

Evaluating Program Impacts: BIBLIOGRAPHY

Agodini, 2004, 'Are Experiments the Only Option? A Look at Dropout Prevention Programs,' The Review of Economics and Statistics, Vol. 86, No. 1, Pages 180-194. (the paper argues that unobserved factors often exert powerful influences on outcomes and these factors are often difficult to control for using statistical methods, I totally agree)

Almutairi, A., Gardner, G. and McCarthy, A. 2014, 'Practical Guidance for the Use of a Pattern-Matching Technique in Case-study Research: A Case Presentation', Nursing and Health Sciences, 16(2), pp. 239-244. (useful example of how to undertake pattern-matching in case studies)

Anderson, 2010, Proven Programs are the exception, not the rule, blog post: <http://blog.givewell.org/2008/12/18/guest-post-proven-programs-are-the-exception-not-the-rule/> (The author argues that examples of proven programs are rare. Their scarcity stems from two main factors: 1) the vast majority of social programs and services have not yet been rigorously evaluated, and 2) of those that have been rigorously evaluated, most, including those backed by expert opinion and less-rigorous studies, turn out to produce small or no effects, and, in some cases negative effects)

Angrist J. & Pischke J. 2015, Mastering Metrics: The paths from cause to effect. (easy to read guidance on what the authors see as the five most valuable econometric methods - random assignment, regression, instrumental variables, regression discontinuity designs, and differences in differences)

Antonakis, J. et al 2010, 'On Making Causal Claims: A Review and Recommendations', The Leadership Quarterly, Volume 21, Issue 6, 1086-1120. (this paper assessed 110 published impact studies in the field of leadership and found that most were significantly flawed, this paper illustrates the challenges of undertaking causal analysis using econometric methods with passive observational data and also includes better practice guidelines, highly recommended)

Arnold Ventures, 2016, Key Items to Get Right When Conducting Randomized Controlled Trials of Social Programs. (brief but helpful guidance)

Asher, 1983, Causal Modeling, Sage. (an introduction to causal/statistical modeling)

Asian Development Bank, 2006, Impact Evaluation: Methodological and Operational Issues. (a brief introduction, interesting discussion of common myths about impact studies)

AusAID, 2012, Impact Evaluation: A discussion paper for AusAID practitioners. (a brief introduction to the topic)

Barlow & Hersen, 1989, Single Case Experimental Designs, Pergamon. (an under-utilized approach in my opinion)

Bamberger, et al 2006, RealWorld Evaluation, Sage. (an excellent overview of how to undertake evaluations of development programs while facing various types of constraints, also includes a discussion of the most commonly used designs for evaluating the impact of development programs)

Barnett and Munslow 2014, Process Tracing The Potential and Pitfalls for Impact Evaluation in International Development, Institute of Development Studies. (a useful and balanced overview of the method, one of the better papers that I have come across on this topic)

Beach, D. 2019, 'Multi-Method Research in the Social Sciences: A Review of Recent Frameworks and a Way Forward', Government and Opposition, 0, 1–20. (a good discussion and comparison of variable-based vs case-based research strategies)

Becker, 2000, Discussion Notes: Causality, <http://web.uccs.edu/lbecker/Psy590/cause.htm> (a brief summary of different philosophical perspectives on causality)

Befani, B. 2012, Models of Causality and Causal Inference – Annex to Stern et al DFID Working Paper 38, UK Department for International Development. (a helpful summary)

Bennet, A. and George, A. 1997, Process Tracing in Case Study Research, MacArthur Foundation Workshop on Case Study Methods. (a comprehensive discussion of the strengths and weaknesses of this method, a challenging method to apply rigorously, highly recommended)

Bernal, J. Cummins, S. and Gasparri, A. 2018, 'The use of controls in interrupted time series studies of public health interventions', International Journal of Epidemiology, Volume 47, Issue 6. (a user friendly summary of the approach, recommended)

BetterEvaluation, 2018, Impact Evaluation, (resources for impact evaluation). Available on the internet: https://www.betterevaluation.org/en/themes/impact_evaluation

Blatter & Blume, 2008, 'In Search of Co-variance, Causal Mechanisms or Congruence? Towards a Plural Understanding of Case Studies', Swiss Political Science Review, 14(2): 315–56. (an excellent discussion of the difference between variable based and case based research, plus the difference between process tracing vs congruence/pattern matching approaches, highly recommended)

Block, 1999, Flawless consulting, Pfeiffer. (highly recommended, excellent chapter on working with resistant clients)

Bloom, Michalopoulos, Hill, & Lei, 2002, Can Nonexperimental Comparison Group Methods Match the Findings from a Random Assignment Evaluation of Mandatory Welfare-to-Work Programs? (the authors say no) Available free on the net at: <http://aspe.hhs.gov/pic/reports/acf/7541.pdf>

Bohte, J. and Meier, K. 2000, 'Goal Displacement: Assessing the Motivation for Organizational Cheating', Public Administration Review, 60, 173-182. (Discusses the motivations, incentives, and methods used by agencies to present an overly positive view of their performance. The lesson here is that accountability and incentive systems need to be designed with great care.)

Bonell, et al 2015, 'Dark Logic: Theorising the Harmful Consequences of Public Health Interventions', Journal Epidemiol Community Health, 69, 95-98. (an interesting paper on identifying and preventing the adverse effects / outcomes of public health interventions and the need to understand the underlying causal mechanisms)

Boruch, 2005, Randomized Experiments for planning and evaluation: A practical guide, Sage. (a good introduction to the topic)

Brady, 2002, Models of Causal Inference: Going Beyond the Neyman-Rubin-Holland Theory, Paper Presented at the Annual Meetings of the Political Methodology Group, University of Washington, Seattle, Washington. (paper reviews four of the more common theories of causality)

Brinkerhoff, 1991, Improving Development Program Performance: Guidelines for Managers, Lynne Rienner. (includes a discussion of the most common causes of performance problems in development programs, recommended)

Brown, E. and Tanner, J. 2019, Integrating Value for Money and Impact Evaluations, World Bank. (overview of the issues and opportunities)

Budhwani, S. and McDavid, J. 2017, 'Contribution Analysis: Theoretical and Practical Challenges and Prospects for Evaluators', Canadian Journal of Program Evaluation, 32.1, 1-24. (a good summary of CA's strengths and weaknesses and how it has evolved over time, recommended).

