

Technological forecasting related to the energy sector: a scientometric overview

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Abstract. Scientometric review of trends and key points of technological forecasting related to the energy sector is carried out in this study. Using co-keyword, co-citation techniques to analyze a set of research and review articles indexed in the Scopus database, clustered networks were built to understand content relationships and research topic evolution within the 2000-2019 period. This study provides an overview of future-oriented research efforts and trends in the energy technology knowledge domain.

Keywords: energy technology, technological forecasting, scientometric analysis, visual analytic, VOSviewer, CiteSpace.

1. Introduction

Technology forecasting is usually determined as decisive and systematic attempts to anticipate and understand the potential direction, rate, characteristics, and effects of technological changes, especially invention, innovation, adoption, and use [1]. In [2] the group of experts systematizes methods and forms of technology forecasting within a future-oriented technology analysis framework. They distinguish several overlapping forms of technology forecasting such as:

- technology monitoring, watch, alerts (gathering and interpreting information);
- competitive intelligence (converting that information into usable intelligence);
- technology forecasting (anticipating the direction and pace of changes);
- technology roadmapping (relating anticipated advances in technologies and products to generate plans);
- technology assessment (anticipating the unintended, indirect, and delayed effects of technological changes);
- technology foresight (effecting development strategy, often involving participatory mechanisms).

In recent decades, the works [3] and [4] review the families of technology forecasting methods, its relationships, and applications. Nevertheless, there are no general overviews of technology forecasting evolution applied to the energy sector. This research tries to investigate the impact of energy technology forecasting in the scientific literature.

The energy technology forecasting concept is not always used to imagine prospects and the coming advances in the energy area. Many works anticipating future energy technologies use “technological change” or widely discussed “energy transition” toward sustainable development by transitioning from fossil-based to zero-carbon energy resources [5]. So these concepts should be additionally involved in the consideration.

The main goals of this study:

- summarizing the recent existing research efforts on energy technology forecasting;
- helping to systematically understand the co-citation documents, term clusters, and keywords clusters, as well as the knowledge pattern of energy technology forecasting;
- quantitative estimation of the status quo and development trend of energy technology forecasting;
- visualization of the research landscape of technology forecasting in the energy area.

2. Methodology, data, and tools

The methodology of the study is a scientometric analysis joint with supporting visualization to provide an in-depth understanding of the research structure and trending topics in energy technology forecasting. The scientometric analysis is a well-established technique to construct a knowledge map of the specific area over a large massive dataset of scientific literature. An example of a scientometric review of global research on

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sustainability and sustainable development can be found in [6]. General workflow of scientometric analysis includes several sequential steps:

1. Publications data retrieval related to a specific problem or knowledge area.
2. Data cleaning manually or automatically to remove irrelevant publications.
3. Scientometric quantitative analysis applying various metrics like betweenness centrality, burst strengths to construct different co-occurrence networks. The network examples are co-authorship network, co-word network, co-terms network, co-citations network, and others. Further cluster analysis over the constructed networks is also a part of the scientometric approach.
4. Knowledge domain visualization and in-depth analysis to obtain status-quo of research, discover emerging trends, hidden interrelations, and other valuable outputs.

In the study, the Scopus database was selected as the most comprehensive and easy-to-use data source. A search in the database was carried out using the base word “energy” and a specific set of additional words related to concepts of future-oriented technology analysis. The last concept has fuzzy semantics and includes such terms as “technology forecasting”, “technology foresight”, “technology monitoring”, “technology roadmapping”, “technology trend”, “technology assessment”, “technology change”, “technology transition” and so on. Symbol “*” is inserted instead of the end of some words to satisfy a fuzzy search. The publications with the language “English” and document type as “Article”, “Review” from reviewed and trusted journals were selected. We consider the period 2000–2019 when the rapid growth of publications in the Scopus database is observed.

The final query text inserted in the bar of “Advanced search” of the Scopus search engine is presented below.

