

Original research article

Evolution and Emerging Trends in HFT Research

ABSTRACT

Aims: In this paper, we try to study the evolution and emerging trends of High Frequency Trading (HFT) research by examining papers published in the Web of Science (WOS) between 1993 and 2017.

Study design: A total of 241 papers were included, and 1876 keywords from these articles were extracted and analyzed.

Place and Duration of Study: For tracing the dynamic changes of the HFT Research, the whole 24 year was further separated three consecutive periods: 1993-2002, 2003-2012, and 2013-2017.

Methodology: The Ucinet is adopted to get keywords network, or knowledge network, to study the relationship of each research theme. NetDraw was applied to visualize network. We used social network analysis (SNA) technique to reveal patterns and trends in the research by measuring the association strength of terms representative of relevant publications produced in HFT field.

Results: Results indicate that HFT research has been strongly influenced by “market”, “prices”, “finance”, “liquidity”, “statistics”, “financial markets”, “stock”, “stochastic”, “model” and “trades” as shown in Table 1, which represent some established research themes. They are major focuses and the bridges connecting to other research themes in HFT. The detailed analysis in results and discussion provides an overview of evolution and emerging trends in HFT Research.

Conclusion: It concludes that market performance related keywords, which represent some established research themes, have become the major focus in HFT research. It also changes rapidly to embrace new themes. Especially, this research may make contribution to enlarge research method in that there is no SNA research in HFT research before.

Keywords: High Frequency Trading, HFT, social network analysis, SNA, emerging trends

1. INTRODUCTION

As the stock market has become nearly exclusively electronic, advances in computer technology and automated algorithm trading have speeding the transmission and execution of security transaction orders, and thus establishing High Frequency Trading (HFT) [1]. HFT is an emerging, ever changing and rapidly evolving area with highly interdisciplinary in nature for the markets, regulators, and the public [2]. This diversity may root from the emerging nature of computing technology and its wide appeal as well as unique researcher and practitioner viewpoints. Many academics raised the controversy concerning about HFT [3]. Even SEC Division of Trading and Markets Director Brett Redfearn admitted, “There are a lot of different definitions of HFT.” The diverse issues and findings in the field of HFT represent the introduction of ideas and even new concepts about HFT. What are the areas

29 of focus in HFT? What are the developing trends in current research? Keywords have been
30 generally identified as the words that reflect the research themes of individual publications
31 that concern researchers. Further, keywords network represents relationships of keywords
32 among HFT papers. When two keywords occur in a same article, it is an indication of
33 connection between the themes which they represent. Therefore, a comprehensive network
34 perspective analysis is required to reveal the developmental trends or future orientation of
35 possible new research field from HFT.

36 Social network analysis (SNA), sometimes also referred to as “structural analysis” [4], is a
37 broad strategy for investigating social structures. For measurement, social network analysis
38 (SNA) measures are a vital tool for understanding the behavior of networks and graphs.
39 These algorithms use graph theory to calculate the importance of any given node in a
40 network [5]. When they’re well implemented, SNA measures allow the analyst to cut through
41 noisy data and hone into the parts of a network that require further attention.

42 In this paper, our focus is to construct and analyze keywords network by using the Social
43 network analysis (SNA) techniques which have already been widely applied in many
44 disciplines of science. Specifically, this study will quantitatively analyze existing empirical
45 and theoretical HFT papers to address the following objectives:

46 1) To construct keywords network from HFT papers published in world leading journals
47 during the period from 1993 to 2017.

48 2) To investigate the characteristics of keywords network of HFT papers by utilizing Social
49 Network Analysis (SNA) techniques.

50 3) To find and compare the changes in keywords network of HFT papers over time.

51 These investigations can help researchers to realize the breadth of HFT research and to
52 establish future research directions and to provide an entry point to any academic,
53 regardless of their prior knowledge of the theme.

