

Lecture Content Summaries for MECH3750

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Week 02

With thanks to Alex Muirhead:

Tutors: Alex Muirhead, Bryce Hill, Kyle McLaren, Luke Bartholomew, William Snell

Quizzes: Weeks 3, 6, 9

Content:

- Taylor Expansions;
- Newton's Method;
- Least Squares.

Taylor Series

a is the fixed point

$$f(x) = f(a) + \frac{f'(a)}{1!} (x-a) + \frac{f''(a)}{2!} (x-a)^2 + \dots$$

↑ ↑ ↑

function we distance from infinite
want to approx the fixed point series

$$= \sum_{n=0}^{\infty} \frac{f^{(n)}(a)}{n!} (x-a)^n$$

But can't always have all infinite terms.

$$f(x) = f(a) + \frac{f'(a)}{1!} (x-a) + \underbrace{o((x-a)^2)}_{\text{error term}}$$

OR

aka Higher in mult-variable
Order
Terms

$$f(x) \approx f(a) + \frac{f'(a)}{1!} (x-a)$$

Accurate to 2nd order

$$\text{Note: } f_x = \frac{\partial f}{\partial x} \text{ etc.}$$

What about multivariable functions?

$$f(x, y, \dots) \text{ OR } f(x_0, x_1, \dots, x_n) = f(\vec{x})$$

two fixed points

$$f(x, y) = f(a, b) + f_x(a, b)(x-a) + f_y(a, b)(y-b)$$

$$\dots + \frac{1}{2} \left[f_{xx}(a, b)(x-a)^2 + 2f_{xy}(a, b)(x-a)(y-b) + f_{yy}(a, b)(y-b)^2 \right]$$

$$\dots + \text{H.O.T} \leftarrow O(\sqrt{(x-a)^2 + (y-b)^2})$$

In general:

$$f(\vec{x}) = f(\vec{x}_0) + [\nabla f(\vec{x}_0)]^T (\vec{x} - \vec{x}_0) + \frac{1}{2} (\vec{x} - \vec{x}_0)^T H (\vec{x} - \vec{x}_0) + \text{H.O.T}$$

where Hessian is $H = \nabla(\nabla f) = \begin{bmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{bmatrix}$

Newton's Method

Want to find roots of function:

$$f(x) = \emptyset$$

$$f(x_0 + h) \approx f(x_0) + f'(x_0)h$$

$\uparrow \quad \uparrow$
initial guess step

Let our next guess $x_1 = x_0 + h$ be a "root".

$$\therefore f'(x_0)h = -f(x_0)$$

$$x_1 = x_0 + h = x_0 - \frac{f(x_0)}{f'(x_0)}$$

Repeat until
 $f(x_n) \approx 0$

For multi-variable functions; and vector functions (aka multiple stacked equations).

$$\vec{f}(\vec{x}) = 0$$

$$\vec{f}'(\vec{x}_0) = \nabla \vec{f}(\vec{x}_0) = J \quad \{ \text{Jacobian} \}$$

$$J = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots \\ \frac{\partial f_2}{\partial x_1} & \ddots & \vdots \\ \vdots & \dots & \frac{\partial f_m}{\partial x_m} \end{bmatrix}, \quad \vec{f}(\vec{x}) = \begin{bmatrix} f_1(\vec{x}) \\ f_2(\vec{x}) \\ \vdots \\ f_m(\vec{x}) \end{bmatrix}$$

$$\vec{f}(\vec{x}_0 + \vec{h}) \approx \vec{f}(\vec{x}_0) + J \vec{h} = \emptyset$$

$$\therefore J \vec{h} = -\vec{f}(\vec{x}_0) \quad \leftarrow \text{easier to solve numerically}$$

$$\text{OR } \vec{h} = -J^{-1} \vec{f}(\vec{x}_0) \quad \leftarrow \text{easier to solve by hand}$$

IFF J is 2×2 matrix.