TITLE-ABS-KEY (energy AND ("technol* forecast*" OR "forecast* technol*" OR "technol* trend*" OR "technol* monitoring" OR "technol* chang*" OR "technol* transit*" OR "technol* transform*" OR "emerging energy technol*" OR "technol* assess*" OR "technol* roadmap*" OR (technolog* AND "future prospect*"))) AND (LIMIT-TO (DOCTYPE , "ar") OR LIMIT-TO (DOCTYPE , "re")) AND (LIMIT-TO (LANGUAGE , "English"))

To avoid including irrelevant documents, for example from medical science, the search results were filtered to remove the subject areas far from “Energy” like “Medicine”, “Nursery”, “Computer Science”, “Arts and Humanities”, etc. On the other hand, since the “energy” is a multidisciplinary topic, such subject categories as “Engineering”, “Chemistry”, “Environmental Science”, “Social Science”, “Material Science” and so on also remain under consideration.

The search with this query gives 3448 articles. Fig.

1 presents the document statistics by years, countries, and sources provided by the standard Scopus tool.

To investigate semantic content, key topics, and its corresponding interrelations the two scientometric techniques were used in this study, namely, co-citation analysis and co-term (keyword) analysis.

In this paper, two software tools are used for scientometric analysis. First, VOSviewer software pays special attention to displaying large bibliometric maps in an easy-to-interpret way [7]. Another one is CiteSpace, which is a very powerful and extremely featured application for analyzing and visualizing co-citation scientific networks [8]. The software developed by Chaomei Chen has rich possibilities to identify the emerging trends and general points in a specific domain.

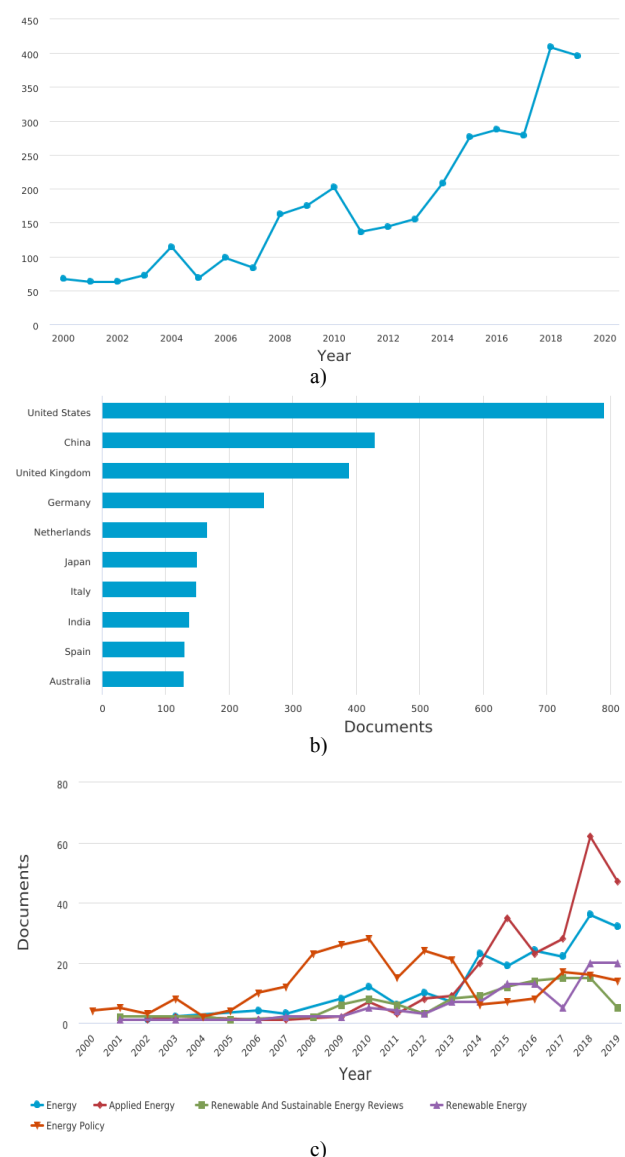


Fig. 1. The number of articles on energy technology forecasting in the Scopus database: (a) during period 2000–2019, (b) country distribution, (c) most productive journals.