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55 2. METHODOLOGY

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58 2.1 Publication search

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60 The objective of the present work is to identify the important keywords from the scientific
61 output on the latest advances in HFT, and to describe the characteristics of the network of
62 keywords of HFT research. To achieve these goals, we selected the Web of Science (WOS),
63 which includes SCIE and SSCI and A&HCI from the Institute of Scientific Information (ISI)
64 Web of Science databases. WOS is the most important and frequently used source for a
65 broad review of scientific accomplishment in all research fields [6]. We constructed a
66 database composed of keywords from HFT papers published in the WOS during the 24-year
67 period from 1993 to 2017. The keywords were obtained from following two sources: (1)
68 Author Keywords and (2) Keywords Plus in the ISI database [7].
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70 2.2 Refinement of keywords and keywords databases

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72 Due to different words may represent same or similar ideas and concepts, we standardize
73 the keywords before constructing the keywords network. The basic rule for the refinement of
74 keywords was that all keywords with identical meaning or similar ideas or concepts or even

75 misspelled keywords from different articles will be grouped and considered as a single
 76 keyword. This refinement leads to a meaningful keywords database. The example of SNA
 77 steps in literature-based research was shown in Fig.1.
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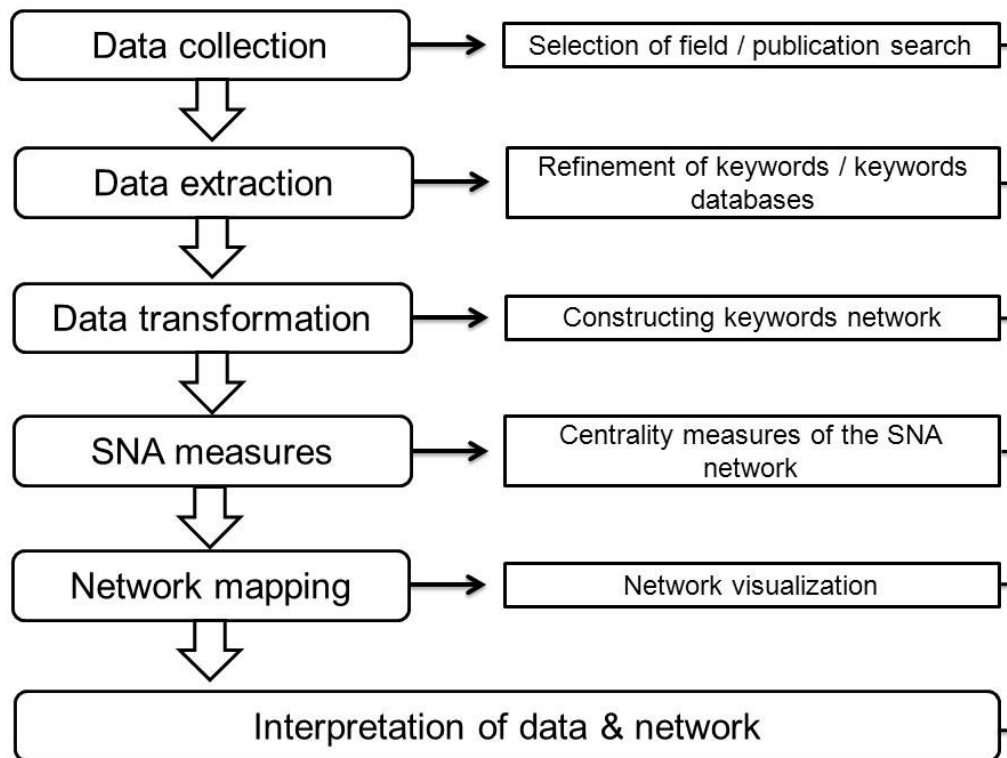


Fig. 1 example of SNA steps in literature-based research

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2.3 Constructing keywords network

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The construction of keywords network is based on three continuous stages which include data collection, data extraction, and data transformation. During the data extraction stage, core keywords are identified from HFT papers and are changed to a standard form. Then in the data transformation stage, all the refined keywords will be input to the most popular social network research tool, Ucinet 6 for Windows [8] to get keywords network, or knowledge network, to study the relationship of each research theme.

2.4 Centrality measures of the SNA network

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Network centrality [9] in the keywords network can measure the degree of relations among keywords. Social network analysis (SNA) measures include measuring degree centrality, betweenness centrality, closeness centrality, EigenCentrality or PageRank for each network quantitatively [10]. In order to understand the characteristics of the overall keywords network in HFT research, we selectively used betweenness centrality measuring to study the relationship of each research theme. Betweenness centrality is the extent to which a node lies on the paths between other nodes. It is measured as the fraction of the shortest paths between all pairs of other nodes in the network containing the node. A keyword that lies between two distinctive research themes can have high betweenness centrality even though

102 it may have a small number of connections to other keywords in each theme [11]. In the
103 keywords network, this represents the importance of a keyword in bridging subsets of
104 keywords.

