

Supplementary material for: Spectral density regression for bivariate extremes

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§A: Asymmetric single-run experiment

Analogously to Fig. 3 in the paper, Fig. 1 displays the true spectral surface from the asymmetric Dirichlet predictor-dependent model detailed in Sect. 3.1, followed by spectral surface estimates for Configurations 1 and 2 using the double kernel estimator, where \mathbb{K}_b is chosen as a normal kernel with standard deviation b .

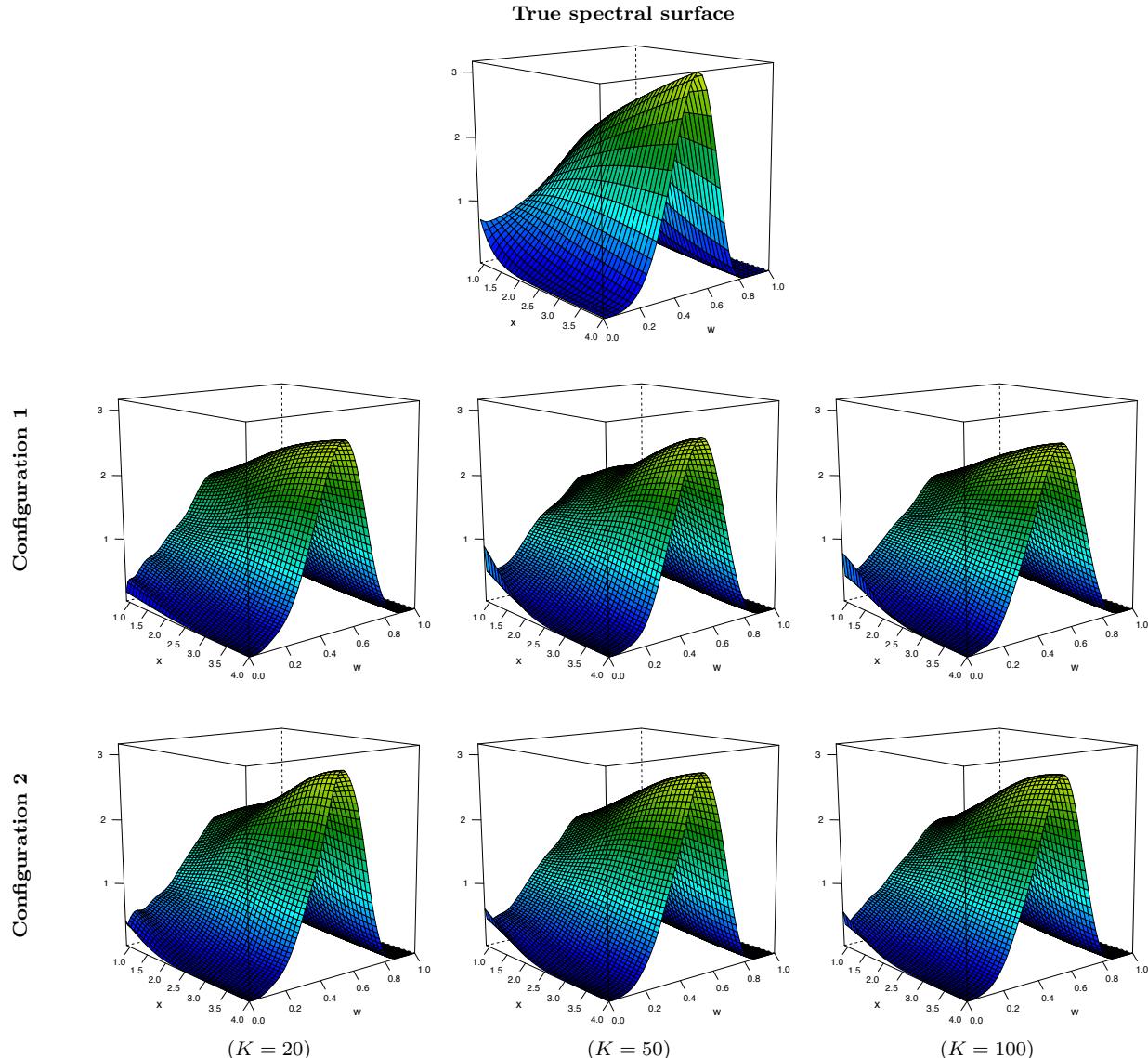


Fig. 1. On the top: true spectral surface from the asymmetric Dirichlet predictor-dependent model detailed in Sect. 3.1, followed by spectral surface estimates for Configurations 1 (above) and 2 (below).

