

DAFTAR PUSTAKA

- Anand, A., & Dubey, Mr. S. (2022). CV Analysis Using Machine Learning. *International Journal for Research in Applied Science and Engineering Technology*, 10(5), 1316–1322. <https://doi.org/10.22214/ijraset.2022.42295>.
- Bender EM, Gebru T, McMillan-Major A, Shmitchell S. On the dangers of stochastic parrots: can language models be too big? In: Proceedings of the 2021 ACM Conference on Fairness, Accountability, and Transparency; 2021. p. 610–623.
- Bozkurt, A., and Sharma, R.C., Generative AI and *prompt* engineering: the art of whispering to let the genie out of the algorithmic world, Asian J. Distance Educ., 2023.
- Brown, T., Mann, B., Ryder, N., Subbiah, M., Kaplan, J. D., Dhariwal, P., Neelakantan, A., Shyam, P., Sastry, G., Askell, A., et al. (2020). Language models are few-shot learners. *Advances in neural information processing systems*, 33, 1877–1901.
- Chen, B., Zhang, Z., Langrené, N., & Zhu, S. (2023). *Unleashing the potential of prompt engineering in Large Language Models: a comprehensive review*. <http://arxiv.org/abs/2310.14735>
- Clavié, B., Ciceu, A., Naylor, F., Soulié, G., & Brightwell, T. (2023). *Large Language Models in the Workplace: A Case Study on Prompt Engineering for Job Type Classification*. <http://arxiv.org/abs/2303.07142>
- Cooper, G., Examining science education in ChatGPT: An exploratorand studand of generative artificial intelligence, *J. Sci. Educ. Technol.*, 32(3), pp. 444-452, 2023. DOI: <https://doi.org/10.1007/s10956-023-10039-and>.
- Chang, E.Y., *Prompting* large language models with the socratic method, in: 2023 IEEE 13th Annual Computing and Communication Workshop and Conference, CCWC 2023, 2023. pp. 351-360. DOI: <https://doi.org/10.1109/CCWC57344.2023.10099179>
- Clinton Gormley & Zachary Tong, *Elasticsearch*, "The Definitive Guide: A Distributed *real-time* search and analytics engine", O'Reilly, January 2015
- Davison, J., Feldman, J., & Rush, A. M. (2019). Commonsense knowledge mining from pretrained models. In Proceedings of the 2019 Conference on Empirical Methods in Natural Language Processing and the 9th International Joint Conference on Natural Language Processing (EMNLP-IJCNLP) (pp. 1173–1178).
- Eager, B., and Brunton, R., *Prompting* higher education towards AI-Augmented teaching and learning practice, *J. Univ. Teach. Learn. Pract.*, 20(5), 2023. DOI: <https://doi.org/10.53761/1.20.5.02>.
- Eandsenbach, G., The role of ChatGPT, generative language models, and artificial intelligence in medical education: a conversation with ChatGPT and a call for papers, *JMIR Med. Educ.*, 9, 2023. DOI: <https://doi.org/10.2196/46885>.
- Epstein, R.H., and Dexter, F., Variabilitand in large language Models'responses to medical licensing and certification examinations. comment on "How Does ChatGPT Perform on the United States Medical Licensing Examination? The Implications of Large Language Models for Medical Education and Knowledge Assessment", *JMIR Med. Educ.*, 9, 2023. DOI: <https://doi.org/10.2196/48305>.
- Gao, T., Fisch, A., & Chen, D. (2021). Making pre-trained language models better *few-shot* learners. In Proceedings of the 59th Annual Meeting of the Association for Computational Linguistics and the 11th International Joint Conference on Natural Language Processing (Volume 1: Long Papers) (pp. 3816–3830). Association for Computational Linguistics. doi: 10.18653/v1/2021.acl-long.295. URL: <https://aclanthology.org/2021.acl-long.295>.
- Giray, L., *Prompt engineering* with ChatGPT: a guide for academic writers, *Ann. Biomed. Eng.*, 2023. DOI: <https://doi.org/10.1007/s10439-023-03272-4>.

- Heston, T.F., and Khun, C., *Prompt engineering in medical education*, Int. Med. Educ., 2(3), pp. 198-205, 2023.