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106 **2.5 Network visualizations**

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108 Network visualizations is generally known as network mapping which can be generated from
109 raw network data within Netdraw, a mapping program in Ucinet. NetDraw was applied to
110 visualize network. It helps to obtain a clear sense of connectivity of keyword networks and to
111 illustrate the overall patterns of networks over time. This method enables the researchers to
112 explicitly understand representation of emerging themes.

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115 **3. RESULTS AND DISCUSSION**

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118 **3.1 Keywords network**

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120 Fig. 2 shows keywords network by co-occurrence (1993-2017). The nodes are the keywords.
121 The size of nodes can reflect the frequency of keywords. Larger size of node means higher
122 frequency of occurrence of keyword. The lines between two nodes stand for the associations
123 of two keywords, or represent the co-occurrence of these keywords in a paper. The
124 thickness of line indicates the co-occurrence frequency of keyword pairs, or represents the
125 number of times each pair of keywords was mentioned together in papers. The thickness of
126 line is proportional to the closeness of connections between two keywords. The thicker line
127 between two keywords, the closer their relationship is. The more co-occurrence between two
128 keywords, the closer their relationship is. It shows the strength of the connection. According
129 to Fig.2, we can see that keywords such as “market”, “prices”, “finance”, “liquidity”,
130 “statistics”, “financial markets”, “stock”, “stochastic”, “model” and “trades” became important
131 keywords, which means that they have played an important role in bridging other research
132 themes.

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134 **3.2 Betweenness centrality measuring for all period (1993-2017)**

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136 Keywords serve as an indicator of the importance of the research themes they represent.
137 The top ten keywords from betweenness centrality measuring for all period (1993-2017) are
138 “market”, “prices”, “finance”, “liquidity”, “statistics”, “financial markets”, “stock”, “stochastic”,
139 “model” and “trades” as shown in Table 1. The results indicate that these research themes
140 are major focuses and the bridges connecting to other research themes in HFT. These
141 findings show that these research themes attract more attention and have a closer
142 relationship with other research themes in HFT. Notice that keywords like “High Frequency
143 Trading” and “Algorithm(s)” have very broad meanings. Due to this kind of keywords are
144 meaningless for this study, we excluded them from the above analysis.

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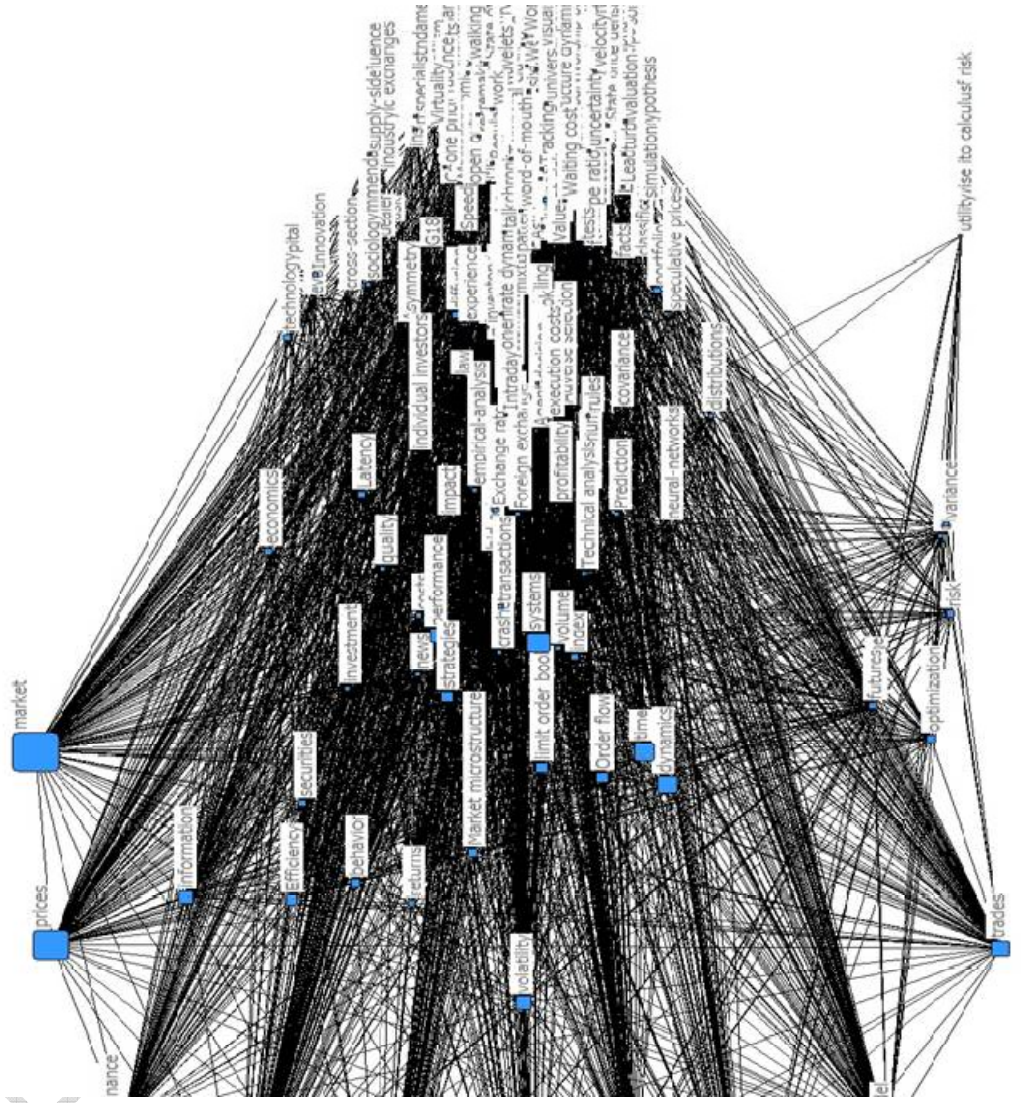
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Prices network by co-occurrence (1993-2017)

