# FACULTY OF ENGINEERING AND INFORMATION TECHNOLOGY

# Topic-based Analysis for Technology Intelligence

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## **ABSTRACT**

Since the past several decades, scientific literature, patents and other semi-structured technology indicators have been generating and accumulating at a very rapid rate. Their growth provides a wealth of information regarding technology development in both the public and private domain. However, it has also caused increasingly severe information overload problems whereby researchers, analysts and decision makers are not able to read, summarize and understand massive technical documents and records manually. The concept and tools of technology intelligence aims to handle this issue.

In the current technology intelligence research, one of the big challenges is that, the frameworks and applications of existing technology intelligence conducted semantic content analysis and temporal trend estimation separately, lacking a comprehensive perspective on trend analysis of the detailed content within an area. In addition, existing research of technology intelligence is mainly constructed on the fundamentals of semantic properties of the semi-structured technology indicators; however, single keywords and their ranking alone, are too general or ambiguous to represent complex concepts and their corresponding temporal patterns. Thirdly, systematic post-processing, forecasting and evaluation on both content analysis and trend identification outputs are still in great demand, for diverse and flexible technological decision support and opportunity discovery.

This research aims to handle these three challenges in both theoretical and practical aspects. It first quantitatively defines and presents temporal characteristics and semantic properties of typical semi-structured technology indicators. Then this thesis proposes a framework of topic-based technology intelligence, with three main functionalities, including data-driven trend identification, topic discovery and comprehensive topic

PHD Thesis, UTS Chapter 1

evaluation, to synthetically process and analyse technological publication count sequence, textual data and metadata of target technology indicators. To achieve the three functionalities, this research proposes an empirical technology trend analysis method to extract temporal trend turning points and trend segments, which help with producing a more reasonable time-based measure; a topic-based technological forecasting method to first discover and characterize the semantic knowledge underlying in massive textual data of technology indicators, meanwhile estimating the future trends of the discovered topics; a comprehensive topic evaluation method that links metadata and discovered topics, to provide integrated landscape and technological insight in depth. In order to demonstrate the proposed topic-based technology intelligence framework and all the related methods, this research presents case studies with both patents and scientific literature. Experimental results on Australian patents, United States patents and scientific papers from Web of Science database, showed that the proposed framework and methods are well-suited in dealing with semi-structured technology indicators analysis, and can provide valuable topic-based knowledge to facilitate further technological decision making or opportunity discovery with good performance.

# TABLE OF CONTENTS

CERTIFICATE OF AUTHORSHIP/ORIGINALITY	
Acknowledgements	
Abstract	ii
TABLE OF CONTENTS	······································
List of Figures	Σ
LIST OF TABLES	xii
CHAPTER 1 Introduction	1
1.1 Background	1
1.2 Research Questions and Objectives	3
1.2.1 Research Questions	3
1.2.2 Research Objectives	5
1.3 Research Significance	8
1.3.1 Theoretical Significance	8
1.3.2 Practical Significance	9
1.4 Research Methodology and Process	10
1.4.1 Research Methodology	10
1.4.2 Research Process	12
1.5 Thesis Structure	13
1.6 Publications Related to This Thesis	12
CHAPTER 2 Literature Review	17
2.1 Technology Intelligence	17
2.1.1 Concept and Framework of Technology Intelligence	17
2.1.2 Technology Intelligence in Practice	20
2.1.3 Tech Mining	21

2.2 Technology Indicators	22
2.2.1 Patent	23
2.2.2 Patent Claims	25
2.2.3 Scientific Literature	26
2.3 Empirical Technology Trend Analysis and Forecasting	27
2.3.1 Curve Fitting Based Approaches	28
2.3.2 Other Approaches	30
2.3.3 Piecewise Linear Representation	32
2.4 Topic Modelling	34
2.4.1 Semantic Space	35
2.4.2 Latent Dirichlet Allocation	36
2.4.3 Topic Modelling in Tech Mining	38
2.5 Summary	39
CHAPTER 3 Framework of Topic-based Technology Intelligence	41
3.1 Introduction	41
3.2 The Limitations of Existing Technology Intelligence Frameworks	42
3.3 Temporal Characteristics and Semantic Properties of Technology Indicate	ors44
3.4 Framework Description.	48
3.4.1 Input and Output	48
3.4.2 Topic-based Technology Intelligence	50
3.4.3 Conceptual Model of Topic-based Technology Intelligence	51
3.4.4 Overall Framework Description	54
3.5 Description of the Framework Components	57
3.5.1 Trend Identification Component	58
3.5.2 Topic Discovery Component	60
3.5.3 Comprehensive Topic Evaluation Component	62
3.6 Summary	64
CHAPTER 4 Empirical Technology Trend Analysis Method	65
4.1 Introduction	65
4.2 Data Preparation for TTA Method	67
4.2.1 Outsider Exclusion	67