- Hu K (2023). ChatGPT sets record for *fasTest*-growing user base - analyst note. Reuters. <https://www.reuters.com/technology/chatgpt-sets-record-fasTest-growing-user-base-analyst-note-2023-02-01/>
- Huang, Z., et al. (2020). "Application of Artificial Intelligence in Human Resource Management."
- Jha, S., Jha, S.K., Lincoln, P., Bastian, N.D., Velasquez, A., and Neema, S., Dehallucinating large language models using formal methods guided iterative *prompting*, in: 2023 IEEE International Conference on Assured Autonomy (ICAA), IEEE, 2023. pp. 149-152.
- Jiang, Z., Xu, F. F., Araki, J., & Neubig, G. (2020). How can we know what language models know? Transactions of the Association for Computational Linguistics, 8, 423–438.
- Jiang, Z., Xu, F. F., Gao, L., Sun, Z., Liu, Q., Dwivedi-Yu, J., Yang, Y., Callan, J., & Neubig, G. (2023). *Active Retrieval Augmented Generation*. <http://arxiv.org/abs/2305.06983>
- Mitra, M. J. (2016), "The Rise of Elastic Stack", November. Available from: <https://doi.org/10.13140/RG.2.2.17596.03203>
- Mnih, V., Kavukcuoglu, K., Silver, D., Rusu, A. A., Veness, J., Bellemare, M. G., Graves, A., Riedmiller, M., Fidjeland, A. K., Ostrovski, G., Petersen, S., Beattie, C., Sadik, A., Antonoglou, I., King, H., Kumaran, D., Wierstra, D., Legg, S., & Hassabis, D. (2015). Human-level control through deep reinforcement learning. Nature, 518, 529–533.
- Lecler, A., Duron, L., and Soander, P., Revolutionizing radiology with GPT-based models: current applications, future possibilities and limitations of ChatGPT, Diagn. Interv. Imaging, 104(6), pp. 269-274, 2023. DOI: <https://doi.org/10.1016/j.diii.2023.02.003>.
- Liu, P., Yuan, W., Fu, J., Jiang, Z., Hayashi, H., Neubig, G.: Pre-train, *prompt*, and predict: A systematic survey of *prompting* methods in natural language processing. arXiv preprint arXiv:2107.13586 (7 2021)
- Lo, L.S., The art and science of *prompt engineering*: a new literacy in the information age, Internet Ref. Serv. Q., 2023. DOI: <https://doi.org/10.1080/10875301.2023.2227621>
- Lester, B., Al-Rfou, R., & Constant, N. (2021). The power of scale for parameter-efficient *prompt tuning*. In Proceedings of the 2021 Conference on Empirical Methods in Natural Language Processing (pp. 3045–3059).
- Lu, Y., Bartolo, M., Moore, A., Riedel, S., & Stenetorp, P. (2021). Fantastically ordered *prompts* and where to find them: Overcoming *few-shot prompt* order sensitivity. arXiv preprint arXiv:2104.08786.
- Ouyang, L., Wu, J., Jiang, X., Almeida, D., Wainwright, C.L., Mishkin, P., Zhang, C., Agarwal, S., Slama, K., Ray, A., et al.: Training language models to follow instructions with human feedback. arXiv preprint arXiv:2203.02155 (2022)
- OpenAI. (2023). *GPT-4 Technical Report*.
- Polak, M. P., & Morgan, D. (2024). Extracting accurate materials data from research papers with conversational language models and *prompt engineering*. Nature Communications, 15(1). <https://doi.org/10.1038/s41467-024-45914-8>.
- Poole, D. L., Mackworth, A. K., & Goebel, R. (1998). Computational intelligence: A logical approach. New York, NY: Oxford University Press.
- Qin, G., & Eisner, J. (2021). Learning how to ask: *Querying LMs with mixtures of soft prompts*. In Proceedings of the 2021 Conference of the North American Chapter of the Association for Computational Linguistics: Human Language Technologies (pp. 5203–5212).
- R. Vidhya, G. Vadivu, "Research Document Search Using *Elasticsearch*", Indian Journal of Science and Technology, Vol. 9(37), September 2016, DOI: 10.17485/ijst/2016/v9i37/102108.