Campbell, 1969, 'Reforms as Experiments', American Psychologist, 24, p.409-429. (classic article on the politics of reforms and practical research designs)

Campbell & Stanley, 1963, Experimental and Quasi-experimental Designs for Research, Rand McNally. (the all-time classic text, discusses the strengths and weaknesses of various research designs for assessing program impacts, highly recommended)

Carroll, C. et al 2007, 'A conceptual framework for implementation fidelity', Implementation Science, 2:40 doi:10.1186/1748-5908-2-40. (only by understanding and measuring whether an intervention has been implemented with fidelity can evaluators and practitioners gain a better understanding of how and why an intervention works, and the extent to which outcomes can be improved)

Carter, Klein and Day, 1992, How organisations measure success: the use of performance indicators in government, Routledge. (offers a useful typology of different types of performance indicators and how they can be used/misused)

Christie and Alkin, 2019, 'Theorists' Models in Action: A Second Look, New Directions for Evaluation, no 163. (a useful illustration of how different theorists would approach the same evaluation task)

Coalition for Evidence-Based Policy, 2014, Which comparison-group (quasi-experimental) study designs are most likely to produce valid estimates of a program's impact? (excellent brief summary of the research evidence) Available free on the net.

Coly, A., & Parry, G. 2017, Evaluating Complex Health Interventions: A Guide to Rigorous Research Designs, AcademyHealth. (a useful overview of research designs for undertaking impact evaluations). Available free on the net at: https://www.academyhealth.org/sites/default/files/AH_Evaluation_Guide_FINAL.pdf

Cook, 2000, 'The false choice between theory-based evaluation and experimentation', in Rogers et al (eds) Program Theory in Evaluation: Challenges and Opportunities, Jossey-Bass. (excellent discussion of the main limitations of theory based methods for impact assessment)

Cook & Campbell, 1979, Quasi-experimentation, Houghton Mifflin. (excellent, contains a useful review of different theories of causality and how to test for causal relationships as well as the application of quasi-experiments for impact evaluations, highly recommended)

Cook, Shadish & Wong, 2008, 'Three Conditions under Which Experiments and Observational Studies Produce Comparable Causal Estimates: New Findings from Within-Study Comparisons', Journal of Policy Analysis and Management, 27, 4, 724-750. (this reference recommends regression-discontinuity designs, matching geographically local groups on pre-treatment outcome measures, and modeling a known selection process)

Corlazzoli and White, 2013, Measuring the Un-Measurable, DFID. (a useful resource).

Cracknell, B. E. 2000. Evaluating Development Aid: Issue, Problems and Solutions. Sage Publications, New Delhi. (interesting discussions of evaluating for learning vs for accountability plus the politics of evaluation)

Cracknell, B. E. 2001, 'Knowing is all: Or is it? Some reflections on why the acquisition of knowledge focusing particularly on evaluation activities, does not always lead to action'. Public Administration and Development, 31, 371-379.

Craig, et al 2018 (draft), Developing and Evaluating Complex Interventions, Medical Research Council. (some useful guidance for health programs)

Davidson, E. J. 2000. 'Ascertaining causation in theory-based evaluation'. In P.J. Rogers, T.A. Hacsí, A. Petrosino, and T.A. Huebner (eds.), "Program theory in evaluation: challenges and opportunities". New Directions in Evaluation, Number 87:17-26, San Francisco, CA. (provides an overview of various methods)

Davidson, E. J. 2004, Evaluation methodology basics: The nuts and bolts of sound evaluation, Sage. (suggests 8 techniques for causal inference)

Davis, 1985, The Logic of Causal Order, Sage. (worth a quick read)

Dawid, A. 2007, Fundamentals of Statistical Causality. (a good intro and description of the challenges of using passive observational research designs). <https://pdfs.semanticscholar.org/c4bc/ad0bb58091ecf9204ddb5db7dce749b0d461.pdf>

Deaton, A. 2010, What Can We Learn From Randomized Control Trials? Chicago: Milton Friedman Institute
http://mfi.uchicago.edu/events/20100225_randomizedtrials/index.shtml

Delahais, T and Toulemonde. J. 2017, ‘Making Rigorous Causal Claims in a Real-life Context: Has Research Contributed to Sustainable Forest Management?’ Evaluation, 23, 4, 370-388. (The authors conclude that substantiating and quantifying casual claims is not something that contribution analysis can deal with on its own. However by combining CA with other analytical approaches such as process tracing and realist evaluation it is possible to formulate rigorous contribution claims)

Dept of Finance, 1987, Evaluating Government Programs, Australian Government Publishing Service. (introductory, includes a useful table comparing different types of research designs)

Dept of Finance and Administration, 2006, Handbook of Cost-Benefit Analysis, Australian Government Publishing Service. (an easy to read introduction). This book is available free on the net at: http://www.finance.gov.au/FinFramework/fc_2006_01.html

Dimova, R. 2019, ‘A Debate that Fatigues...: To Randomise or Not to Randomise; What’s the Real Question?’, The European Journal of Development Research, Volume 31, Issue 2, pp 163–168. (argues that the RTC debate is increasingly fatiguing and tends to overemphasize methodological peculiarities at the expense of conceptual issues, the resolution of which is crucial for successful policy making)