Table 1: Betweenness centrality measuring for all period (1993-2017)

1993 - 2017			1993 - 2017			
No.	rank	Keywords	No.	rank	Keywords	
196	1	High-frequency trading	20972.666	99	costs	479.408
13	2	Algorithms	14873.707	142	empirical-analysis	428.283
258	3	market	11343.833	339	quality	425.657
333	4	prices	9020.214	435	volume	425.474
169	5	Finance	8998.353	430	variance	420.01
250	6	Liquidity	8372.251	378	sociology	403.265
389	7	Statistics	6550.62	329	power	381.97
170	8	Financial markets	6365.009	410	Technical analysis	378.11
391	9	stock	5479.472	175	Foreign exchange	343.948
390	10	Stochastic	5255.224	425	universal portfolios	339.003
278	11	model	5207.57	206	impact	305.744
419	12	trades	4817.368	158	Exchange rate	291.833
408	13	systems	4643.722	332	Prediction	284.327
132	14	dynamics	4483.216	301	options	283.819
416	15	time	3957.8	149	equilibrium	253.83
434	16	volatility	3471.744	78	competition	241.319
255	17	management	3040.512	10	Agent-based modelling	209.288
215	18	information	2945.935	216	Innovation	207.791
302	19	Order flow	2761.069	291	neural-networks	194.051
392	20	strategies	2683.609	213	individual investors	186.9
319	21	performance	2651.219	100	covariance	172.821
139	22	Efficiency	2355.799	91	Content-based	161.023
37	23	behavior	1895.172	39	bid-ask spread	158.613
259	24	Market microstructure	1890.566	115	decision	150.691
300	25	optimization	1622.698	371	70 sharpe ratio	140.984
352	26	returns	1461.738	155	71 evolution	129.907
210	27	index	1404.703	337	72 profitability	124.693
249	28	limit order book	1389.61	364	73 selection	123.284
54	29	capital	1313.427	256	74 Manipulation	117.729
326	30	portfolio	1312.679	422	75 turbulence	115.056
19	31	arbitrage	1276.097	159	76 execution costs	98.269
240	32	Latency	1247.931	242	77 law	94.131
184	33	futures	1177.844	360	78 rules	85.183
363	34	securities	1130.502	244	79 Lead-lag relationship	82.599
411	35	technology	1065.972	157	80 exchange	76.936
354	36	risk	1020.679	9	81 Adverse selection	70.177
226	37	investment	974.707	18	82 Approximation	68.299
138	38	economics	806.251	160	83 experience	63.962
293	39	news	750.328	22	84 ask	51.969
420	40	transactions	713.42	27	85 Asymmetry	40.406
30	41	Automation	681.092	70	86 Codings	40.025
122	42	diffusion	652.382	112	87 dealer	39.832
128	43	distributions	584.501	62	88 classification	39.252
101	44	crashes	532.398	382	89 speculative prices	38.453
297	45	Online learning	501.978	223	90 Intraday	38.163