4.2.2 Parameter Setup for Piecewise Linear Representation	68
4.3 TTA Method and Afterwards Trend Forecasting	71
4.3.1 Trend Turning Points Identification.	71
4.3.2 Trend Segments Identification	72
4.3.3 Trend Movement Intensity	73
4.3.4 TTA-based Trend Forecasting	74
4.4 Case Study 1: Patent Data in IP Australia	75
4.4.1 Data Sets	76
4.4.2 Outliers Exclusion	76
4.4.3 Trend States and Trend Turning Points Identification	77
4.5 Case Study 2: Patent Data in USPTO	79
4.5.1 Data Collection and Parameter Setting	79
4.5.2 Trend Forecasting for Telecommunications Technologies	81
4.5.3 Trend Forecasting for Solar Cell Technologies	84
4.5.4 Trend Forecasting for Radar-related Technologies in USPC 342	87
4.5.5 Comparison and Discussion	90
4.6 Summary	95
CHAPTER 5 Topic-based Technological Forecasting Method and afterwards Cor	itent
Analysis	97
5.1 Introduction	97
5.2 Patent Crawling and Cleaning	98
5.3 Topic-based Technological Forecasting Method	101
5.3.1 Method Framework	101
5.3.2 Topic Modelling	104
5.3.3 Topic-based Technological Forecasting and Analysis	105
5.4 Case Study of Topic-based Technological Forecasting by Patent Data	108
5.4.1 Trend Pattern Identification	109
5.4.2 Topic Modelling and Prominent Topic Selection	110
5.4.3 Topic Annual Weight Matrix and Topic-based Trend Coefficients Estin	nation
	112
5.4.4 Topic-based Trend Forecasting and Analysis	113

5.4.5 Discussion	116
5.5 Technological Topic Change Identification Approach	117
5.5.1 Framework of the TTCI Approach	117
5.5.2 Topic Modelling	119
5.5.3 Topic Change Identification Model	120
5.5.4 Case Study of TTCI Approach by Patent Data	122
5.6 Fuzzy Number-based Technological Trend Measurement Approach	128
5.6.1 Framework of the FTTM Approach	128
5.6.2 Fuzzy Set	130
5.6.3 Topic Weight Estimation	131
5.6.4 Fuzzy-based Technological Development Measurement	131
5.6.5 Case Study of FTTM Approach by Patent Data	134
5.7 Summary	138
CHAPTER 6 Topic Detection and Comprehensive Evaluation Method	139
6.1 Introduction	139
6.2 Methodology Framework	141
6.3 Topic Modelling	143
6.3.1 Parameters Setting	143
6.3.2 Final Topic Set Determination	145
6.4 Topic Evaluation Indices	146
6.4.1 Topic Weight Index	147
6.4.2 Topic Trend Index	147
6.4.3 Topic Activeness Index	149
6.4.4 Topic-based Citation	149
6.4.5 Topic-based Prominent Topics and Documents Identification Using M	[etadata
	151
6.5 Case study: Dye-sensitized Solar Cells Scientific Literature	151
6.5.1 Data	151
6.5.2 Scientific Literature Text Cleaning.	152
6.5.3 Parameters Setting and Final Topic Set Determination	153
6.5.4 Topic Evaluation Result	154

6.5.5 Topic-based Evaluation Maps	156
6.5.6 Prominent Topics and Papers Analysis	160
6.5.7 Discussion	165
6.6 Summary	166
CHAPTER 7 Conclusions and Further Study	167
7.1 Conclusions	167
7.2 Further Study	171
References	173
Abbreviations	186
Appendix	187

# LIST OF FIGURES

Figure 1-1. The methodology of design research of this thesis	11
Figure 1-2. Thesis structure	13
Figure 2-1. Technology intelligence service define by Savioz (2004)	19
Figure 2-2. The technology intelligence process defined by Kerr at al. (2006)	20
Figure 2-3. An example of curve fitting approaches in technology forecasting	29
Figure 2-4. Comparison of growth curves, PLR-based approaches and time series	
analysis	33
Figure 2-5. Brief introduction of the 'semantic space' of topic modelling	35
Figure 2-6. The graphical model of Latent Dirichlet Allocation	37
Figure 3-1. The three-dimensional schematic structure of the semantic property and	
temporal characteristic	46
Figure 3-2. 'Topics' and 'Trend Segments'	48
Figure 3-3. Schematic diagram of the input and output of the topic-based technology	
intelligence	50
Figure 3-4. Brief introduction of topic-based technology intelligence	51
Figure 3-5. The Conceptual model of topic-based technology intelligence	53
Figure 3-6. Framework of topic-based technology intelligence	56
Figure 3-7. Details of the trend identification component	59
Figure 3-8. Details of the topic discovery component	61
Figure 3-9. Details of the comprehensive topic evaluation component	63
Figure 4-1. An example of PLR threshold setup	70
Figure 4-2. An example of transforming original data to trend segments step by step	73
Figure 4-3. The detailed process of the TTA-based trend forecasting	75
Figure 4-4. The outliers exclusion for ICT patent count sequence	77