Donaldson, Christie & Mark, 2009, What Counts as Credible Evidence in Applied Research and Evaluation Practice?, Sage. (reviews the debates on this topic, offers a range of perspectives)

Dong, N. et al 2017, “Can Propensity Score Analysis Approximate Randomized Experiments Using Pretest and Demographic Information in Pre-K Intervention Research?” Evaluation Review, XX(X). (The answer is no; propensity score analysis can sufficiently remove bias only if certain key assumptions are satisfied. In practice these assumptions are usually not tested for and when they are tested we generally find that the assumptions do not hold)

M. C. Elze, 2017, “Comparison of Propensity Score Methods and Covariate Adjustment, Evaluation in 4 Cardiovascular Studies”, Journal of the American College of Cardiology, vol 69, no 3. (PS methods are not necessarily superior to conventional covariate adjustment, and care should be taken to select the most suitable method)

Epstein & Klerman, 2012, “When is a Program Ready for Rigorous Impact Evaluation?” Evaluation Review, vol 36, pp. 373-399. (when it has a plausible theory of change in place and its implementation has been assessed as sound)

European Evaluation Society, 2007, Statement: The Importance of a Methodologically Diverse Approach to Impact Evaluation.

Evidence in Governance and Politics, 2020, Methods Guides. (a very useful website with technical guidance for conducting impact evaluations). See: <https://egap.org/list-methods-guides>

Evidence in Governance and Politics, 2020, 10 Things to Know about Causal Inference. (highly recommended). See: <https://egap.org/methods-guides/10-things-you-need-know-about-causal-inference>

Evidence in Governance and Politics, 2020, 10 Types of Treatment Effects You Should Know About. (highly recommended). See: <http://egap.org/methods-guides/10-types-treatment-effect-you-should-know-about>

Freedman and Collier, 2009, Statistical Models and Causal Inference: A Dialogue with the Social Sciences, Cambridge University Press. (argues that statistical techniques are seldom an adequate substitute for substantive knowledge of the topic, having a good research design, relevant data and undertaking empirical testing in diverse settings; I totally agree)

Fretheim A., Tomic O. 2015, ‘Statistical process control and interrupted time series: a golden opportunity for impact evaluation in quality improvement’, BMJ Qual Saf; 24:748–752. (helpful introduction to this topic)

Gaarder, M. 2019, A Commentary to ‘Bridging to Action Requires Mixed Methods, Not Only Randomised Control Trials’, The European Journal of Development Research, Volume 31, Issue 2, pp 169–173. (this paper puts the debates about RTCs into context, highly recommended)

Gates & Dyson, 2017, “Implications of the Changing Conversation About Causality for Evaluators”, American Journal of Evaluation, vol. 38, no. 1 (an introductory overview of issues for consideration plus six guidelines for evaluators seeking to make causal claims: (1) being responsive to the situation and intervention, (2) building relevant and defensible causal arguments, (3) being literate in multiple ways of thinking about causality, (4) being familiar with a range of causal designs and methods, (5) layering theories to explain causality at multiple levels; and (6) justifying the causal approach taken to multiple audiences)

Gao, X. et al 2019, ‘Evaluating Program Effects: Conceptualizing and Demonstrating a Typology’, Evaluation and Program Planning, 72, 88-96. (some interesting comments about collective impact and different types of program effects)

Gertler, P. et al 2010, Impact Evaluation in Practice, World Bank. It is available free on the net at: <http://documents.worldbank.org/curated/en/2011/01/13871146/impact-evaluation-practice>

Glazerman, Levy & Myers, 2002, Nonexperimental Replications of Social Experiments: A Systematic Review, Mathematica Policy Research Inc. (this research paper concludes that more often than not, statistical models do a poor job of estimating program impacts, highly recommended). This report is available free on the net at: <http://www.mathematica-mpr.com/publications/PDFs/nonexperimentalreps.pdf>

Gleason, et al 2018, “RD or Not RD: Using Experimental Studies to Assess the Performance of the Regression Discontinuity Approach”, Evaluation Review, 42,(1), 3-33. (paper concludes that RDs often provide accurate estimates of impacts although the results are sensitive to the manipulation of the assignment variable)

Glennerster & Takavarasha, 2013, Running Randomized Evaluations: A Practical Guide, Princeton University Press. (a comprehensive and handy guide to undertaking randomized impact evaluations of social programs)

Gilovich, 1991, How we know what isn't so: The fallibility of human reason in everyday life, Free Press, New York. (demonstrates how cognitive, social and motivational processes distort our thoughts, beliefs, judgments and decisions)

Goodman, L. et al, 2018, “Beyond the RCT: Integrating Rigor and Relevance to Evaluate the Outcomes of Domestic Violence Programs”, American Journal of Evaluation, vol. 39, no. 1 (argues that it is important to match the evaluation design to the nature of the intervention, in addition a more inclusive conceptualization of credible evidence is required)

Gouleta, M. et al, 2020, ‘Understanding the dynamic interinfluences of implementation processes: An illustration by multiple case studies’, Evaluation and Program Planning, <https://doi.org/10.1016/j.evalprogplan.2020.101798>. (a clear illustration of how implementation fidelity can affect the achievement of desired outcomes)

Gray, K. 2019, Making Causal Inferences from Observational Data, LinkedIn. (a helpful short summary of the challenges plus some useful references).

gsocialchange, 2017, How do you know whether your intervention had an effect (a website with a range of resources)
<https://sites.google.com/site/gsocialchange/cause>

Guba and Lincoln, 1989, Fourth Generation Evaluation, Sage. (the authors argue that 'cause and effect' do not exist except by imputation, a constructivist perspective; I don't agree with this approach but many 'qualitative' researchers do)