- Rae, J.W., Borgeaud, S., Cai, T., Millican, K., Hoffmann, J., Song, F., Aslanides, J., Henderson, S., Ring, R., Young, S., et al.: Scaling language models: Methods, analysis & insights from training gopher. arXiv preprint arXiv:2112.11446 (2021)
- Reynolds, L., & McDonell, K. (2021). *Prompt programming for large language models: Beyond the few-shot paradigm*. In Extended Abstracts of the 2021 CHI Conference on Human Factors in Computing Systems (pp. 1–7).
- Russell, S. J., & Norvig, P. (2010). *Artificial intelligence: A modern approach* (3rd ed.). Upper Saddle River: Prentice Hall.
- Sanh, V., Webson, A., Raffel, C., Bach, S., Sutawika, L., Alyafeai, Z., Chaffin, A., Stiegler, A., Le Scao, T., Raja, A., et al. (2022). Multitask *prompted* training enables *zero-shot* task generalization. In The Tenth International Conference on Learning Representations.
- Schick, T., & Schütze, H. (2021). Exploiting cloze-questions for *few-shot text* classification and natural language inference. In Proceedings of the 16th Conference of the European Chapter of the Association for Computational Linguistics: Main Volume (pp. 255–269).
- Tenenbaum, J. B., Kemp, C., Griffiths, T. L., & Goodman, N. D. (2011). How to grow a mind: Statistics, structure, and abstraction. *Science*, 331, 1279–1285.
- Velásquez-Henao, J. D., Franco-Cardona, C. J., & Cadavid-Higuita, L. (2023). Prompt Engineering: a methodology for optimizing interactions with AI-Language Models in the field of engineering. *DYNA*, 90(230), 9–17. <https://doi.org/10.15446/dyna.v90n230.111700>
- Wang, J., Shi, E., Yu, S., Wu, Z., Ma, C., Dai, H., Yang, Q., Kang, Y., Wu, J., Hu, H., Yue, C., Zhang, H., Liu, Y., Li, X., Ge, B., Zhu, D., Yuan, Y., Shen, D., Liu, T., & Zhang, S. (2023). *Prompt Engineering for Healthcare: Methodologies and Applications*. <http://arxiv.org/abs/2304.14670>
- Webson, A., & Pavlick, E. (2021). Do *prompt*-based models really understand the meaning of their *prompts*? arXiv preprint arXiv:2109.01247.
- White, J. et al., A *prompt* pattern catalog to enhance *prompt engineering* with chatgpt, ArXiv Prepr. ArXiv230211382, 2023.
- Wu, T., Terrand, M., and Cai, C.J., AI chains: transparent and controllable Human-AI interaction band chaining large language model *prompts*, in: Conference on Human Factors in Computing Sandstems - Proceedings, 2022.DOI: <https://doi.org/10.1145/3491102.3517582>.
- Wu, Y., & Hu, G. (n.d.). *Exploring Prompt Engineering with GPT Language Models for Document-Level Machine Translation: Insights and Findings*. <https://github.com/wmt-conference/wmt23-news->
- Ye, J., Chen, X., Xu, N., Zu, C., Shao, Z., Liu, S., Cui, Y., Zhou, Z., Gong, C., Shen, Y., Zhou, J., Chen, S., Gui, T., Zhang, Q., & Huang, X. (n.d.). *A Comprehensive Capability Analysis of GPT-3 and GPT-3.5 Series Models*. <https://platform.openai.com/docs/model-index-for-researchers>
- Yong, G., Jeon, K., Gil, D., and Lee, G., *Prompt engineering for zero-shot and few-shot defect detection and classification using a visual-language pretrained model*, *Comput.-Aided Civ. Infrastruct. Eng.*, 38(11), pp. 1536-1554, 2023. DOI: <https://doi.org/10.1111/mice.12954>
- Young, T., Hazarika, D., Poria, S., & Cambria, E. (2018). Recent trends in deep learning based natural language processing. *IEEE Computational Intelligence Magazine*, 13,55–75.
- Zhou, C., Qiu, C., & Acuna, D. E. (2022). Paraphrase Identification with Deep Learning: A Review of Datasets and Methods (arXiv:2212.06933). arXiv. <https://doi.org/10.48550/arXiv.2212.06933>.