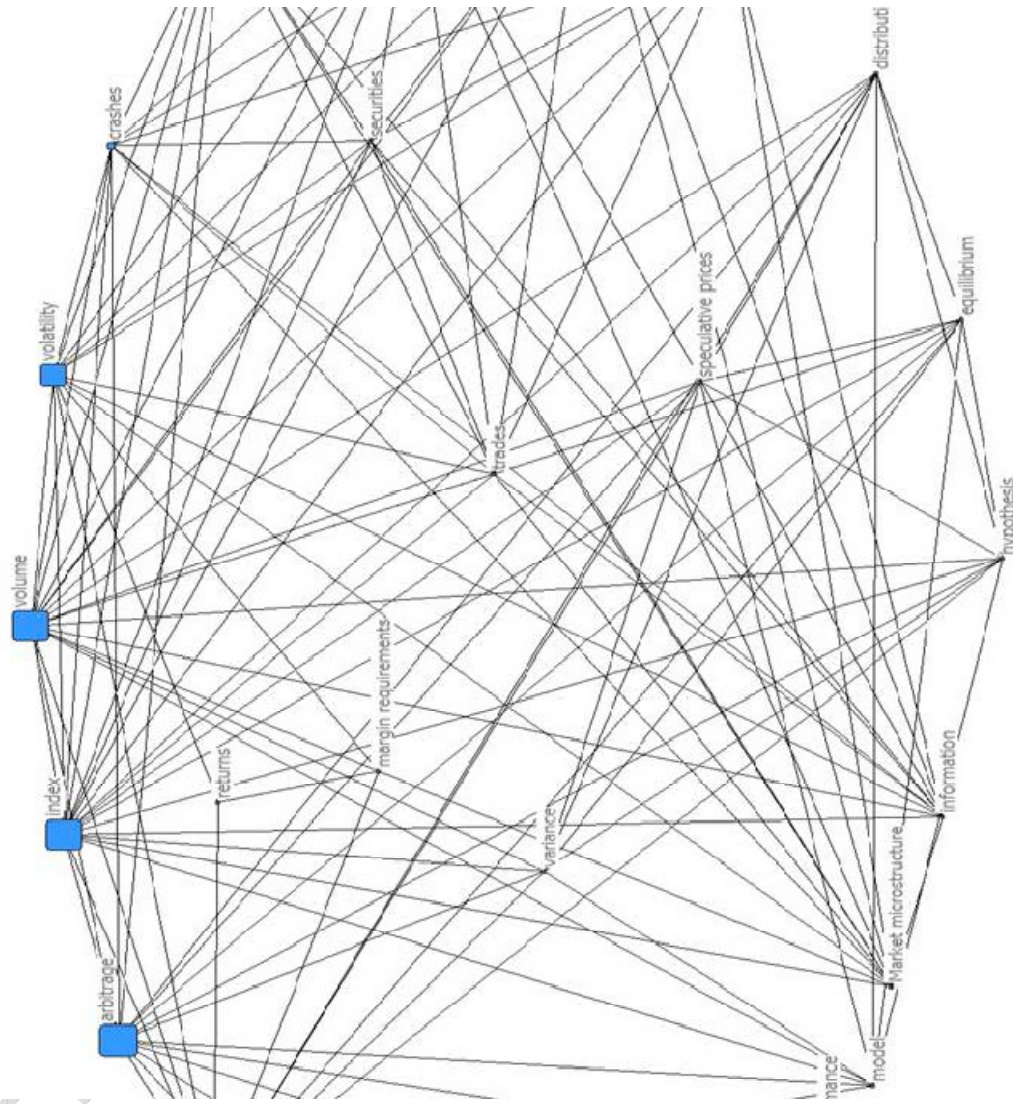
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3.3 Changes in important keywords over time