Gugerty & Summer 2018, “Ten Reasons Not to Measure Impact and What to do Instead”, Stanford Social Innovation Review, (argues that impact evaluations are only a good investment in the right circumstances, i.e. when matched to the evaluation question, it's feasible to undertake an impact evaluation, the program is stable, etc. Recommended)

Handley, M. et al, 2018, ‘Selecting and improving quasi experimental designs in effectiveness and implementation research’, Annual Review Public Health; 39:5–25. (a good summary of options and the merits of different approaches)

Hatry et al 1981, Practical Program Evaluation for State and Local Governments, The Urban Institute Press (a classic introductory text with an excellent discussion of when experiments are feasible and appropriate)

Hay, C. 2016, ‘Process tracing: a laudable aim or a high-tariff methodology?’ New Political Economy, 21 (5). pp. 500-504. ISSN 1356-3467 (explains the significant limitations of process tracing, I think the method has been over sold)

Hedström, P. 2009, ‘Studying Mechanisms To Strengthen Causal Inferences In Quantitative Research’, in Janet M. Box-Steffensmeier, Henry E. Brady, and David Collier (eds) The Oxford Handbook of Political Methodology. (Benefits of studying mechanisms: an understanding of the mechanisms at work can improve statistical inference by guiding the specification of the statistical models to be estimated; mechanism-based models can strengthen causal inferences by showing why, acting as they do, individuals bring about the social outcomes they do)

Hernán M. and Robins, J. 2020, Causal Inference: What If, CRC Press. (a book that aims to help health and social scientists generate and analyze data to make causal inferences that are explicit about both the causal question and the assumptions underlying the data analysis, comprehensive and very technical)

HM Treasury, 2020, Magenta Book Annex A - Analytical methods for use within an evaluation, (excellent summary of the strengths and limitations of various methods, recommended). Available free on the internet at: https://assets.publishing.service.gov.uk/government/uploads/system/uploads/attachment_data/file/877202/Magenta_Book_Annex_A_Analytical_methods_for_use_within_an_evaluation.pdf

Holland, P. 1986, 'Statistics and Causal Inference'. Journal of the American Statistical Association. Vol. 81 pp. 945-960.

Hopkins, A. et al, 2020, The Experimenter's Inventory: A Catalogue of Experiments for Decision-Makers and Professionals, Alliance for Useful Evidence, Nesta, London. (a useful plain English summary of the pros and cons of different research designs for impact evaluation, addresses common criticisms of RTCs, uses nonstandard terminology at times which can be a bit confusing)

Hughs & Hutchings, 2011, Can We Obtain the Required Rigour Without Randomisation? Oxfam GB's Non-Experimental Global Performance Framework, 3IE. (suggests alternative approaches for NGOs)

Illari, P. 2014, Causality: Philosophical Theory meets Scientific Practice, OUP Oxford. (an introduction to the philosophy of causality, useful for evaluators unacquainted with philosophy)

Impact and Innovation Unit, Government of Canada, 2019, Measuring Impact by Design: A guide to methods for impact measurement: (a new reference guide for those involved in the design, delivery, procurement or appraisal of impact measurement strategies in Canada, a helpful summary of different methods).

Informed Choices Network, 2019, A framework for thinking critically about claims, evidence, and choices. (useful resources). Available free on the internet at: <https://thatsclaim.org/>

Intrac 2017, Most significant Change, (a description of MSC plus a summary of strengths and weaknesses). Available free on the internet at: <https://www.intrac.org/wpcms/wp-content/uploads/2017/01/Most-significant-change.pdf>

Intrac, 2017, Impact Assessment, (a useful summary of issues to consider). Available free on the internet at: <https://www.intrac.org/wpcms/wp-content/uploads/2017/01/Impact-Assessment.pdf>

Intrac, Reporting Change, 2018, (a useful easy to read overview, recommended). Available free on the internet at: <https://www.intrac.org/wpcms/wp-content/uploads/2017/01/Reporting-change.pdf>

J-PAL, 2017, Impact Evaluation Toolkit. (a guide to impact evaluation methods)

J-PAL, 2020, Library of Research Resources. (with an emphasis on experiments)
https://www.povertyactionlab.org/research-resources?utm_source=newsletter&utm_medium=email&utm_campaign=aug20&view=toc