3. Results of scientometric analysis

Initially, the co-occurrence network based on the article’s keywords was generated with VOSViewer. General terms like “article”, “review”, “technology” were excluded from the keywords list summarized from all articles. The visualization of the clustered network graph is shown in Fig. 2.

Several colored clusters found over the keywords network reflect the main research topics of two recent decades. Technological change implies the advancement of renewable energy resources such as wind energy and solar energy (mixed violet and blue clusters). The red cluster contains another class of renewable energy technologies based on biomass. The bioenergy cluster is linked to renewable solar and wind technologies through hydrogen technologies including fuel cells. The green cluster presents energy policy issues including planning and market development for renewable energy sources on the one hand. On the other hand, climate change and emission control issues are directly related to environmental protection and carbon dioxide emissions topics located in the yellow cluster. Such general terms as

energy efficiency, energy conversion, energy consumption, performance assessment, electric power system development, and other issues remain important research topics.

Generating a co-citation network using CiteSpace software with default parameters is the next step of the analysis. For the correct construction of the co-citation network, the publications of the years preceding 2000 were also included to consider previous research impact. Co-citation network in Fig. 3 presents an evolution of technology forecasting research from 2000 to 2019 years. It’s observed that the presented co-citation graph becomes sparser during the last decade 2010-2020.

The list of top 47 papers having the strongest citation burst is shown in Fig. 4. These papers are sorted by start year of burst to show the dynamics of the “hottest” documents and its corresponding topics along with the considered period. The main theme of these papers is a discussion about appropriate technological changes as responses on the global problems of climate change and sustainable development of the world economy. Types of almost all highly cited works are reviews, surveys or theory foundation books.

