The List of Papers used to train the customised Chat-GPT

State of the art review paper:

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2. Ghoroghi, A., Rezgui, Y., Petri, I. et al. Advances in application of machine learning to life cycle assessment: a literature review. Int J Life Cycle Assess 27, 433–456 (2022). https://doi.org/10.1007/s11367-022-02030-3
3. Lotrič, Andrej, Mihael Sekavčnik, Igor Kuštrin, and Mitja Mori. “Life-Cycle Assessment of Hydrogen Technologies with the Focus on EU Critical Raw Materials and End-of-Life Strategies.” International Journal of Hydrogen Energy 46, no. 16 (2021): 10143–60. https://doi.org/10.1016/j.ijhydene.2020.06.190.

LCA- related papers:

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3. Wang, L., et al. "Artificial Intelligence in Product Lifecycle Management." *The International Journal of Advanced Manufacturing Technology*, vol. 114, 2021, 10.1007/s00170-021-06882-1
4. D'Amico, A., et al. "Artificial Neural Networks to Assess Energy and Environmental Performance of Buildings: An Italian Case Study." *Journal of Cleaner Production*, vol. 254, 2020, https://doi.org/10.1016/j.enbuild.2020.110220
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