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§B: Monte Carlo mean spectral surfaces

Monte Carlo mean spectral surfaces are plotted in Figures 2 and 3, where we average over 1000 simulation for $K \in \{20, 50, 100\}$ and Configurations 1 and 2 described in Sect. 3.1.

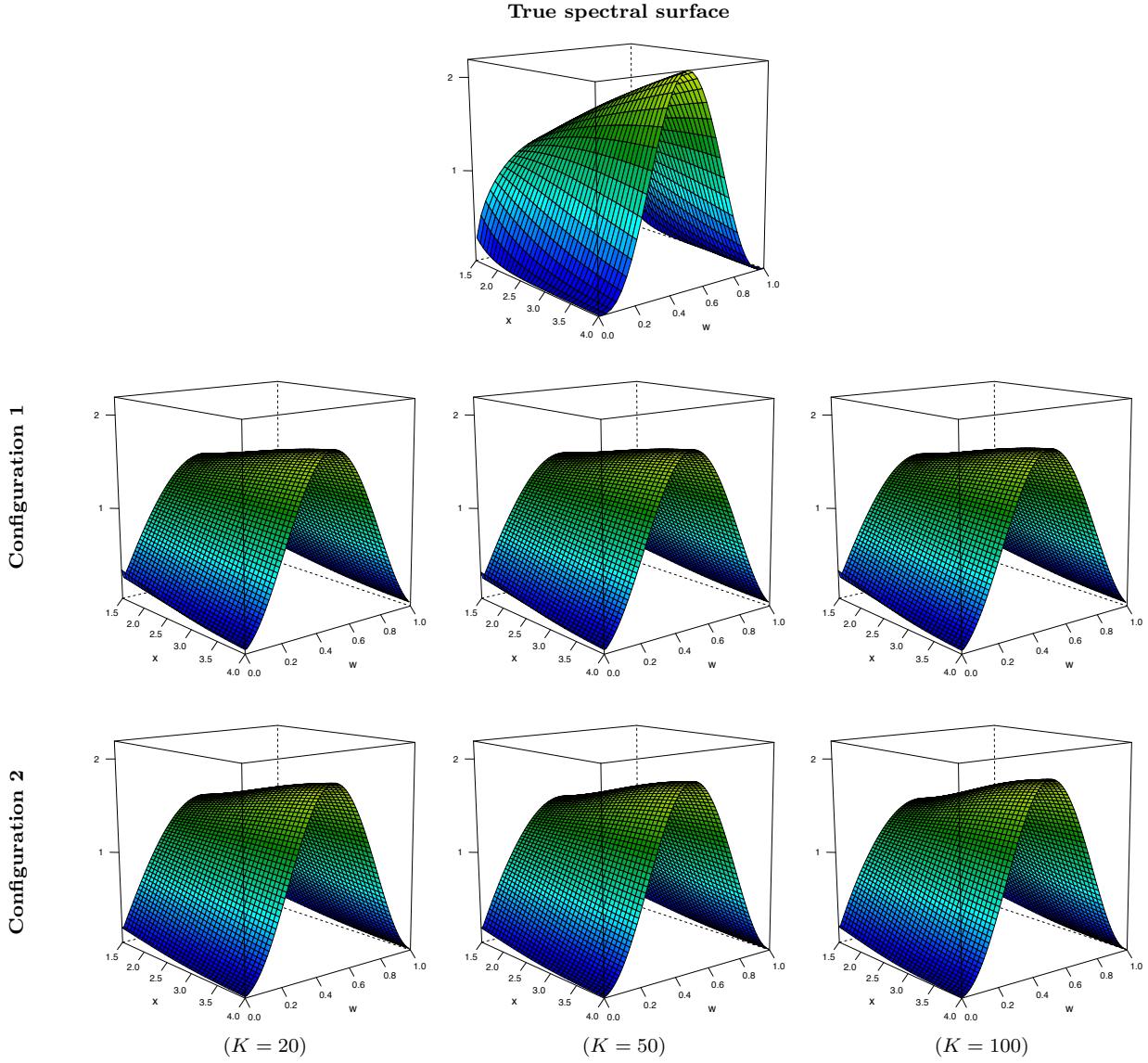


Fig. 2. On the top: true spectral surface from the symmetric Dirichlet predictor-dependent model detailed in Sect. 3.1, followed by Monte Carlo mean spectral surface estimates for Configurations 1 (above) and 2 (below).