- Jabeen, S. 2018, 'Unintended Outcomes Evaluation Approach: A Plausible Way to Evaluate Unintended Outcomes of Social Development Programmes', Evaluation and Program Planning, 68, 262-274. (a good paper addressing one of evaluation's weak areas, highly recommended)
- Jacob, R. et al 2012, A Practical Guide to Regression Discontinuity, MDRC. (the title says it all, available free on the internet)
- Jimenez. E. 2019, Be careful what you wish for: cautionary tales on using single studies to inform policymaking, 3IE Blog, (explains why we need to be careful about generalising from single studies)
<https://www.3ieimpact.org/blogs/be-careful-what-you-wish-cautionary-theses-using-single-studies-inform-policymaking>
- Johns, M. et al 2012, 'Evaluating the NYC Smoke-Free Parks and Beaches Law: A Critical Multiplist Approach', American Journal of Community Psychology, DOI: 10.1007/s10464-012-9519-5. (a helpful explanation of this approach, recommended)
- Judd & Kenny, 1981, Estimating the Effects of Social Interventions, Cambridge. (heavy emphasis on statistical applications, for the enthusiast)
- Kahneman, D. 2013, Thinking, Fast and Slow; Farrar, Straus and Giroux. (discusses why cognitive biases are common across all aspects of our lives and why human beings are generally unable to accurately perceive causal relationships, highly recommended)
- Kazdin, A. 2011, Single-Case Research Designs, Oxford University Press. (an under-utilized approach in my opinion)
- Kenny, 2004, Correlation and causality, (a very technical book about analysing causal impacts using statistical models). This book is available free on the net at: <http://davidakenny.net/cm/cc.htm>
- King, J. 2018, OPM's Approach to Assessing Value for Money, Oxford Policy Management. (useful and practical guidance)
- Kotvojs, F. and Carolina Lasambouw, C. nd, MSC: Misconceptions, Strengths and Challenges. (a good summary). Available free on the internet at: <https://www.aes.asn.au/images/stories/files/conferences/2009/Papers/Kotvojs,%20Fiona%20-%20MSC.pdf>
- Kraft MA. 2019, "Interpreting Effect Sizes of Education Interventions", Educational Researcher; 49 (4) :241-253. (this tends to be a weak area in most published evaluations, the author presents guidelines for interpreting effect sizes that are applicable across the social sciences, recommended)
- Krauss, A. 2018, "Why All Randomized Controlled Trials Produce Biased Results", Annals of Medicine, 50:4. (highly recommended, RTCs are never conducted without some degree of bias and this paper explains why. The idea of a single study that provides the ultimate definitive answer is flawed. We slowly and steadily build up our knowledge base over time. Undertaking research is similar to an Easter egg hunt combined with a jigsaw puzzle with no instructions; we are putting a mosaic together piece by piece over time)
- LaLonde, 1986, "Evaluating the Econometric Evaluations of Training Programs with Experimental Data", American Economic Review, vol 76, pp. 604-620. (the classic article explaining why the econometric analysis of correlational research designs usually fail to achieve accurate estimates of program impact)

Lam, S. 2020, "Toward Learning from Change Pathways: Reviewing Theory of Change and Its Discontents", Canadian Journal of Program Evaluation doi: 10.3138/cjpe.69535 (this paper identifies 7 common problems with ToC based evaluations, makes some good points)

Lam & Valencia, 2019, 'Retrospective Pretest and Counterfactual Self-Report: Different or Same?', Journal of MultiDisciplinary Evaluation, Volume 15, Issue 33. (the evidence shows that people are very inaccurate judges of change over time)

Lance, P. et al 2014, How Do we Know If a Program Made a Difference? A Guide to Statistical Methods for Impact Evaluation. Chapel Hill, North Carolina: MEASURE Evaluation. (a useful overview of methods)

Langbein, 1980, Discovering Whether Programs Work, Goodyear. (good but technical)

Larson, 1980, Why government programs fail, Praeger. (The reasons: a faulty theory of change/strategy; poor implementation; a changing external environment; or the evaluation itself is faulty)

Ledford, J. 2018, "No Randomization? No Problem: Experimental Control and Random Assignment in Single Case Research", American Journal of Evaluation, vol. 39, no. 1. (an overview of the use of single subject designs for impact evaluation to assess changes in level, trend and variability)

Li, S and Liu, Y, 2020. Using big data to evaluate the impacts of transportation infrastructure investment: the case of subway systems in Beijing, 3ie Impact Evaluation Report 115, New Delhi: International Initiative for Impact Evaluation (3ie). Available at: <https://doi.org/10.23846/DPW1IE115>. (this report uses interrupted time series analysis combined with comparison groups and pattern matching, a good example of using multiple methods to arrive at a defensible conclusion a la Reynolds & West)

Light, P. 2014, A Cascade of Failures: Why Government Fails and How to Stop It, Centre for Effective Public Management at Brookings. (analysis of government failures in the USA, insightful for evaluators. The reasons: poor policy; inadequate resources; culture; structure; lack of leadership)

Ioannidis, J. 2005, 'Why Most Published Research Findings Are False', PLOS Medicine. (discusses a range of potential biases). Available free on the internet at: <https://journals.plos.org/plosmedicine/article/file?id=10.1371/journal.pmed.0020124&type=printable>

Manning, R Ian Goldman, I. and Licona, G. 2020, The Impact of Impact Evaluation, WIDER Working Paper 2020/20, United Nations University. (a discussion of various political aspects of impact evaluations)

Mark & Reichardt, 2004, 'Quasi-experimental and correlational designs: Methods for the real world when random assignment isn't feasible'. In Sansone, Morf and Panter, (eds), Handbook of methods in social psychology, (pp. 265-286), Sage. (useful introductory overview, recommended)

Mayne, J. 2008, Contribution analysis: An approach to exploring cause and effect, ILAC Brief 16. (a popular approach based on using program theory and performance indicators - shares similar strengths and weaknesses)

