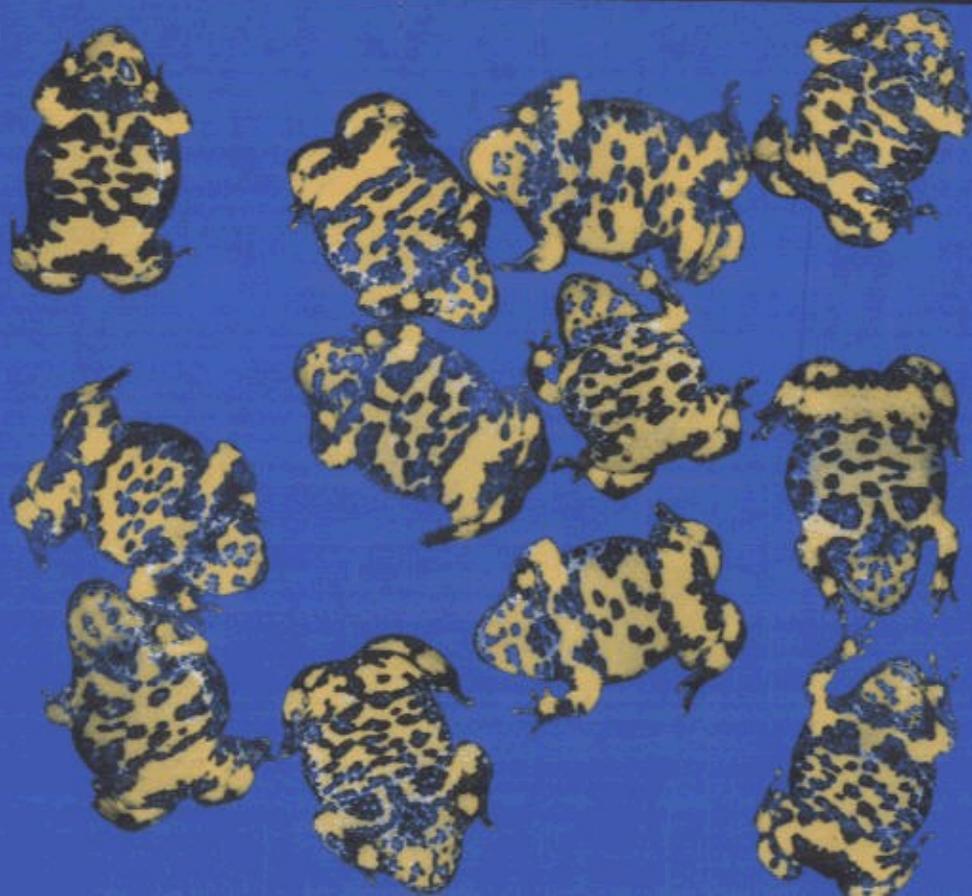


MODEL SELECTION AND MULTIMODEL INFERENCE
A Practical Information-Theoretic Approach

SECOND EDITION

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