

Additional Readings for Stance Detection

Supplementary Material to SIGIR 2021 Tutorial Titled

“*Stance Detection: Concepts, Approaches, Resources, and Outstanding Issues*”

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