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# AI That Benefits Humanity Is All You Need

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## Abstract

Ever since I was young, I have been dreaming of building a platform that allows more people to access useful information and knowledge they need, breaking down information barriers. I was greatly inspired when I saw the mission Elon Musk assigned to xAI. On March 17, 2024, xAI was open-sourced.<sup>2</sup> I think it's time for me to take the first step and delve deep into the AI industry, and make my dream come true. And now, it's time to go!

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\*Thanks for *Mao Tse-Tung*, Elon Musk and Eminem. You guys gave me the drive to move forward.

<sup>2</sup>xAI is a new company working on building artificial intelligence to accelerate human scientific discovery.

## Part I

# Paper Reading

- 1 ImageNet Classification with Deep Convolutional Neural Networks (AlexNet)
- 2 Deep Residual Learning for Image Recognition (ResNet)

## Part II

# Mathematical Principles

- 3 Support Vector Machine
- 4 Normal Equation

To minimize  $\|\beta - Ax\|$  is to get the distance of  $\beta$  to  $W$ .

## 5 Why Gradient

If  $f$  is differentiable at  $(x_0, y_0)$ , then the directional derivative exists for any direction  $\vec{v} = (\cos \alpha, \sin \alpha)$ , and is given by:

$$\frac{\partial f}{\partial v}(x_0, y_0) = \frac{\partial f}{\partial x}(x_0, y_0) \cos \alpha + \frac{\partial f}{\partial y}(x_0, y_0) \sin \alpha$$

The derivation is:

$$\begin{aligned} \frac{\partial f}{\partial v}(x_0, y_0) &= \lim_{t \rightarrow 0^+} \frac{f(x_0 + t \cos \alpha, y_0 + t \sin \alpha) - f(x_0, y_0)}{t} \\ &= \lim_{t \rightarrow 0^+} \frac{\frac{\partial f}{\partial x}(x_0, y_0) \cos \alpha + \frac{\partial f}{\partial y}(x_0, y_0) \sin \alpha + o(t)}{t} \\ &= \frac{\partial f}{\partial x}(x_0, y_0) \cos \alpha + \frac{\partial f}{\partial y}(x_0, y_0) \sin \alpha \end{aligned}$$

The definition of gradient is  $\mathbf{grad} f(x_0, y_0) = f_x(x_0, y_0)\mathbf{i} + f_y(x_0, y_0)\mathbf{j}$

## References

References follow the acknowledgments.

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