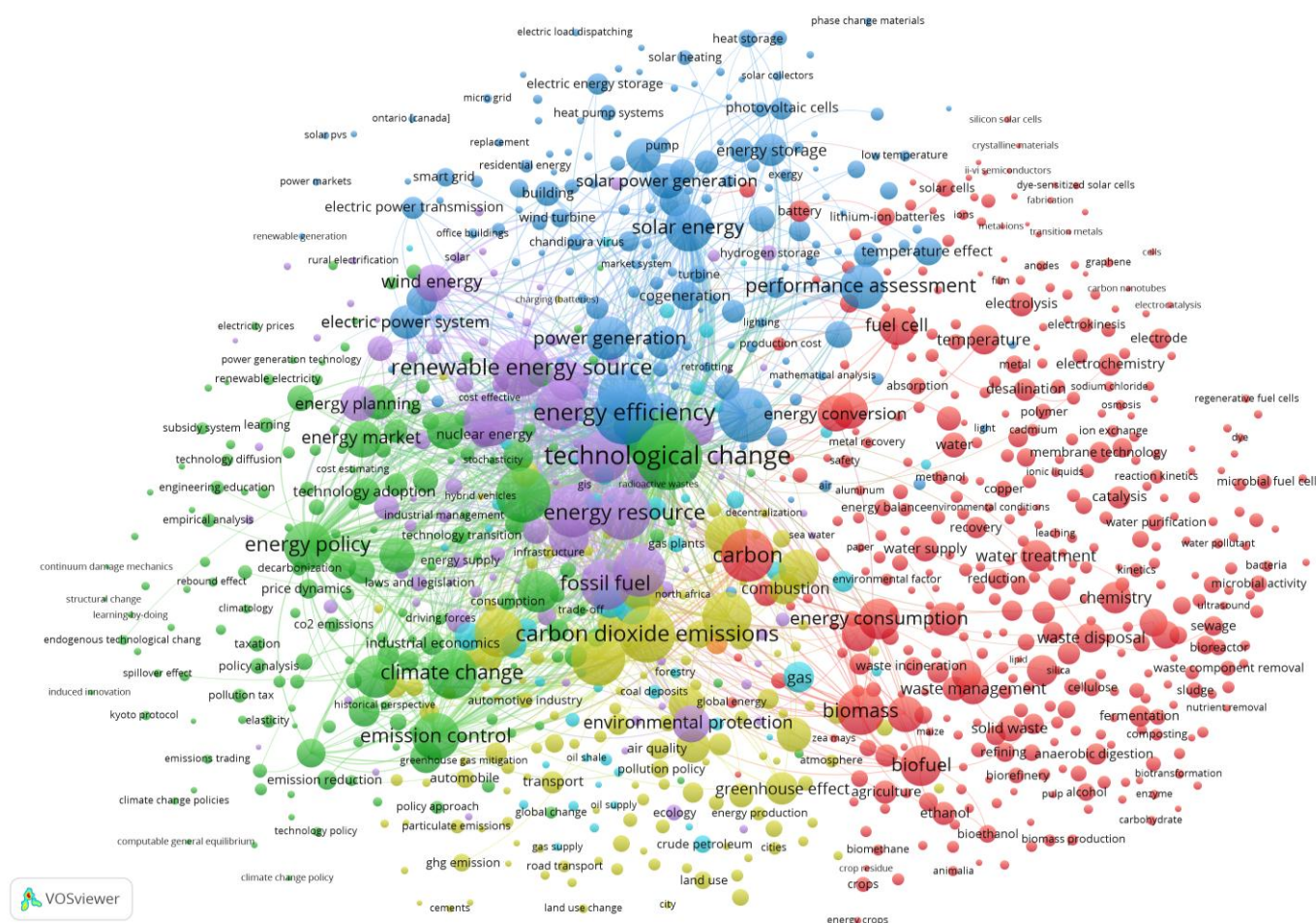


Fig. 2. Colored clusters of keywords co-occurrence network generated using VOSviewer software.

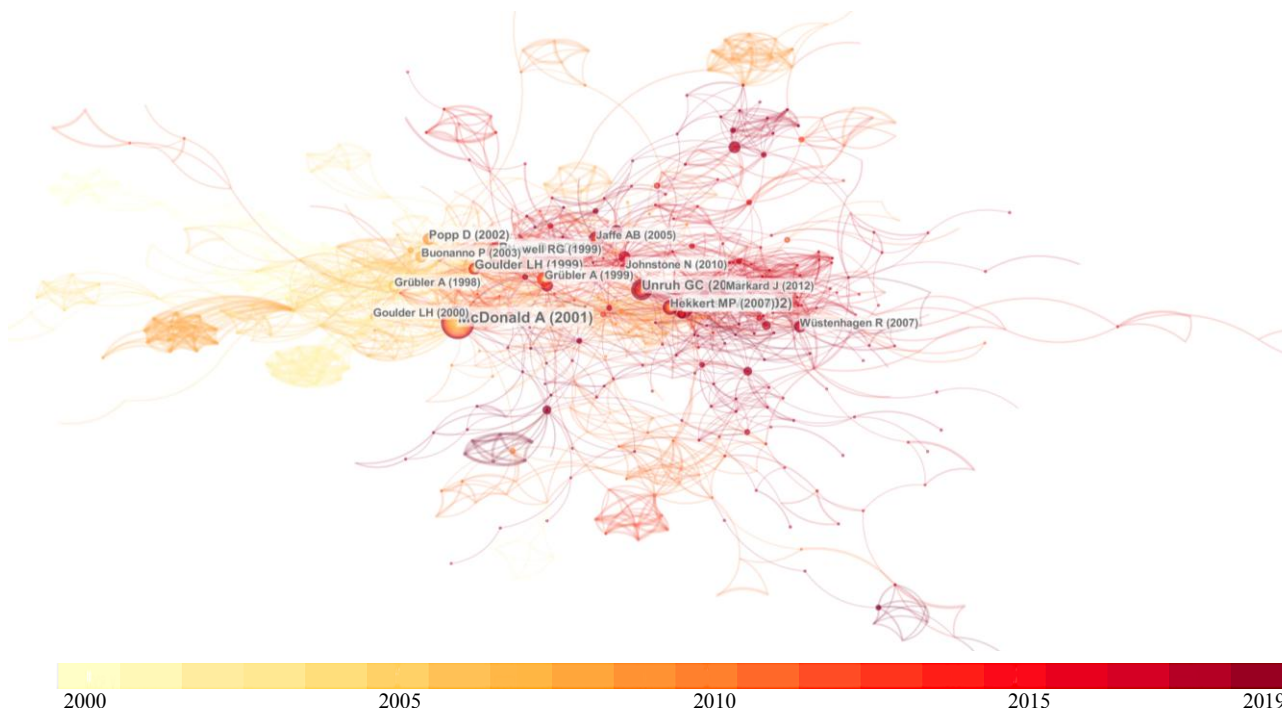


Fig. 3. Co-citation network from publications of Scopus database over period of 2000-2019 years.

