

Pruning and Quantizing MobileNet trained on DREAMT Dataset

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노기현

박정민

Why do we need to quantize / prune? → To achieve On-Device Inference

Processing Data on Your Device

After anonymization, processing data directly on your device adds another layer of privacy by keeping raw information local. Al sleep apps handle sensitive sleep data on your device, reducing the need to send it to external servers and lowering privacy risks.

Vatch • 6 mo. ago

umption with pulse sensor + how it affects sleep

= .

does it work at night when sleep analysis is active with all extras such as temperature and oxygen saturation informations. is the sleep data still accurate at all, I am afraid that I will leave it on "every 10 minutes" or completely active and that will take care of less hours of battery. Currently fighting very much with battery level and consumption. Has anyone more information about that? My Galaxy Watch Ultra is 2 Days old. Am I acting too fast against it, do I need to wait for AI analysing my battery first? I'm a little bit confused. The first 2 days with the new watch are horrible. I have no battery life at all: (. I turned lte off, on weather apps I put gps accuracy on "off", stress and heart rate monitoring manual. I don't have problems with my heart. But I love the sleep monitoring feature. It's one of the reasons I want it at night with accurate formations for the next morning. Anyone have more important informations about it, and how it works, or drains the

Why do we need to quantize / prune?

- → To achieve On-Device Inference
- → Model that can actually run in wearable devices "locally" (e.g. Smart Watch)



Goal

- Minimal Accuracy Decline
- Reduced Model Size & Parameters
- Improved Memory & Latency

About Dataset

Dataset for Real-time sleep stage Estimation using Multisensor wearable Technology



DREAMT: Dataset for Real-time sleep stage EstimAtion using Multisensor wearable Technology

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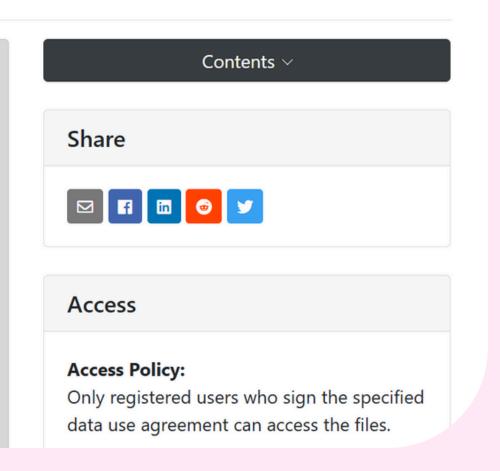
Wang, K., Yang, J., Shetty, A., & Dunn, J. (2025). DREAMT: Dataset for Real-time sleep stage EstimAtion using Multisensor wearable Technology (version 2.1.0). *PhysioNet*. RRID:SCR_007345. https://doi.org/10.13026/7r9r-7r24

Additionally, please cite the original publication:

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Please include the standard citation for PhysioNet: (show more options)

Goldberger, A., Amaral, L., Glass, L., Hausdorff, J., Ivanov, P. C., Mark, R., ... & Stanley, H. E. (2000). PhysioBank, PhysioToolkit, and PhysioNet: Components of a new research resource for complex physiologic signals. Circulation [Online]. 101 (23), pp. e215-e220. RRID:SCR_007345.



Baseline Model

MobileNet v2 adapted for 1D Time-Series

Calculate Acc / F1-score / Confusion Matrix

Pruning

torch.nn.utils.prune() 이용

- → weight sparsity (30%, 50%, 70%, ...)
- → Check the difference in Accuracy

Post-training dynamic quantization torch.quantization.quantize_dynamic 이용

- → convert to int8
- → check the model size/Accuracy



Comparing Acc, latency, and size

- Baseline model
- Pruned
- Quantized
- Pruned + Quantized

감사합니다