

Supplement information

32 Papers included in the meta-analysis: (Bai et al., 2021; Dou and Yang, 2018, 2017; Fang et al., 2020; Feng et al., 2020; Gerken et al., 2019; Granata, 2019; Granata and Di Nunno, 2021; Guo et al., 2019; Jung et al., 2011; Kafer et al., 2020; Li et al., 2018, 2021; Lu and Zhuang, 2010; Pang et al., 2021; Papale et al., 2015; Qin et al., 2005b, a; Safa et al., 2018; Shang et al., 2021; Van Wijk and Bouten, 1999; Vrugt et al., 2002; Vulova et al., 2021; Wang et al., 2021a, b; Xie et al., 2021; Xu et al., 2018; Yin et al., 2021; Zhang et al., 2021, 2020; Zhao et al., 2019)

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