

	key	annotation
Craig-1936	Craig-1936	NULL
Aroian-1947	Aroian-1947	NULL
Cochran-1952	Cochran-1952	NULL
Goodman-1960	Goodman-1960	NULL
Bradley-1978	Bradley-1978	robustness
Rubin-1976	Rubin-1976	NULL
Baron-Kenny-1986	Baron-Kenny-1986	mediation
Browne-1984	Browne-1984	NULL
Efron-1987	Efron-1987	NULL
Efron-1988	Efron-1988	NULL
James-Brett-1984	James-Brett-1984	mediation
Judd-Kenny-1981	Judd-Kenny-1981	mediation
Micceri-1989	Micceri-1989	NULL
Sobel-1982	Sobel-1982	NULL
Sobel-1986	Sobel-1986	NULL
Sobel-1987	Sobel-1987	NULL
Venzon-Moolgavkar-1988	Venzon-Moolgavkar-1988	NULL
White-1980	White-1980	NULL
Cohen-1988	Cohen-1988	NULL
NationalResearchCouncil-1982	NationalResearchCouncil-1982	data
Rubin-1987	Rubin-1987	Lib-Miss
Bollen-Stine-1990	Bollen-Stine-1990	NULL
Li-Raghunathan-Rubin-1991	Li-Raghunathan-Rubin-1991	missing,
Arbuckle-1996	Arbuckle-1996	NULL
Davison-Hinkley-1997	Davison-Hinkley-1997	bootstrap
Efron-Tibshirani-1993	Efron-Tibshirani-1993	NULL
Schafer-1997	Schafer-1997	NULL
Bauer-Preacher-Gil-2006	Bauer-Preacher-Gil-2006	NULL
Cheung-2009a	Cheung-2009a	NULL
Cheung-2009b	Cheung-2009b	NULL
Cheung-Lau-2007	Cheung-Lau-2007	NULL
CribariNeto-Souza-Vasconcellos-2007	CribariNeto-Souza-Vasconcellos-2007	NULL
Fritz-MacKinnon-2007	Fritz-MacKinnon-2007	NULL

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