- Mayne, J. 2019, A Brief on Contribution Analysis: Principals and Concepts, available via LinkedIn. (explains what CA can and cannot do, although I disagree with his approach to assessing causality).
- Mayne, J. 2019, 'Revisiting Contribution Analysis', Canadian Journal of Program Evaluation, 34.2, 171-191. (an update on the method which continues to evolve over time becoming more rigorous and also more complex/resource intensive)
- Mayne, 2019, Assessing the Relative Importance of Causal Factors, Centre for Development Impact, Practice Paper. (written from the perspective of CA and hence this approach shares the same advantages and disadvantages, in my opinion path analysis is a much better methodology for comparing the strength of causal pathways)
- Mayo, D. (2018). Statistical Inference as Severe Testing: How To Get Beyond the Statistics Wars, Cambridge: Cambridge University Press. (essential reading for those interested in data analysis and what constitutes credible evidence)
- Masset, et al, 2019. Successful Impact Evaluations: Lessons from DFID and 3ie, Centre of Excellence for Development Impact and Learning. (achieving the goals of credibility, relevance and policy impact is very challenging)
- McMillan, 2007, Randomized Field Trials and Internal Validity: Not So Fast My Friend, (good overview of the limitations). Available free on the net at: <http://pareonline.net/pdf/v12n15.pdf>
- McMurry et al, 2015, "Propensity scores: Methods, considerations, and applications", Journal of Thoracic and Cardiovascular Surgery, 150:14-9. (the authors conclude that results from most of the examples of PS that they examined were not convincing due to methodological problems)
- Michalopoulos, 2004, 'Can Propensity-Score Methods Match the Findings from a Random Assignment Evaluation of Mandatory Welfare-to-Work Programs?' The Review of Economics and Statistics, Vol. 86, No. 1, Pages 156-179. (the answer: occasionally, but not consistently)
- Miles and Huberman, 1994, Qualitative data analysis, Sage. (contains examples of undertaking causal analysis with qualitative data, recommended)
- Mill, J. S. 1888, A System of Logic Ratiocinative and Inductive, 8th edn. New York: Harper and Brothers. (Mill proposes 3 criteria for testing causal relationships: association, temporal order; and non-spuriousness)
- Mohr, 1995, Impact Analysis for Program Evaluation, Sage. (an advanced discussion of research designs and impact analysis)
- Mohr, 1999, 'The Qualitative Method of Impact Analysis', American Journal of Evaluation, 20, 1, pp. 69-84. (useful introduction to the topic)
- Muller, J. 2018, The Tyranny of Metrics, Princeton University Press. (a useful discussion of the limits of measurement as a management tool)
- Meyer, B. D. 1995, 'Natural and quasi-experiments in economics', Journal of Business & Economics Statistics, 13(2), 151–161. (a useful paper, discusses internal validity from an econometric perspective)

Murnane and Willett, 2010, Methods Matter: Improving Causal Inference in Educational and Social Science Research, Oxford University Press. (not overly technical and includes many examples)

National Institute for Health Research, 2016, Assessing claims about treatments effects: Key concepts that people need to understand. (a useful summary). Available on the net at: <http://www.testingtreatments.org/key-concepts-for-assessing-claims-about-treatment-effects/?nabm=1>

Network of Networks on Impact Evaluation, 2009, Impact Evaluations and Development – NONIE Guidance on Impact Evaluations. (a review of the methods commonly used by development agencies). Available free on the net at: <http://www.worldbank.org/ieg/nonie/guidance.html>

Nisbett and Ross, 1985, Human Inference: Strategies and Shortcomings of Social Judgments, Prentice-Hall. (explains why we all struggle to accurately perceive causal relationships, basically people are terrible at this due to various ‘involuntary’ cognitive biases)

Noble, J. et al, 2020, Understanding Impact – Using your theory of change to develop a measurement and evaluation framework, NPC.

Norad, 2008, The Challenge of Assessing Aid Impact: A Review of Norwegian Evaluation Practice, (provides a number of examples of problematic impact evaluations along with various lessons for better practice). Available free on the net at: http://www.norad.no/default.asp?V_ITEM_ID=12314

Norgbey, E, 2016, ‘Debate on the Appropriate Methods for Conducting Impact Evaluation of Programs within the Development Context’, Journal of Multidisciplinary Evaluation, vol 12, issue 27. (argues that method choices should respond to contextual and situational aspects of the program, I agree)

Nutley, et al, 2013, What counts as Good Evidence? Alliance For Useful Evidence.

Nutt, 2002, Why Decisions Fail, Berrett-Koehler Publishers. (interesting review of why strategic decisions often fail, e.g. lack of consultation, poor analysis, faulty implementation,)

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Olsen & Orr, 2016, ‘On The ‘Where’ of Social Experiments: Selecting More Representative Samples to Inform Policy’, New Directions in Evaluation, no 152. (useful suggestions for improving the external validity of experiments through better sampling)

Orr, L. et al 2019, ‘Using the Results from Rigorous Multisite Evaluations to Inform Local Policy Decisions’, Journal of Policy Analysis and Management. (this is generally a difficult undertaking as the size of the local error is often equal or larger than the estimated effect size).

Patton, 1982, Practical Evaluation, Sage. (of all his books this one is my favorite, great chapters on data analysis and on preparing useful recommendations, highly recommended)

- Patton, 1987, How to use qualitative methods in evaluation, Sage. (excellent discussion of combining qualitative and quantitative methods)
- Patton, 1990, Qualitative Evaluation and Research Methods, Sage. (good all round reference, helpful description of different types of purposeful sampling)
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- Pearl, J. and Mackenzie, D. 2018, The Book of Why: The New Science of Cause and Effect, Basic Books, New York. (an advanced text for the enthusiast, challenging to read)
- Peck, (ed) 2016, "Social Experiments in Practice: The What, Why, When, Where and How of Experimental Design and Analysis", New Directions in Evaluation, no 152. (a good overview of various issues from an econometric perspective)
- Peck, L. 2017, When is Randomization Right for Evaluation? (offers principles for when experiments are appropriate). See: <http://abtassociates.com/Perspectives/March/When-Is-Randomization-Right-for-Evaluation>
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Posthumus, H. and Wanitphon, P. 2015, Measuring Attribution: a practical framework to select appropriate attribution methods, with cases from ALCP in Georgia, MDF in East Timor, Propcom Mai-Karfi in Nigeria and Samarth-NMDP in Nepal, Hans Posthumus Consultancy. (a useful introduction from the DCED perspective with some helpful graphical illustrations).