References	Year	Strength	Begin	End	2000 - 2020
Messner S, 1997, Endogenised technological learning in an energy systems ..., V7, P291-313	1997	5.1843	2000	2004	-----
Grübler A, 1998, Technology and Global Change @, V0, P0	1998	7.6396	2000	2007	-----
Nakicenovic N, 1998, Global Energy Perspectives @, V0, P0	1998	3.4511	2000	2004	-----
McDonald A, 2001, Learning rates for energy technologies @ ENERGY POLICY ..., V29, P255-261	2001	8.5072	2002	2012	-----
Goulder LH, 2000, Optimal CO2 abatement in the presence of induced technological ..., V39, P1-38	2000	5.5228	2002	2008	-----
Goulder LH, 1999, Induced technological change and the attractiveness ..., V21, P211-253	1999	4.1715	2003	2009	-----
Mattsson N, 1997, Assessing new energy technologies using an energy system ..., V21, P385-393	1997	3.1934	2003	2004	-----
Newell RG, 1999, The induced innovation hypothesis and energy-saving ..., V114, P941-975	1999	3.6389	2004	2008	-----
Argote L, 1990, Learning curves in manufacturing @ SCIENCE, V247, P920-924	1990	4.4955	2004	2007	-----
Grubb M, 2002, Induced technical change in energy and environmental ..., V27, P271-308	2002	3.5716	2004	2008	-----
Popp D, 2001, The effect of new technology on energy consumption ..., V23, P215-239	2001	3.5856	2004	2006	-----
Jacobsson S, 2000, The diffusion of renewable energy technology, V28, P625-640	2000	3.9312	2005	2010	-----
Kemp R, 1998, Regime shifts to sustainability through processes of ..., V10, P175-195	1998	3.9312	2005	2010	-----
Nordhaus WD, 1994, Managing the Global Commons, V0, P0	1994	3.1554	2005	2006	-----
Gerlagh R, 2003, Gross world product and consumption in a global warming ..., V25, P35-57	2003	5.1502	2005	2008	-----
Popp D, 2002, Induced innovation and energy prices @ AMERICAN ECONOMIC ..., V92, P160-180	2002	8.3398	2005	2010	-----
Popp D, 2004, ENTICE, V48, P742-768	2004	7.6643	2005	2009	-----
Buonanno P, 2003, Endogenous induced technical change and the costs of ..., V25, P11-34	2003	8.7436	2005	2008	-----
Acemoglu D, 2002, Directed technical change @ REVIEW OF ECONOMIC STUDIES ..., V69, P781-809	2002	3.1912	2006	2010	-----
Rip A, 1998, Technological change @ HUMAN CHOICE AND CLIMATE CHANGE ..., V2, P0	1998	4.0037	2007	2010	-----
Stern N, 2007, The Economics of Climate Change, V0, P0	2007	3.4291	2008	2011	-----
Markard J, 2008, Technological innovation systems and the multi-level ..., V37, P596-615	2008	3.5256	2009	2010	-----
De Pante Henriksen A, 1997, A technology assessment primer for management of technology ..., V13, P615-638	1997	3.5939	2011	2012	-----
Färe R, 1994, Productivity growth, technical progress, and ..., V84, P66-83	1994	4.0341	2011	2014	-----
Kern F, 2008, Restructuring energy systems for sustainability?, V36, P4093-4103	2008	3.302	2012	2014	-----
Grübler A, 1999, Dynamics of energy technologies and global change @ ..., V27, P247-280	1999	3.6705	2012	2014	-----
Ferrioli F, 2009, Use and limitations of learning curves for energy technology ..., V37, P2525-2535	2009	3.2291	2012	2013	-----
Nemet GF, 2006, Beyond the learning curve, V34, P3218-3232	2006	3.4727	2012	2018	-----
Johnstone N, 2010, Renewable energy policies and technological innovation ..., V45, P133-155	2010	6.3747	2013	2017	-----
Chung YH, 1997, Productivity and undesirable outputs, V51, P229-240	1997	5.1062	2013	2016	-----
Fischer C, 2008, Environmental and technology policies for climate mitigation ..., V55, P142-162	2008	4.8062	2014	2018	-----
Wüstenhagen R, 2007, Social acceptance of renewable energy innovation, V35, P2683-2691	2007	3.7744	2014	2020	-----
Markard J, 2012, Sustainability transitions, V41, P955-967	2012	5.0054	2014	2017	-----
Acemoglu D, 2012, The environment and directed technical change @ AM ..., V102, P131-166	2012	3.2681	2014	2020	-----
Kriegler E, 2014, The role of technology for achieving climate policy ..., V123, P353-367	2014	3.4436	2015	2017	-----
Geels FW, 2011, The multi-level perspective on sustainability transitions ..., V1, P24-40	2011	3.9079	2016	2020	-----
Jacobson MZ, 2011, Providing all global energy with wind, water, ..., V39, P1154-1169	2011	3.8205	2016	2017	-----
Jacobsson S, 2006, The politics and policy of energy system transformation ..., V34, P256-276	2006	5.3655	2016	2018	-----
Peters M, 2012, The impact of technology-push and demand-pull policies ..., V41, P1296-1308	2012	3.2731	2016	2017	-----
Su B, 2012, Structural decomposition analysis applied to energy ..., V34, P177-188	2012	4.0555	2017	2020	-----
Turnheim B, 2013, The destabilisation of existing regimes, V42, P1749-1767	2013	3.1737	2017	2018	-----
Chu S, 2012, Opportunities and challenges for a sustainable energy ..., V488, P294-303	2012	3.7045	2017	2018	-----
Kivimaa P, 2016, Creative destruction or mere niche support?, V45, P205-217	2016	3.1737	2017	2018	-----
Luo X, 2015, Overview of current development in electrical energy ..., V137, P511-536	2015	4.3839	2017	2020	-----
Ang BW, 2004, Decomposition analysis for policymaking in energy, V32, P1131-1139	2004	3.7045	2017	2018	-----
Nykvist B, 2015, Rapidly falling costs of battery packs for electric ..., V5, P329-332	2015	6.081	2018	2020	-----
Rubin ES, 2015, A review of learning rates for electricity supply technologies ..., V86, P198-218	2015	5.5589	2018	2020	-----

