

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 solutions

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

Exponential decay

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

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

In general

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

Finite difference schemes  
Taylor expand

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

Discretize  $t \rightarrow t_i'$

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

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

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

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

$$g_{i+1} \stackrel{\sim}{=} g_i + \Delta t \dot{g}_i$$

$$\frac{dg}{dt} \Big|_{t=t_i} = \dot{g}_i$$

start  $i=0$

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

$$g_1 \simeq g_0 - \gamma g_0 \quad -\gamma g(t=0)$$

$$g_2 \simeq g_1 - \gamma g_1 \quad -\gamma g_0$$

⋮

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

$$\sum (\Delta t)^2$$

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)

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

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

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

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

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

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

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

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

$$\Rightarrow Ag = \Delta x^2 f$$

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

$$g_0 = g_n = 0$$

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

# Neural Networks

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

Trial solution

$$g_t(x) = h_1(x) + b_2(x, N(x, \epsilon))$$

↑  
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

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

$$\hat{g} = \arg \min_{\hat{g}} \frac{1}{n} \sum_i [f(x_i; g_t(x_i; \epsilon)), \\ g_t'(x_i; \epsilon), \dots g_t^{(m)}(x_i; \epsilon)]^T$$

exponential trial decay

$$C(X, g) = \frac{1}{n} \sum_i \left( g_t'(x_i; \epsilon) - (-\gamma g_t(x_i; \epsilon)) \right)^2$$