

Supplementary Materials

List of publications used for the extractions of conditioning factors and predictors

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13. Goldfarb, A., & Lindsay, J. R. (2021). Prediction and Judgment: Why Artificial Intelligence Increases the Importance of Humans in War. *International Security*, 46(3), 7-50. https://doi.org/10.1162/isec_a_00425
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