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Key challenges for delivering clinical impact with artificial intelligence, Kelly et al., 2019

The myth of generalisability in clinical research and machine learning in health care, Futoma et al., 2020

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Where I got some of them + some not read yet

Google “results for clinicians machine learning”

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Check ref 19 for regression guidelines

Check ref 22 for missing values for predictive models

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