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Jan **<u>Wira</u>** Gotama Putra adalah PhD Candidate (JSPS Research Fellow) di Computational Linguistics laboratory, Artificial Intelligence Major, Department of Computer Science, Tokyo Institute of Technology. Sebelumnya, penulis mendapatkan gelar sarjana di jurusan Teknik Informatika, Institut Teknologi Bandung. Kemudian, melanjutkan magister di Tokyo Institute of Technology dengan pendanaan MEXT scholarship. Penulis memiliki pengalaman menulis makalah ilmiah pada bidang pemrosesan bahasa alami dan menerapkan teknik pembelajaran mesin untuk perusahaan IT (saat *internship*), e.g., *discourse processing*, klasifikasi teks, peringkasan teks, *automatic hashtag-ing*, dan *content filtering* (*moderation*). Penulis pernah mendapatkan *best paper award* di workshop internasional dan memenangkan lomba *data mining* (terbuka untuk publik) di Indonesia. Penulis masih terus mempelajari teknik *machine learning*. Buku ini adalah catatan yang ingin ia bagikan.

https://wiragotama.github.io/