

# Past, Present, and Future of Statistical Science

Edited by

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**COPSS**

Committee of Presidents  
of Statistical Societies

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