

Solving differential equations with neural networks, October 21, 2024

Solving differential equations with deep neural networks

$$\frac{dg}{dx} = -\gamma g(x)$$

$$x \in [0, \infty)$$

$$g(x) = g_0 \exp(-\gamma x)$$

$$g_0 = g(x=0)$$

Population growth

$$\frac{dg}{dt} = \alpha g(t)(A - g(t))$$

$\alpha > 0$, growth rate $A > 0$
max population

$$g(0) = g_0$$

analytical solution:

$$g(t) = \frac{A g_0}{g_0 + (A - g_0)(\exp(-\alpha A t))}$$

Exponential decay

$$\frac{dg}{dx} - (-\gamma g(x)) = 0$$

$$= f(x, g(x), g'(x)) = 0$$

In general

$$f(x, g(x), g'(x), g''(x) - \dots \\ g^{(n)}(x)) = 0$$

Finite difference scheme
Taylor expand

$$g(t + \Delta t) = g(t) + \Delta t \cdot \frac{dg}{dt} \Big|_t + \frac{(\Delta t)^2}{2!} \frac{d^2 g}{dt^2} \Big|_t + O(\Delta t^3)$$

Discretize $t \rightarrow t_i'$

$$t_i' = t_0 + i \Delta t \quad i = 0, 1, 2, \dots, n$$

$$t_n = t_0 + n \Delta t \Rightarrow$$

$$\Delta t = \frac{t_n - t_0}{n}$$

$$g(t + \Delta t) \rightarrow g(t_i' + \Delta t) = g_{i+1}$$

$$g_{i+1} \approx g_i + \Delta t g_i'$$

$$\frac{dg}{dt} \Big|_{t=t_i} = g_i'$$

$$\text{start } i = 0$$

$$\frac{dg}{dt} = -\gamma g(t)$$

$$g_1 \approx g_0 - \gamma g_0$$

$$= -\gamma g(t=0)$$

$$g_2 \approx g_1 - \gamma g_1$$

$$= -\gamma g_0$$

:

$$g_n \approx g_{n-1} - \gamma g_{n-1}$$

$$\left. \begin{array}{l} \\ \\ \\ \end{array} \right\} \Sigma (\Delta t^2)$$

Two-point boundary value

$$-g''(x) = f(x) \quad \text{known function}$$

$$x \in [0, 1]$$

$$g(x=0) = g(x=1) = 0$$

(Dirichlet boundary conditions)

$$f(x) = (3x + x^2) \exp(x)$$

$$g(x) = x(1-x) \exp(x)$$

$$g''(x) = \frac{g(x+\Delta x) + g(x-\Delta x) - 2g(x)}{(\Delta x)^2}$$

$$+ O(\Delta x^2)$$

$$-g''(x) \leq - \frac{(g(x+\Delta x) + g(x-\Delta x) - 2g(x))}{\Delta x^2}$$

$$= f(x)$$

$$x \rightarrow x_i' = x_0 + i' \Delta x$$

$$i = 0, 1, \dots, n \quad \Delta x = \frac{x_n - x_0}{n}$$

$$-g''_i = - \frac{d^2 g}{dx^2} \Big|_{x=x_i'}$$

$$-g''_i = \frac{-g_{i+1} - g_{i-1} + 2g_i}{\Delta x^2} = f_i$$

$$\Rightarrow A g = \Delta x^2 f$$

$$g = \begin{bmatrix} g_1 \\ g_2 \\ \vdots \\ g_{n-1} \end{bmatrix}$$

$$g_0 = g_n = 0$$

$$A = \begin{bmatrix} 2 & -1 & & & \\ -1 & 2 & -1 & & \\ & -1 & 2 & -1 & \\ & & \ddots & \ddots & \ddots \\ & & & -1 & 2 \end{bmatrix}$$

Neural Networks

$$f(x, g(x), g'(x), \dots, g^{(n)}(x)) = 0$$

Trial solution

$$g_t(x) = h_1(x) + h_2(x, N(x, \epsilon))$$

↑
satisfies
initial and/or
Boundary
conditions

Exponential decay

$$g_t(x) = g_0 + x N(x, \epsilon)$$

Two-point boundary value

$$g_t(x) = x(1-x)N(x, \epsilon)$$

$$x=0 \quad g_t(0) = 0$$

$$x=1 \quad g_t(1) = 0$$

Define Cost function

$$C(x, \epsilon) = \frac{1}{n} \sum_{i=0}^{n-1} \left(f(x_i, g_t(x_i)) - g_t^{(m)}(x_i) \right)^2$$

$$\hat{\theta} = \arg \min_{\theta} \frac{1}{n} \sum_i \left[f(x_i', g_{\theta}(x_i', \theta), g_{\theta}'(x_i, \theta), \dots, g_{\theta}^{(m)}(x_i', \theta)) \right]^2$$

exponential decay

$$C(x_i, \theta) = \frac{1}{n} \sum_i \left(g_{\theta}'(x_i', \theta) - (-\gamma g_{\theta}(x_i', \theta)) \right)^2$$