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CS229 ch3 Generalized linear model = $g(\mu) = \theta^T x$ linear model (linear in θ) \downarrow
link function

$$\text{ex } x \in \mathbb{R}^d, \mu = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \dots + \theta_d x_d = \theta^T x$$

$$\mu = \theta_0 x + \theta_1 x^2 + \theta_2 x^3 \dots \theta_{d-1} x^d = [\theta_0 \theta_1 \dots \theta_d] \begin{bmatrix} x \\ x^2 \\ x^3 \\ \vdots \\ x^d \end{bmatrix} = \theta^T T(x)$$

The exponential family

$$p(y; \eta) = b(y) e^{\eta^T y - a(\eta)}$$

$$= b(y) \exp(\eta^T y - a(\eta))$$

Bernoulli distribution (mean = ϕ)

$$p(y) = \phi^y (1-\phi)^{1-y}$$

$$= e^{y \ln \phi + (1-y) \ln(1-\phi)}$$

$$= e^{y(\ln \phi - \ln(1-\phi)) + \ln(1-\phi)}$$

$$= e^{y \ln \frac{\phi}{1-\phi} + \ln(1-\phi)}$$

Bernoulli

$$b(y) = 1, \quad y = \ln \frac{\phi}{1-\phi} \Rightarrow e^y = \frac{\phi}{1-\phi} \Rightarrow \phi = \frac{e^y}{1+e^y} \Rightarrow 1-\phi = \frac{1}{1+e^y}$$

$$a(y) = -\ln(1-\phi) = \ln(1+e^y)$$

$$\text{GLM: } h_0(x) = E(y|x)$$

$$= \phi$$

$$= \frac{1}{1+e^{-x}} = \frac{1}{1+e^{-\theta^T x}}$$

Poisson distribution (mean λ)

$$p(y; \lambda) = \frac{\lambda^y e^{-\lambda}}{y!} = \frac{1}{y!} \exp(y \ln \lambda - \lambda)$$

$$b(y) = \frac{1}{y!} \Rightarrow \eta = \ln \lambda \Rightarrow \lambda = e^\eta$$

$$GLM = E[y|x] = \lambda = e^\eta = e^{\theta^T x}$$

Normal distribution (mean μ) (var=1)

$$\begin{aligned} p(y; \mu) &= \frac{1}{\sqrt{2\pi}} e^{-\frac{(y-\mu)^2}{2}} \\ &= \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2} + \mu y - \frac{\mu^2}{2}} \\ &= \left(\frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} \right) \cdot e^{\mu y - \frac{\mu^2}{2}} \end{aligned}$$

$$b(y) = \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} \quad \eta = \mu \quad a(\eta) = \frac{\mu^2}{2} = \frac{y^2}{2}$$

$$GLM = \eta(\eta) = E[y|x] = \mu = \eta = \theta^T x$$

The exponential family

$$p(y; \eta) = b(y) \exp(\eta^T T(y) - a(\eta))$$

Consider normal distribution

$$\begin{aligned} p(y; \mu, \sigma) &= \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{(y-\mu)^2}{2\sigma^2}} = \frac{1}{\sqrt{2\pi}} e^{-\ln \sigma} e^{-\frac{y^2}{2\sigma^2} + \frac{\mu y}{\sigma^2} - \frac{\mu^2}{2\sigma^2}} \\ &= \frac{1}{\sqrt{2\pi}} e^{\left[\frac{\mu}{\sigma^2}, \frac{-1}{2\sigma^2} \right] \begin{bmatrix} y \\ 1 \end{bmatrix} - \frac{\mu^2}{2\sigma^2} - \ln \sigma} \end{aligned}$$

* normal distribution is an exponential family with two parameter

$$b(\eta) = \frac{1}{\sqrt{2\pi}}$$

$$a(\eta) = \frac{\eta^2}{2\sigma^2} + \ln \sigma = \frac{-\eta_1^2}{4\pi} - \frac{1}{2} \ln(-2\eta_2)$$

$$\eta = \begin{bmatrix} \frac{\mu}{\sigma^2} \\ -\frac{1}{2\sigma^2} \end{bmatrix} = \begin{bmatrix} \eta_1 \\ \eta_2 \end{bmatrix} \quad T(\eta) = \begin{bmatrix} \mu \\ \sigma^2 \end{bmatrix}$$

$$\eta = \begin{bmatrix} \eta_1 \\ \eta_2 \end{bmatrix} = \begin{bmatrix} \frac{\mu}{\sigma^2} \\ -\frac{1}{2\sigma^2} \end{bmatrix}$$

$$\rightarrow \eta_2 = \frac{-1}{2\sigma^2} \Rightarrow \sigma^2 = \sqrt{\frac{1}{2\eta_2}}$$

$$\eta_1 = \frac{\mu}{\sigma^2} \Rightarrow \mu = \eta_1 \cdot \sigma^2 = \frac{-\eta_1}{2\eta_2}$$

$$\frac{\mu^2}{2\sigma^2} = \frac{\eta_1^2}{4\eta_2^2} \cdot \frac{2\eta_2}{\eta_2(-1)} = -\frac{1}{4} \frac{\eta_1^2}{\eta_2}$$

1-2 1A

1 2A

2-3 next

✓ PINNs / deep Ritz method.
 ✓ Neural ODE.
 ✓ DDPM / Diffusion model