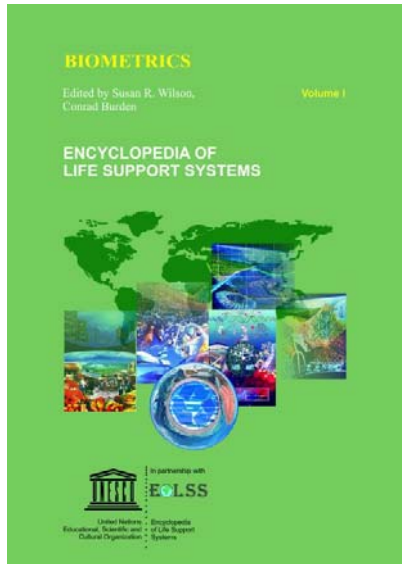


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BIOMETRICS



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 Susan R. Wilson, *Mathematical Sciences Institute, Australian National University, Australia*

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