Least Squares

Want to approximate some vector \vec{f} with vectors $\vec{p}^{(1)}, \vec{p}^{(2)}, \dots, \vec{p}^{(n)}$.

$$\vec{f} \approx \alpha_1 \vec{p}^{(1)} + \alpha_2 \vec{p}^{(2)} + \dots + \alpha_n \vec{p}^{(n)}$$

Define positive error:

$$E = \sum_i (\alpha_1 p_i^{(1)} + \alpha_2 p_i^{(2)} + \dots + \alpha_n p_i^{(n)} - f_i)^2$$

\nwarrow summing over components
of the vectors

square to ensure error is positive

Minimise error by setting derivatives to zero.

$$\frac{\partial E}{\partial \alpha_j} = 0 \quad \forall j \in 1, 2, \dots, n$$

$$\frac{1}{2} \frac{\partial E}{\partial \alpha_j} = \sum_i p_i^{(j)} (\alpha_1 p_i^{(1)} + \alpha_2 p_i^{(2)} + \dots + \alpha_n p_i^{(n)} - f_i)$$

$$0 = \underbrace{\sum_i p_i^{(j)} (\alpha_1 p_i^{(1)} + \alpha_2 p_i^{(2)} + \dots + \alpha_n p_i^{(n)})}_{\text{unknowns}} - \underbrace{\sum_i p_i^{(j)} f_i}_{\text{knowns}}$$

In matrix notation:

$$\begin{bmatrix} p^{(1)} \cdot f \\ p^{(2)} \cdot f \\ \vdots \\ p^{(n)} \cdot f \\ P^T f \end{bmatrix} = \begin{bmatrix} p^{(1)} \cdot p^{(1)} & p^{(1)} \cdot p^{(2)} & \dots & p^{(1)} \cdot p^{(n)} \\ p^{(2)} \cdot p^{(1)} & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \vdots \\ p^{(n)} \cdot p^{(1)} & \dots & \dots & p^{(n)} \cdot p^{(n)} \\ P^T P \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \\ \alpha \end{bmatrix}$$

$$\vec{\alpha} = (P^T P)^{-1} P^T \vec{f}$$

Week 03

With thanks to Alex Muirhead:

Tutors: Alex Muirhead, Bryce Hill, Kyle McLaren, Luke Bartholomew, William Snell

Quiz: Weeks 3 - 1pm

Content:

- Least Squares continued;
- Inner product;
- Orthogonal polynomials;
- Fourier Series.

Least Squares (cont.)

Can represent discrete points on some function $f(x)$ by a vector.

$$f(\vec{x}) = (f(x_0), f(x_1), \dots, f(x_n))^T = \vec{f}$$

Approximate vector with basis "vectors" made from functions.

$$\vec{p}^{(1)} = (p^{(1)}(x_0), p^{(1)}(x_1), \dots, p^{(1)}(x_n))^T$$

etc.

Can be any set of linearly independent functions $\{p^{(1)}, p^{(2)}, \dots, p^{(m)}\}$

$$\therefore f(\vec{x}) = \vec{f} \approx \sum_m \alpha_m \vec{p}^{(m)}$$

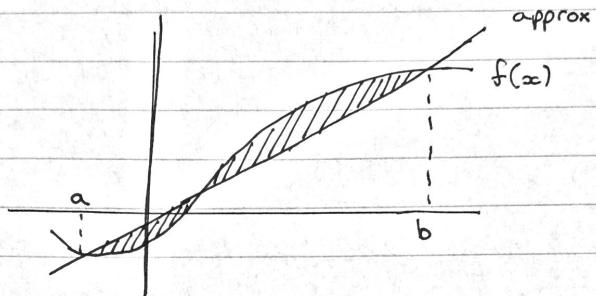
$$\begin{bmatrix} \vec{p}^{(1)} \cdot \vec{p}^{(1)} & \vec{p}^{(1)} \cdot \vec{p}^{(2)} & \dots & \vec{p}^{(1)} \cdot \vec{p}^{(m)} \\ \vec{p}^{(2)} \cdot \vec{p}^{(1)} & \dots & & \vdots \\ \vdots & \ddots & \vdots & \vdots \\ \vec{p}^{(m)} \cdot \vec{p}^{(1)} & \dots & \dots & \vec{p}^{(m)} \cdot \vec{p}^{(m)} \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \vdots \\ \vdots \\ \alpha_m \end{bmatrix} = \begin{bmatrix} \vec{f} \cdot \vec{p}^{(1)} \\ \vdots \\ \vdots \\ \vec{f} \cdot \vec{p}^{(m)} \end{bmatrix}$$

Same form as before.

