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Remark: Paper 35 and 36 for ACM are the same. They are given as duplicate during search results. For correctness both are included. Same holds for paper 130 and 131.