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Reinertsen, Bjørkdahl, and McNeill, 2017, Confronting the Contradiction- An Exploration into the Dual Purpose of Accountability and Learning in Aid Evaluation, SIDA. (their main conclusion is that, in practice, the dual purposes of accountability and learning leads to difficult trade-offs; there are tensions and sometimes direct contradictions between maximizing accountability vs learning)

Reschovsky, Heeringa and Colby, 2018, Selecting the Best Comparison Group and Evaluation Design: A Guidance Document for State Section 1115 Demonstration Evaluations, Mathematica Policy Research. (an excellent paper with a useful decision making flow chart, highly recommended). Available on the net at: <https://www.medicaid.gov/medicaid/section-1115-demo/downloads/evaluation-reports/comparison-grp-eval-dsgn.pdf>

Reynolds & West 1987, 'A multiplist strategy for strengthening nonequivalent control group designs', Evaluation Review, 11, 6, 691-714. (an excellent example of how to fix up a weak research design by adding additional features thereby improving your overall assessment of the program's impact, a classic article and highly recommended)

Rogers et al 2000, 'Program Theory in Evaluation: Challenges and Opportunities', New Directions for Evaluation, No. 87. (a series of papers on the strengths and weaknesses of using program theory to assist with causal analysis)

Rohrer, J. 2018, 'Thinking Clearly About Correlations and Causation: Graphical Causal Models for Observational Data', Advances in Methods and Practices in Psychological Science, Vol. 1(1) 27–42. (this paper argues that it is very difficult to undertake causal analysis with observational data although graphical methods can help, highly recommended)

Roodman, 2008, Through the Looking Glass, and What OLS Found There: On Growth, Foreign Aid, and Reverse Causality, Working Paper 137, Center for Global Development. (discussion of assessing the impact of foreign aid)

Rosenbaum, P. 2019, Observation and Experiment: An Introduction to Causal Inference, Harvard University Press. (contains interesting examples and is quite insightful but it is not an easy read)

Rosling, H. 2018, Factfulness: Ten Reasons We're Wrong About the World--and Why Things Are Better Than You Think, Sceptre. (highly recommended for its real-world examples of outcome trajectories)

Rossi, Lipsey & Freeman 2003, Evaluation – A Systematic Approach, Sage. (recommended, includes an excellent discussion of different types of research designs and when to use each of them)

Rothman & Greenland, 2005, 'Causation and Causal Inference in Epidemiology', American Journal of Public Health, Vol 95, No. S1.

Rubin, 2008, "For Objective Causal Inference, Design Trumps Analysis", Annals of Applied Statistics, vol 2, pp. 808-840. (the title says it all and I agree with this position)

Salkind, N. J. (ed) 2010, Encyclopedia of Research Design, Sage. (very comprehensive at 1800 pages, and expensive)

Sauerbrei, W., Perperoglou, A., Schmid, M. et al. 2020, 'State of the art in selection of variables and functional forms in multivariable analysis—outstanding issues' Diagnostic and Prognostic Research 4, 3. <https://doi.org/10.1186/s41512-020-00074-3>. (multivariate analysis requires both advanced technical skills and expert subject matter knowledge, when undertaking observational research a number of subjective choices about analytical strategies are required and these can generate quite different results).

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Scriven, M. 1976, 'Maximizing the power of causal investigations: The modus operandi method'. In Gene V. Glass (ed.) Evaluation studies review annual, Volume 1, 101-118, Beverly Hills, CA: Sage Publications. (in my view Scriven's argument that undertaking an impact evaluation is similar to a detective investigating a murder is simply incorrect)

Shadish, Clark, & Steiner, 2008, 'Can nonrandomized experiments yield accurate results? A randomized experiment comparing random and nonrandom assignments;', Journal of the American Statistical Association, 103(484), pp. 134-1343. (yes, provided that all key variables are observed and we have good covariates to facilitate adjustment)

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Shadish, Cook and Campbell, 2002, Experimental and Quasi-Experimental Designs for Generalized Causal Inference, Houghton Mifflin. (a classic advanced text, recommended)

Shadish, Cook, and Houts, 1986, 'Quasi-Experimentation in a Critical Multiplist Mode', in W. M. K. Trochim (ed.). 'Advances in Quasi-Experimental Design and Analysis', New Directions for Program Evaluation, no. 31. San Francisco: Jossey-Bass. (in my view critical multiplism has a lot to offer when undertaking impact evaluations)

Shadish, Cook and Leviton, 1991, Foundations of Program Evaluation, Sage. (the final chapter contains an excellent summary of evaluation theory in relation to program design, evaluation practice, and theory of use, highly recommended)

Social Programs that Work, 2020, <https://evidencebasedprograms.org/>
(This site seeks to identify those social programs shown in rigorous studies to produce sizable, sustained benefits to participants and/or society, so that they can be deployed to help solve social problems)

Somers, M. et al, 2013, The Validity and Precision of the Comparative Interrupted Time Series Design and the Difference-in-Difference Design in Educational Evaluation, MDRC. (technical but interesting)

Spector, 1981, Research Designs, Sage. (a basic introduction)

St. Clair, T., Cook, T. and Hallberg, K. 2014, 'Examining the Internal Validity and Statistical Precision of the Comparative Interrupted Time Series Design by Comparison With a Randomized Experiment', American Journal of Evaluation, vol. 35, no 3, pp. 311-327, 2014. (CITS often provide accurate estimates but are sensitive to choices in design parameters)

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Streiner & Norman, 2012, 'The Pros and Cons of Propensity Scores', CHEST Journal, Volume 142, Issue 6, Pages 1380–1382, DOI: <https://doi.org/10.1378/chest.12-1920> (PSM's apparent simplicity masks a number of rather arbitrary statistical assumptions and tradeoffs)

Stuart, E. A., & Rubin, D. B. 2008, 'Matching with multiple control groups with adjustment for group differences'. Journal of Educational and Behavioral Statistics, 33(3), 279–306. (an illustration of this method)

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The International Campbell Collaboration, 2020, <https://campbellcollaboration.org/> (this site offers a registry of systematic reviews of evidence on the effects of interventions in the social, behavioral, and educational arenas)

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Treasury Board of Canada, no date, Program Evaluation Methods: Measurement and Attribution of Program Results, (useful overview). It is available free on the net at: http://www.tbs-sct.gc.ca/eval/pubs/meth/pem-mep_e.pdf