Written more concisely as:

$$\begin{aligned} P^T P \vec{\alpha} &= P^T \vec{f} \\ \downarrow \\ \vec{\alpha} &= (P^T P)^{-1} P^T \vec{f} \end{aligned}$$

What if we want to approximate a continuous function over an interval?



Define positive error based on the area between $f(x)$ and approximation.

$$E = \int_a^b \left(\sum_i^n \alpha_i p^{(i)}(x) - f(x) \right)^2 dx$$

Follow previous derivation, set $\partial_{\alpha_i} E = 0$

$$0 = \sum_i^n \int_a^b \alpha_i p^{(i)}(x) p^{(j)}(x) dx - \int_a^b p^{(j)}(x) f(x) dx$$

In matrix form:

$$\begin{bmatrix} \int_a^b p^{(1)} p^{(1)} dx & \dots & \int_a^b p^{(1)} p^{(m)} dx \\ \vdots & \ddots & \vdots \\ \int_a^b p^{(m)} p^{(1)} dx & \dots & \int_a^b p^{(m)} p^{(m)} dx \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix} = \begin{bmatrix} \int_a^b p^{(1)} f dx \\ \vdots \\ \int_a^b p^{(m)} f dx \end{bmatrix}$$

Discrete vector form & continuous function
forms of least squares very similar...

Generalise by introducing inner product.

$$\begin{array}{ccc} \langle p, q \rangle & & \\ \text{vectors} \swarrow & & \searrow \text{functions} \\ \vec{p} \cdot \vec{q} & & \int p(x)q(x) dx \end{array}$$

Some properties:

$$\text{Norm (or length/size)} \Rightarrow \|p\| = \sqrt{\langle p, p \rangle}$$

$$\text{Distance} \Rightarrow d(p, q) = \|p - q\|$$

$$\text{Orthogonality} \Rightarrow \langle p, q \rangle = 0$$

Distance between two functions becomes:

$$d(p, q) = \sqrt{\int (p(x) - q(x))^2 dx}$$

\therefore Least squares minimises distance squared!

There exist sets of linearly independent orthogonal functions. These simplify matrix form, as non-diagonal elements become \emptyset .

Legendre Polynomials

$$\int_0^1 p^{(i)} p^{(j)} dx \propto \delta_{ij} \equiv \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$$

$$P_0(x) = 1$$

$$P_1(x) = x$$

$$P_2(x) = \frac{1}{2}(3x^2 - 1)$$

$$P_3(x) = \frac{1}{2}(5x^3 - 3x)$$

$$P_n(x) = \frac{1}{2^n n!} \frac{d^n}{dx^n} (x^2 - 1)^n$$

\curvearrowleft Don't need to remember this!

Trigonometric Functions

Specifically sine and cosine, as they can be represented as:

$$e^{ix} = \cos x + i \sin x$$

Note that orthogonality holds:

$$\int_{-\pi}^{\pi} \sin(mx) \sin(nx) dx = \begin{cases} \pi & \text{if } m=n \\ 0 & \text{otherwise} \end{cases}$$

$$\int_{-\pi}^{\pi} \cos(mx) \cos(nx) dx = \begin{cases} \pi & \text{if } m=n * \\ 0 & \text{otherwise} \end{cases}$$

$$\int_{-\pi}^{\pi} \sin(mx) \cos(nx) dx = 0$$

* if $m, n > 0$

Fourier Series

$$f(x) = \frac{a_0}{2} + \sum_{n=1}^{\infty} a_n \cos(nx) + b_n \sin(nx)$$

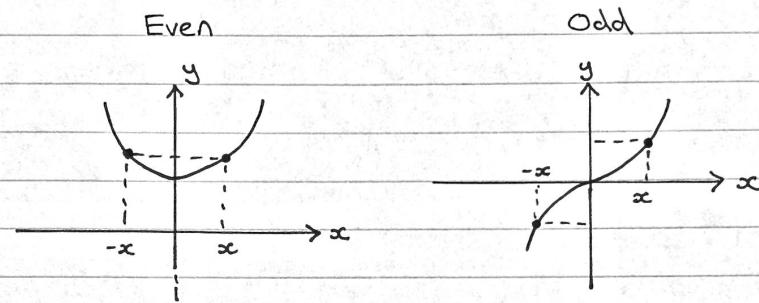
↑
exactly equal for an
infinite sum

$$\text{where } a_n = \frac{1}{\pi} \int_{-\pi}^{\pi} f(x) \cos(nx) dx$$

$$b_n = \frac{1}{\pi} \int_{-\pi}^{\pi} f(x) \sin(nx) dx$$

Shortcuts

Introduced even and odd functions.



