}^n$ ,  $\mathbf{y} \in \mathbb{R}^p$  such that

$$\begin{aligned} (\mathbf{x}, \mathbf{y}) \text{ minimizes } & \mathbf{x}^T \mathbf{A} \mathbf{x} \\ \text{subject to } & \mathbf{B}^T \mathbf{x} + \mathbf{C}^T \mathbf{y} = \mathbf{d}. \end{aligned}$$

First we write  $\mathbf{x}$  in terms of  $\mathbf{B}$  as follows:

$$\mathbf{x} = \mathbf{A}^{-1} \mathbf{B} \boldsymbol{\xi} + \mathbf{B}_\perp \boldsymbol{\xi}_\perp,$$

where  $\mathbf{B}_\perp$  is  $n \times n - m$  such that  $(\mathbf{B} \mid \mathbf{B}_\perp)$  has full rank, and the columns of  $\mathbf{B}_\perp$  are perpendicular to the columns of  $\mathbf{B}$ . The  $m$ -vector  $\boldsymbol{\xi}$  and the  $n - m$ -vector  $\boldsymbol{\xi}_\perp$  are the new unknowns replacing  $\mathbf{x}$ . This implies that

$$\begin{aligned} \mathbf{x}^T \mathbf{A} \mathbf{x} &= \boldsymbol{\xi}^T \mathbf{B}^T \mathbf{A}^{-1} \mathbf{B} \boldsymbol{\xi} + \boldsymbol{\xi}_\perp^T \mathbf{B}_\perp^T \mathbf{A} \mathbf{B}_\perp \boldsymbol{\xi}_\perp \\ \mathbf{d} &= \mathbf{B}^T \mathbf{A}^{-1} \mathbf{B} \boldsymbol{\xi} + \mathbf{C}^T \mathbf{y} \end{aligned}$$

Thus, one should choose  $\mathbf{x}_\perp = \mathbf{0}$ . One may now solve for  $\boldsymbol{\xi}$  in terms of  $\mathbf{y}$ :

$$\boldsymbol{\xi} = (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} (\mathbf{d} - \mathbf{C}^T \mathbf{y}).$$

Then the quantity to minimize becomes

$$\begin{aligned} \mathbf{x}^T \mathbf{A} \mathbf{x} &= \boldsymbol{\xi}^T \mathbf{B}^T \mathbf{A}^{-1} \mathbf{B} \boldsymbol{\xi} \\ &= (\mathbf{d} - \mathbf{C}^T \mathbf{y})^T (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} (\mathbf{d} - \mathbf{C}^T \mathbf{y}) \\ &= \mathbf{y}^T \mathbf{C} (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} \mathbf{C}^T \mathbf{y} - 2 \mathbf{d}^T (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} \mathbf{C}^T \mathbf{y} + \mathbf{d}^T (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} \mathbf{d} \end{aligned}$$

The value of  $\mathbf{y}$  that minimizes this quantity is

$$\mathbf{y} = [\mathbf{C} (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} \mathbf{C}^T]^{-1} \mathbf{C} (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} \mathbf{d}$$

To solve this numerically in a stable way, perhaps we should use singular value decompositions. First we form  $(\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1}$ :

$$\begin{aligned} \mathbf{A} &= \mathbf{V}_1 \boldsymbol{\Lambda}_1^2 \mathbf{V}_1^T, \quad \mathbf{A}_1, \mathbf{V}_1 \ n \times n, \quad \boldsymbol{\Lambda}_1 \text{ diagonal}, \quad \mathbf{V}_1^T \mathbf{V}_1 = \mathbf{I}, \\ \boldsymbol{\Lambda}_1^{-1} \mathbf{V}_1^T \mathbf{B} &= \mathbf{U}_2 \boldsymbol{\Lambda}_2 \mathbf{V}_2^T, \quad \mathbf{U}_2 \ n \times m, \quad \boldsymbol{\Lambda}_2, \mathbf{V}_2 \ n \times n, \\ \mathbf{U}_2^T \mathbf{U}_2 &= \mathbf{V}_2^T \mathbf{V}_2 = \mathbf{I}, \quad \boldsymbol{\Lambda}_2 \text{ diagonal}, \\ \mathbf{B}^T \mathbf{A}^{-1} \mathbf{B} &= \mathbf{B}^T \mathbf{V}_1 \boldsymbol{\Lambda}_1^{-2} \mathbf{V}_1^T \mathbf{B} = \mathbf{V}_2 \boldsymbol{\Lambda}_2^2 \mathbf{V}_2^T, \\ (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} &= \mathbf{V}_2 \boldsymbol{\Lambda}_2^{-2} \mathbf{V}_2^T. \end{aligned}$$

