

Assignment 11

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W1

Q: Mini-Batch SGD 的 batch size 要如何選擇?

- Gradient Noise Scale & 臨界批量：McCandlish et al., An Empirical Model of Large-Batch Training, 2018 (OpenAI 解釋文 + 論文)。提出以梯度雜訊尺度預測「最大有用 batch size」。
連結：[OpenAI 解說頁](#)、[arXiv 論文](#)
- 線性放大學習率 + warmup (大批量 ImageNet)：Goyal et al., Accurate, Large Minibatch SGD: Training ImageNet in 1 Hour, 2017。提出linear scaling rule 與 warmup，在 batch 高達 8192 時維持準確率。
連結：[arXiv PDF](#)。

W3

Q: Why do the weights in the construction get so large when the required error ϵ is very small?

1. Lipschitz 與權重範數/結構

- Virmaux, A., & Scaman, K. (2018). Lipschitz regularity of deep neural networks: analysis and efficient estimation. NeurIPS.
PDF：<https://papers.neurips.cc/paper/7640-lipschitz-regularity-of-deep-neural-networks-analysis-and-efficient-estimation.pdf>
arXiv：<https://arxiv.org/abs/1805.10965>
- 深度 ReLU 近似率 (ϵ \epsilonpsilon 與模型複雜度/常數量化)
Yarotsky, D. (2017). Error bounds for approximations with deep ReLU networks. Neural Networks, 94, 103–114.
arXiv：<https://arxiv.org/abs/1610.01145>
Journal 版摘要：
<https://www.sciencedirect.com/science/article/abs/pii/S0893608017301545>

W5

Q: Why are multivariate Gaussian contours ellipsoids determined by eigenvectors/eigenvalues of Σ ?

- Reference: Johnson, R. A., & Wichern, D. W. (2002). Applied Multivariate Statistical Analysis (6th ed.), Ch. 4 — 等密度橢球的主軸與 Σ 的特徵向量/特徵值對應 (掃描章節)。連結：[Johnson & Wichern Ch.4 PDF](#)。

Q: What breaks if Σ is not positive definite?

- Reference: Multivariate normal distribution (Wikipedia) — “Degenerate case”
小節： Σ 非滿秩/非正定時，對 Lebesgue 測度不具密度；需要用偽逆/偽行列式等處理。連結：[Wikipedia 條目](#)。

Q: Why does MLE use $1/n$ while the unbiased estimator uses $1/(n-1)$?

- Reference: Bessel's correction (Wikipedia) — 解釋為何 MLE 取 $1/n$ 有偏、而以 $1/(n-1)$ 校正可得無偏。連結：[Wikipedia 條目](#)。

W7

Q: Why do we train on multiple noise levels instead of one fixed σ ?

- Reference: Song, Y., & Ermon, S. (2019). *Generative Modeling by Estimating Gradients of the Data Distribution (Noise-Conditional Score Networks)*. NeurIPS.
連結：[論文 PDF](#)。

Q: How is DSM related to denoising autoencoders?

- Reference: Vincent, P. (2011). A Connection Between Score Matching and Denoising Autoencoders. Neural Computation. 連結：[期刊頁 \(摘要/全文\)](#)。

W8

Q: Why does score matching allow training without knowing the true data density $p(x)$?

- Hyvärinen, A. (2005). Estimation of Non-Normalized Statistical Models by Score Matching. JMLR, 6, 695–709.
(提出以匹配「分數函數」 $\nabla_x \log p(x)$ 的方式進行估計，目標可寫成只含模型可微項與邊界項，無需正規化常數與真實密度。) 連結：[JMLR PDF](#)。

W10

Q: Can we derive the same probability flow ODE form for higher-dimensional SDEs, and what complications arise when the diffusion term $g(x, t)$ becomes a matrix?

- Song, Y., Sohl-Dickstein, J., Kingma, D. P., Kumar, A., Ermon, S., & Poole, B. (2021). Score-Based Generative Modeling through Stochastic Differential Equations. ICLR — Appendix D derives the probability-flow ODE for the general multidimensional SDE with matrix diffusion, and discusses the resulting terms (e.g., involving $A=\sigma\sigma^T$) and their implications for likelihoods/sampling. 連結：[arXiv \(含附錄\)](#)