$$\left. \begin{array}{l} f(-x) = f(x) \\ \text{Reflection Symmetry} \\ \text{about } y\text{-axis} \end{array} \right\}$$

$$\left. \begin{array}{l} f(-x) = -f(x) \\ \text{Rotation Symmetry} \\ \text{about origin} \end{array} \right\}$$

$$\text{Even} \times \text{Even} \Rightarrow f(-x)g(-x) = f(x)g(x) \Rightarrow \text{Even}$$

$$\text{Odd} \times \text{Odd} \Rightarrow f(-x)g(-x) = (-1)^2 f(x)g(x) \Rightarrow \text{Even}$$

$$\text{Even} \times \text{Odd} \Rightarrow f(-x)g(-x) = -f(x)g(x) \Rightarrow \text{Odd}$$

$$\int_{-a}^a \text{Even } dx = \int_{-a}^0 f(x) dx + \int_0^a f(x) dx = 2 \int_0^a f(x) dx$$

$$\int_{-a}^a \text{Odd } dx = \int_{-a}^0 g(x) dx + \int_0^a g(x) dx = \int_0^a -g(x) + g(x) dx = \emptyset$$

∴ As cosine is even & sine is odd

if $f(x)$ is even:

$$f(x) = \frac{a_0}{2} + \sum_{n=1}^{\infty} a_n \cos(nx)$$

if $f(x)$ is odd:

$$f(x) = \sum_{n=1}^{\infty} b_n \sin(nx)$$

Week 04

Tutors: Luke Bartholomew, Bryce Hill, Kyle McLaren, Travis Mitchell, Alex Muirhead, William Snell

Assessment:

- Quiz 1 results are available for collection;
- Next quiz week 6;
- Assignment 1 due week 7.

Content:

- Least Squares continued;
- Inner product;
- Orthogonal polynomials;
- Fourier Series.

1 Fourier Series continued

Example 1.1.

Find the Fourier Series of $f(x) = x^2$ on $[-\pi, \pi]$.

This example shows a key point to note (particularly for quizzes/exams), and that is to check if the function is *odd* or *even*! The first step here is to determine our Fourier coefficients:

$$\begin{aligned} b_j &= \frac{1}{\pi} \int_{-\pi}^{\pi} \underbrace{x^2}_{\text{even}} \underbrace{\sin(jx)}_{\text{odd}} dx \\ &= 0 \\ a_j &= \frac{1}{\pi} \int_{-\pi}^{\pi} \underbrace{x^2}_{\text{even}} \underbrace{\cos(jx)}_{\text{even}} dx \\ &= \frac{2}{\pi} \int_0^{\pi} x^2 \cos(jx) dx \end{aligned}$$

The coefficients, a_j , can then be determined through integration by parts to be,

$$a_j = \left(\frac{4(-1)^j}{j^2} \right)$$

We now evaluate, a_0 , and then we can determine the Fourier series,

$$\begin{aligned} a_0 &= \frac{2}{\pi} \int_0^{\pi} x^2 dx \\ &= 2\pi^2/3 \end{aligned}$$

And therefore, as $f(x)$ is even,

$$\begin{aligned} f(x) &= a_0/2 + \sum_{n=1}^{\infty} a_n \cos(nx) \\ &= \pi^2/3 - \frac{4 \cos(x)}{1^2} + \frac{4 \cos(2x)}{2^2} - \frac{4 \cos(3x)}{3^2} + \dots \end{aligned}$$

Remarks from Fourier Theorem, namely is we can restate the formulation as,

$$\begin{aligned} f(x) &= \sum_{j=-n}^n c_j e^{ijx} \\ c_j &= \frac{1}{2\pi} \int_{-\pi}^{\pi} f(x) e^{-ijx} dx, \quad j = -n, \dots, -1, 0, 1, \dots, n \end{aligned}$$

This makes use of Euler's formula (or we can also show with Taylor series) that, $e^{ijx} = \cos jx + i \sin(jx)$.

2 Discrete Fourier Transform