Fig. 4. Top 47 papers with strongest citation bursts. Most intensive citation period is selected by red color.

There are several key topics identified from top-cited papers. First, the *learning rates* of energy technologies to assess forthcoming technological changes were the most important subject of interest for researchers. A highly cited research study [9] by Alan McDonald and Leo Schrattenholzer considers assembled data on cost reductions for many energy technologies to estimate learning rates. The work [10] by G.F. Nemet discusses the factors influencing cost reductions in photovoltaics. A comprehensive review of learning rates for electricity supply technologies [11] is highly cited in past 3 years. Understanding of nature of learning rates remains a key hot topic during all period 2000 – 2019.

The *directed technological change* related to is another topic widely discussed (see for example [12]). The challenge of restructuring energy systems to provide *sustainable technological transition* with large-scale involvement of *renewable energy sources* receives special attention as an important issue of energy policy [13].

Conclusions

A preliminary scientometric overview was carried out for the research domain of energy technological forecasting. General trends of technology forecasting in the energy sector were quantitatively estimated and visualized. The findings show the research spectrum from environmental policy issues like climate change and emission control to a set of alternative energy technologies including renewable solar, wind, biomass, and hydrogen technologies. However, to discover non-evident topics and relationships the deeper analysis is needed together with further comprehensive critical review similar to the methodology used in [14]. This analysis can also be improved on the base of the iterative procedure using preliminary prepared hierarchical concept maps or applied ontologies of energy technologies and forecasting methods.

Acknowledgments

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