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Iván Palomares Carrascosa

Large Group Decision Making

Creating Decision Support Approaches
at Scale

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*To my parents, Tomás and María,
and my sister Miriam.*

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