

Derivations

This worksheet is intended to document the derivation of the least squares procedure followed to fit Model 1 and Model 2 to acceleration data. It is not needed during the activity, but may be adapted to use in student research or marking period project.

1 Piecewise constant model (Model 1)

We are given acceleration values a_1, a_2, \dots, a_N for the times t_1, t_2, \dots, t_N . Our goal is to fit a constant c_1 to the data over the interval $0 \leq t \leq t_1$, and another constant c_2 over the interval $t_1 < t \leq t_2$. In doing so, we must make sure that the total area under the fitted graph is equal to zero, corresponding to zero initial and final racer velocities. This is a constrained least squares problem. Let N_1 and N_2 denote the number of acceleration data points that fall in time intervals $0 \leq t \leq t_1$ and $t_1 < t \leq t_2$. We can then minimize

$$F = \sum_{i=1}^{N_1} (c_1 - a_i)^2 + \sum_{i=N_1+1}^{N_2} (c_2 - a_i)^2 \quad (1)$$

subject to the constraint

$$G = c_1 t_1 + c_2 (t_2 - t_1) = 0. \quad (2)$$

We can solve this problem by forming the Lagrangian

$$\mathbf{grad}F = \lambda \mathbf{grad}G, \text{ and } G = 0, \quad (3)$$

which gives us three equations in three unknowns

$$\begin{aligned}
2 \sum_{i=1}^{N_1} (c_1 - a_i) &= \lambda t_1, \\
2 \sum_{i=N_1+1}^{N_2} (c_2 - a_i) &= \lambda (t_2 - t_1), \\
c_1 t_1 + c_2 (t_2 - t_1) &= 0.
\end{aligned} \tag{4}$$

Eliminating λ from equations 1 and 2 leaves us with two linear equations in two unknowns c_1 and c_2

$$\begin{aligned}
\frac{N_1}{t_1} c_1 + \frac{N_2}{t_1 - t_2} c_2 &= \frac{1}{t_1} \sum_{i=1}^{N_1} a_i + \frac{1}{t_1 - t_2} \sum_{i=N_1+1}^{N_2} a_i = \beta \\
c_1 t_1 + c_2 (t_2 - t_1) &.
\end{aligned} \tag{5}$$

When we solve these two equations for the unknowns c_1 and c_2 , we get

$$\begin{aligned}
c_1 &= \frac{t_1 (t_1 - t_2)^2 \beta}{(t_1^2 N_2 + (t_1 - t_2)^2 N_1)}, \\
c_2 &= \frac{t_1^2 (t_1 - t_2) \beta}{(t_1^2 N_2 + (t_1 - t_2)^2 N_1)},
\end{aligned} \tag{6}$$

where β is

$$\beta = \frac{1}{t_1} \sum_{i=1}^{N_1} a_i + \frac{1}{t_1 - t_2} \sum_{i=N_1+1}^{N_2} a_i. \tag{7}$$

These formulas are implemented in the Analysis Spreadsheet.

2 Piecewise linear model (Model 2)

We are given acceleration data a_1, a_2, \dots, a_N along with times t_1, t_2, \dots, t_N , and we want to fit a line $\alpha_1 + \beta_1 t$ to data over the interval $0 \leq t \leq t_1$ and another line $\alpha_2 + \beta_2 t$ to data over the interval $t_1 < t \leq t_2$. Whereas we had to find only two unknowns in the case of the piecewise constant model, in the piecewise linear model, we need to find four unknowns. Let N_1 and

N_2 denote the number of acceleration data points that fall in time intervals $0 \leq t \leq t_1$ and $t_1 < t \leq t_2$. We can then minimize

$$\sum_{i=1}^{N_1} (\alpha_1 + \beta_1 t_i - a_i)^2 + \sum_{i=N_1+1}^{N_2} (\alpha_2 + \beta_2 t_i - a_i)^2 \quad (8)$$

subject to the zero area constraint

$$\alpha_1 t_1 + \frac{1}{2} \beta_1 t_1^2 + \alpha_2 (t_2 - t_1) + \frac{1}{2} \beta_2 (t_2^2 - t_1^2) = 0. \quad (9)$$

This problem is more easily solved by working with matrices. Let's define a vector of unknowns

$$\theta = (\alpha_1 \quad \beta_1 \quad \alpha_2 \quad \beta_2)^t, \quad (10)$$

two vectors of acceleration measurements

$$\begin{aligned} A_1 &= (a_1 \quad \cdots \quad a_{N_1})^t, \\ A_2 &= (a_{N_1+1} \quad \cdots \quad a_{N_2})^t, \end{aligned} \quad (11)$$

a vector of constants

$$C = (t_1 \quad \frac{1}{2} t_1^2 \quad (t_2 - t_1) \quad \frac{1}{2} (t_2^2 - t_1^2))^t, \quad (12)$$

and two matrices T_1 and T_2 as

$$T_1 = \begin{pmatrix} 1 & \cdots & 1 \\ t_1 & \cdots & t_{N_1} \\ 0 & \cdots & 0 \\ 0 & \cdots & 0 \end{pmatrix}^t, \quad (13)$$

$$T_2 = \begin{pmatrix} 0 & \cdots & 0 \\ 0 & \cdots & 0 \\ 1 & \cdots & 1 \\ t_{N_1+1} & \cdots & t_{N_2} \end{pmatrix}^t. \quad (14)$$

We can now write our constrained optimization as follows. Minimize the quantity

$$(A_1 - T_1 \theta)^t (A_1 - T_1 \theta) + (A_2 - T_2 \theta)^t (A_2 - T_2 \theta) \quad (15)$$

with respect to θ , subject to the constraint

$$C^t \theta = 0. \quad (16)$$

To solve this problem, we form the Lagrangian

$$J = \frac{1}{2} (A_1 - T_1\theta)^t (A_1 - T_1\theta) + \frac{1}{2} (A_2 - T_2\theta)^t (A_2 - T_2\theta) - (C^t\theta)^t \lambda. \quad (17)$$

Equating the derivative to zero gives us

$$\theta = (T_1^t T_1 + T_2^t T_2)^{-1} (T_1^t A_1 + T_2^t A_2) - (T_1^t T_1 + T_2^t T_2)^{-1} C \lambda. \quad (18)$$

To solve for λ , we can plug in for the constraint to get

$$\lambda = \left(C^t (T_1^t T_1 + T_2^t T_2)^{-1} C \right)^{-1} C^t \left((T_1^t T_1 + T_2^t T_2)^{-1} (T_1^t A_1 + T_2^t A_2) \right). \quad (19)$$

Substituting this expression for λ in equation (18) gives us the solution

$$\theta = M_1^{-1} M_2 - \frac{M_1^{-1} C C^t}{C^t M_1^{-1} C} M_1^{-1} M_2, \quad (20)$$

where

$$\begin{aligned} M_1 &= T_1^t T_1 + T_2^t T_2 \\ M_2 &= T_1^t A_1 + T_2^t A_2. \end{aligned} \quad (21)$$

This formula is implemented in the Analysis Spreadsheet.