

# IWML 2021 Reading List

## User research

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## Visual analytics

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# Interpretation and explainability

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