References on Federated Learning

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This is a list of references on Federated Learning (FL), a.k.a. Federated Machine Learning (FML). Since it is impossible for me to know every single reference on FL, please pardon me if I missed any of your work. If you want me to add your work to this list, you are welcome to send me a message. I will try my best to keep this list up-to-date. **Last updated: Dec. 07, 2019, by Dr. Yong Cheng.**

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