Next we form  $[\mathbf{C} (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} \mathbf{C}^T]^{-1}$ :

$$\begin{aligned} \boldsymbol{\Lambda}_2^{-1} \mathbf{V}_2^T \mathbf{C}^T &= \mathbf{U}_3 \boldsymbol{\Lambda}_3 \mathbf{V}_3^T, \quad \mathbf{U}_3 \ m \times p, \quad \boldsymbol{\Lambda}_3, \mathbf{V}_3 \ p \times p, \\ \mathbf{U}_3^T \mathbf{U}_3 &= \mathbf{V}_3^T \mathbf{V}_3 = \mathbf{I}, \quad \boldsymbol{\Lambda}_3 \text{ diagonal}, \\ \mathbf{C} (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} &= \mathbf{V}_3 \boldsymbol{\Lambda}_3 \mathbf{U}_3^T \boldsymbol{\Lambda}_2^{-1} \mathbf{V}_2^T, \\ \mathbf{C} (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} \mathbf{C}^T &= \mathbf{V}_3 \boldsymbol{\Lambda}_3^2 \mathbf{V}_3^T, \\ [\mathbf{C} (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} \mathbf{C}^T]^{-1} &= \mathbf{V}_3 \boldsymbol{\Lambda}_3^{-2} \mathbf{V}_3^T. \end{aligned}$$

Then we solve for  $\mathbf{y}$ :

$$\begin{aligned} \mathbf{y} &= [\mathbf{V}_3 \boldsymbol{\Lambda}_3^{-2} \mathbf{V}_3^T] [\mathbf{V}_3 \boldsymbol{\Lambda}_3 \mathbf{U}_3^T \boldsymbol{\Lambda}_2^{-1} \mathbf{V}_2^T] \mathbf{d} \\ &= \mathbf{V}_3 \boldsymbol{\Lambda}_3^{-1} \mathbf{U}_3^T \boldsymbol{\Lambda}_2^{-1} \mathbf{V}_2^T \mathbf{d}. \end{aligned}$$

Finally we solve for  $\mathbf{x}$  via  $\boldsymbol{\xi}$ :

$$\begin{aligned}
\mathbf{C}^T \mathbf{y} &= [\mathbf{V}_2 \boldsymbol{\Lambda}_2 \mathbf{U}_3 \boldsymbol{\Lambda}_3 \mathbf{V}_3^T] \mathbf{V}_3 \boldsymbol{\Lambda}_3^{-1} \mathbf{U}_3^T \boldsymbol{\Lambda}_2^{-1} \mathbf{V}_2^T \mathbf{d} \\
&= \mathbf{V}_2 \boldsymbol{\Lambda}_2 \mathbf{U}_3 \mathbf{U}_3^T \boldsymbol{\Lambda}_2^{-1} \mathbf{V}_2^T \mathbf{d} \\
\mathbf{d} - \mathbf{C}^T \mathbf{y} &= \mathbf{V}_2 \boldsymbol{\Lambda}_2 (\mathbf{I} - \mathbf{U}_3 \mathbf{U}_3^T) \boldsymbol{\Lambda}_2^{-1} \mathbf{V}_2^T \mathbf{d} \\
\boldsymbol{\xi} &= (\mathbf{B}^T \mathbf{A}^{-1} \mathbf{B})^{-1} (\mathbf{d} - \mathbf{C}^T \mathbf{y}) \\
&= [\mathbf{V}_2 \boldsymbol{\Lambda}_2^{-2} \mathbf{V}_2^T] \mathbf{V}_2 \boldsymbol{\Lambda}_2 (\mathbf{I} - \mathbf{U}_3 \mathbf{U}_3^T) \boldsymbol{\Lambda}_2^{-1} \mathbf{V}_2^T \mathbf{d} \\
&= \mathbf{V}_2 \boldsymbol{\Lambda}_2^{-1} (\mathbf{I} - \mathbf{U}_3 \mathbf{U}_3^T) \boldsymbol{\Lambda}_2^{-1} \mathbf{V}_2^T \mathbf{d} \\
\mathbf{x} &= \mathbf{A}^{-1} \mathbf{B} \boldsymbol{\xi} \\
&= [\mathbf{U}_2 \boldsymbol{\Lambda}_2 \mathbf{V}_2^T] \mathbf{V}_2 \boldsymbol{\Lambda}_2^{-1} (\mathbf{I} - \mathbf{U}_3 \mathbf{U}_3^T) \boldsymbol{\Lambda}_2^{-1} \mathbf{V}_2^T \mathbf{d} \\
&= \mathbf{U}_2 (\mathbf{I} - \mathbf{U}_3 \mathbf{U}_3^T) \boldsymbol{\Lambda}_2^{-1} \mathbf{V}_2^T \mathbf{d}
\end{aligned}$$