Figure 4-5. Original data, PLR segmentation and trend segments of technologies in IC	CT
industry	78
Figure 4-6. Normalized original and cumulative patent data in the case study	80
Figure 4-7. PLR threshold setup for the three technologies in the case study	81
Figure 4-8. Original data, PLR segmentation result and trend segments of	
telecommunication technologies	82
Figure 4-9. The forecasting result of the trend segments of Telecommunication	
technologies	84
Figure 4-10. Original data, PLR segmentation result and trend segments of solar cell	
technologies solar cell technologies	85
Figure 4-11. The forecasting result of the trend segments of solar cell technologies	87
Figure 4-12. Original data, PLR segmentation result and trend segments of radar-rela	ted
technologies	88
Figure 4-13. The forecasting result of the trend segments of radar-related technologie	s.90
Figure 4-14. The growth curves fitting result for the three technologies	92
Figure 4-15. Experimental comparison between the proposed approach and the growt	th
curves model	93
Figure 5-1. An example of webpage crawling result	100
Figure 5-2. Framework for the topic-based technological forecasting method	103
Figure 5-3. An example of topic distribution matrix in chronological order	107
Figure 5-4. The trend turning points and trend segments generated from patenting	
activities	109
Figure 5-5. The curve fitting result for topic trend estimation of the 10 selected topics	s 114
Figure 5-6. The framework of the proposed technological topic change identification	
approach	118
Figure 5-7. Relationships between sub-collections and topics	119
Figure 5-8. Topic change identification model	120
Figure 5-9. Topics became newly important in each year of 2010 to 2013 and topmos	st
frequent words of each topic	125
Figure 5-10. An example of the topic "antibody" evolving over time	126
Figure 5-11. An example of the topic-based trend estimation of the theme "antibody"	.127

Figure 5-12. The framework of FTTM approach	29
Figure 5-13. Linguistic terms and their membership functions	33
Figure 5-14. Membership functions of linguistic terms in term set I	36
Figure 6-1. The framework of the topic detection and comprehensive evaluation method	Ĺ
14	12
Figure 6-2. An example of annual weight matrix	18
Figure 6-3. The log likelihood of the probability of the observation under models with	
different setting of the number of topics	54
Figure 6-4. The annual weight growth of the top 10 high weighted topic	55
Figure 6-5. The topic-based evaluation map based on weight, trend and activeness	
characteristics of DSSCs corpus	57
Figure 6-6. The topic-based citations map based on the total citation	59
Figure 6-7. The topic evaluation map based on weight, trend, activeness and topic-based	l
citations of DSSCs corpus	50
Figure 6-8. Graphical illustration of the main content of the 6 prominent topics16	51
Figure 6-9. The cumulative citation distribution of DSSCs corpus from year 1991 to 201	4
	52
Figure 6-10. Publications country distribution of DSSCs corpus from year 1991 to 2014	
	54

# LIST OF TABLES

Γable 4-1. The detailed outliers' values of ICT patent count sequence	77
Γable 4-2. Trend signal value and trend tags of ICT technologies in Australia	79
Γable 4-3. Description of the data sets for case study 2	80
Γable 4-4. Curve fitting information for PLR threshold determination	81
Table 4-5. The trend segments information of Telecommunication technologies	83
Table 4-6. The trend segments information of solar cell technologies	86
Table 4-7. The trend segments information of radar-related technologies	89
Table 4-8. Growth curves selection for comparison experiments	91
Table 4-9. Forecasting result comparison	94
Table 5-1. The start and end of webpage source code while patent crawling	99
Table 5-2. The basic notations throughout this chapter	105
Table 5-3. Trend forecasting indicators and future trend estimation	107
Table 5-4. The trend turning points, document numbers, and term numbers for each	trend
segment	110
Table 5-5. The 50 topics generated from the patent claims collection and their weig	ht
indicator	111
Γable 5-6. The annual weight matrix of the selected top 10 significant topics	112
Γable 5-7. Topic-based trend coefficients for all 10 prominent topics	113
Γable 5-8. The details of trend estimation for the 10 selected topics	115
Table 5-9. The number of documents, terms and USPC of patents published each years.	ear 122
Γable 5-10. Linguistic terms and fuzzy numbers	133
Table 5-11. The Temporal-weight coefficients of topic 1 to topic 30	135
Γable 5-12. Development states measure result	136
Table 6-1. The number of documents, terms and subjects of documents each year	152

Table 6-2. Similarity evaluation result for the final topic set selection	154
Table 6-3. The top 10 topics with the highest evaluation indices values	156
Table 6-4. Top 10 topics with the highest values (normalized) of the three indices	158
Table 6-5. Detected topics and the publications they covered	163
Table 6-6. The most contributory papers of the prominent topic set	165