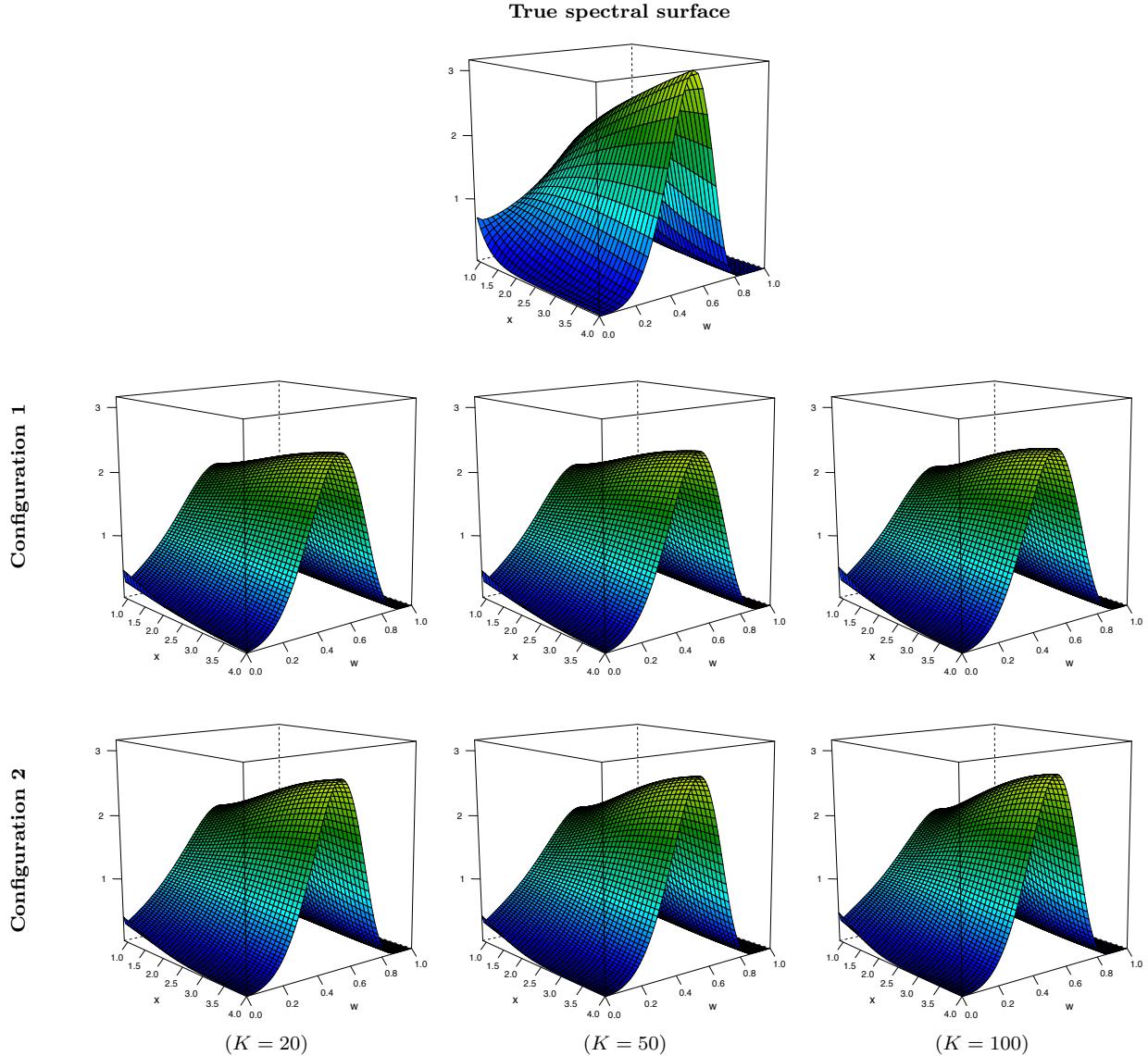


Fig. 3. On the top: true spectral surface from the asymmetric Dirichlet predictor-dependent model detailed in Sect. 3.1, followed by Monte Carlo mean spectral surface estimates for Configurations 1 (above) and 2 (below).