Trochim, 1984, Research designs for program evaluation: The regression discontinuity approach, Sage. (excellent method for evaluating impacts where entry into the program depends upon meeting a numerical eligibility criterion, e.g. income less than X, academic grades more than Y)

Trochim, 1989, 'Outcome Pattern Matching and Program Theory'. Evaluation and Program Planning, Vol. 12:355-366. (an interesting alternative for impact evaluations using case studies and pattern matching)

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U.S. Department of Education, 2003, Identifying and Implementing Educational Practices Supported by Rigorous Evidence: A User Friendly Guide. (a useful summary of what counts as good evidence in evaluation, this publication caused some controversy in the USA)

U.S. Department of Education, 2008, Technical Methods Report: Guidelines for Multiple Testing in Impact Evaluations.

USAID 2018, Guide for planning long-term impact evaluations. (useful introductory guidance)

Wauters and Beach, 2018, 'Process Tracing and Congruence Analysis to Support Theory Based Impact Evaluation', Journal of Evaluation. (this article illustrates the difference between process tracing and congruence analysis and their relative advantages).

Weisburd, D. 2010, 'Justifying the use of non-experimental methods and disqualifying the use of randomized controlled trials: challenging folklore in evaluation research in crime and justice', Journal of Experimental Criminology, volume 6, pages 209–227. (The key limitation of passive observational impact evaluation methods is that they require an assumption that all confounding factors related to treatment are identified in the statistical model being applied. The author explains why this assumption is so critical and challenges what he describe as “folklores” that are used to justify the use of non-randomized studies despite this major limitation).

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West, S. and Thoemmes, F. 2010, “Campbell’s and Rubin’s Perspectives on Causal Inference”, Psychological Methods, Vol. 15, No. 1, 18–37. (Campbell’s approach tends to be used in education and psychology and Rubin’s in economics, medicine and statistics, they are actually complementary)

Weyrauch, V. and Langou, G. D. 2011, Sound Expectations: From Impact Evaluations to Policy Change, 3ie Working Paper 12. London: 3ie. It is available free on the net at: http://www.3ieimpact.org/3ie_working_papers.html.

What Works Clearinghouse, 2017, Standards Handbook. (A good but very technical set of standards for impact evaluation. The What Works Clearinghouse identifies existing research on education interventions, assesses the quality of this research, and then summarizes and disseminates the evidence from studies that meet the Clearinghouse’s standards)

White, H. 2010, ‘A Contribution to Current Debates in Impact Evaluation’, Evaluation, 16(2) 153-164. (a useful summary of some of the issues and controversies)

White, H. And Raitzer, D. 2017, Impact Evaluation of Development Interventions – A Practical Guide, Asian Development Bank. (a useful resource with an econometric flavour). It is available free on the net at: <https://www.adb.org/publications/impact-evaluation-development-interventions-practical-guide>

Williams, R. 2015, Logic of Scientific Inference/ What is Causality? University of Notre Dame, <https://www3.nd.edu/~rwilliam/>. (a concise overview, explains Mill’s 3 causal criteria)

Winship, C. and Morgan S. 1999. “The estimation of causal effects from observational data”, Annual Review of Sociology. 25: 659-706. (a useful summary of the field; the author’s overall conclusion: the challenges of estimating causal effects with observational data are often formidable, the authors tend to favour longitudinal methods)

Winston, J. 1993, ‘Performance indicators: Do they perform?’ Evaluation News and Comments, 2(3), 22-39. (this paper outlines a number of concerns with how KPIs get used)

Wong, Steiner & Anglin, 2018, “What Can Be Learned from Empirical Evaluations of Nonexperimental Methods?”, Evaluation Review, 42(2), 147-175. (a lot, this paper provides a summary of the key issues)

Woolcock, M. 2019, Reasons for Using Mixed Methods in the Evaluation of Complex Projects, CID Faculty Working Paper No. 348, Centre for International Development at Harvard University. (a helpful overview of the topic)

Work-Learning Research, 2019, Levels of Evidence for the Learning Profession. (a useful summary as to what counts as better evidence). See: <https://www.worklearning.com/2019/11/20/levels-of-evidence-for-the-learning-profession/>

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Yang and Hendra, 2018, "The Importance of Using Multiple Data Sources in Policy Assessments: Lessons from Conditional Cash Transfer Programs in New York City", Evaluation Review, 1-25. (this paper argues that it is important to triangulate data sources in order to reach accurate conclusions about program effects)

Yeaton and Thompson, 2016, "Transforming the Canons of John Stuart Mill from Philopshy to Replicative, Empirical Research: The Common Cause Design", Journal of Methods and Measurement in the Social Sciences, Vol. 7, No . 2, p 122-143. (discussion of the strengths and weakensses of this little known approach).

Yin, 2000, 'Rival Explanations as an Alternative to Reforms as Experiments', in Bickman (ed) Validity and Social Experimentation, Sage. (good review of how to identify and test rival explanations when evaluating reforms or complex social change)

Yin, 2003, Applications of Case Study Research, Sage. (very good reference, includes advice on undertaking causal analysis using case studies and pattern matching - in my view this is preferable to process tracing)

Young, J. and Mendizabal, E. 2009, Helping researchers become policy entrepreneurs, ODI Briefing Papers 53. London: ODI <http://www.odi.org.uk/resources/download/1127.pdf> (guidance on the research:policy interface)

Zhao, Y. 2018, "What Works May Hurt: Side Effects in Education, Journal of Educational Change, DOI10.1007/s10833-016-9294-4. (If an intervention can potentially help it can also potentially harm. Unintended side effects are inseparable from intended effects – both are outcomes from the same intervention; highly recommended)

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