

Resources

Other Fairness Libraries of Note

- [Aequitas](#)
- [AIF360](#)
- [Awesome Fairness in AI](#)
- [Dalex](#)
- [Fairlearn](#)
- [Fairness Comparison](#)
- [FAT Forensics](#)
- [ML Fairness Gym](#)
- [Themis ML](#)

Recorded References

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Stucchio, C. (2018, October). [AI Ethics, Impossibility Theorems and Tradeoffs](#) [Conference presentation]. Crunch Data Conference 2018, Budapest, Hungary. <https://www.youtube.com/watch?v=Zn7oWlhFffs>

Additional Resources and Tutorials

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