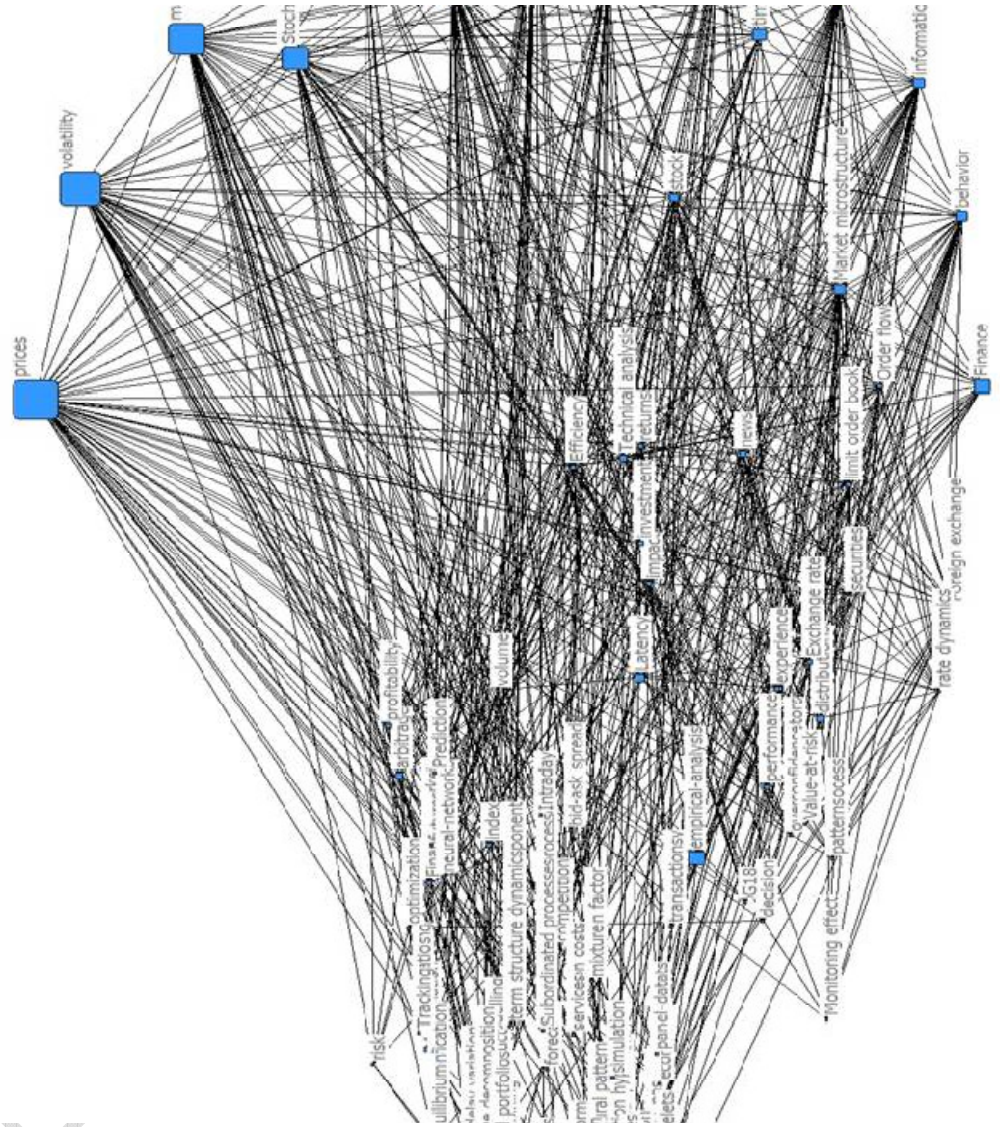
How have the important keywords changed over time and what are the recent important keywords? In order to trace dynamic changes of the HFT Research, the whole 24 year was further separated three consecutive periods: 1993-2002, 2003-2012, and 2013-2017. We constructed three keywords networks as shown in Fig. 3 to 5. For showing statistics of keywords network in different time slices, we compared the rank of the important keywords in the three keywords networks constructed as shown in Fig. 6 in order to thoroughly and precisely analyze the variations of trends. Please notice that the important keywords are from top ten keywords in Table 1. For the full lists of keywords in these three periods, see Appendix A through Appendix C.

This comparison reveals some notable results. "Market" revealed to be the most important keyword by betweenness centrality measuring for all three periods, because it has received consistent upward attention. "Stock" even received sharply upward attention since 1993 until 2017. "Liquidity" and "finance" and "financial markets" are emerging theme since 2003 year due to they were not appeared in period of 1993 to 2002. "Model" and "trades" were paid growing attention from 1993 through 2012 period, while 2013 to 2017 were not. The reason may be that "model" and "trades" are viewed as common sense already in HFT research until recent years. "Prices" and "stochastic" emerged since 2003 year, but they were paid less attention from 2013 through 2017 period. The above analysis provides an overview of HFT research and it concludes that market performance related keywords, which represent some established research themes, have become the major focus in HFT research. It also changes rapidly to embrace new themes.

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ds network by co-occurrence (1993-2002)



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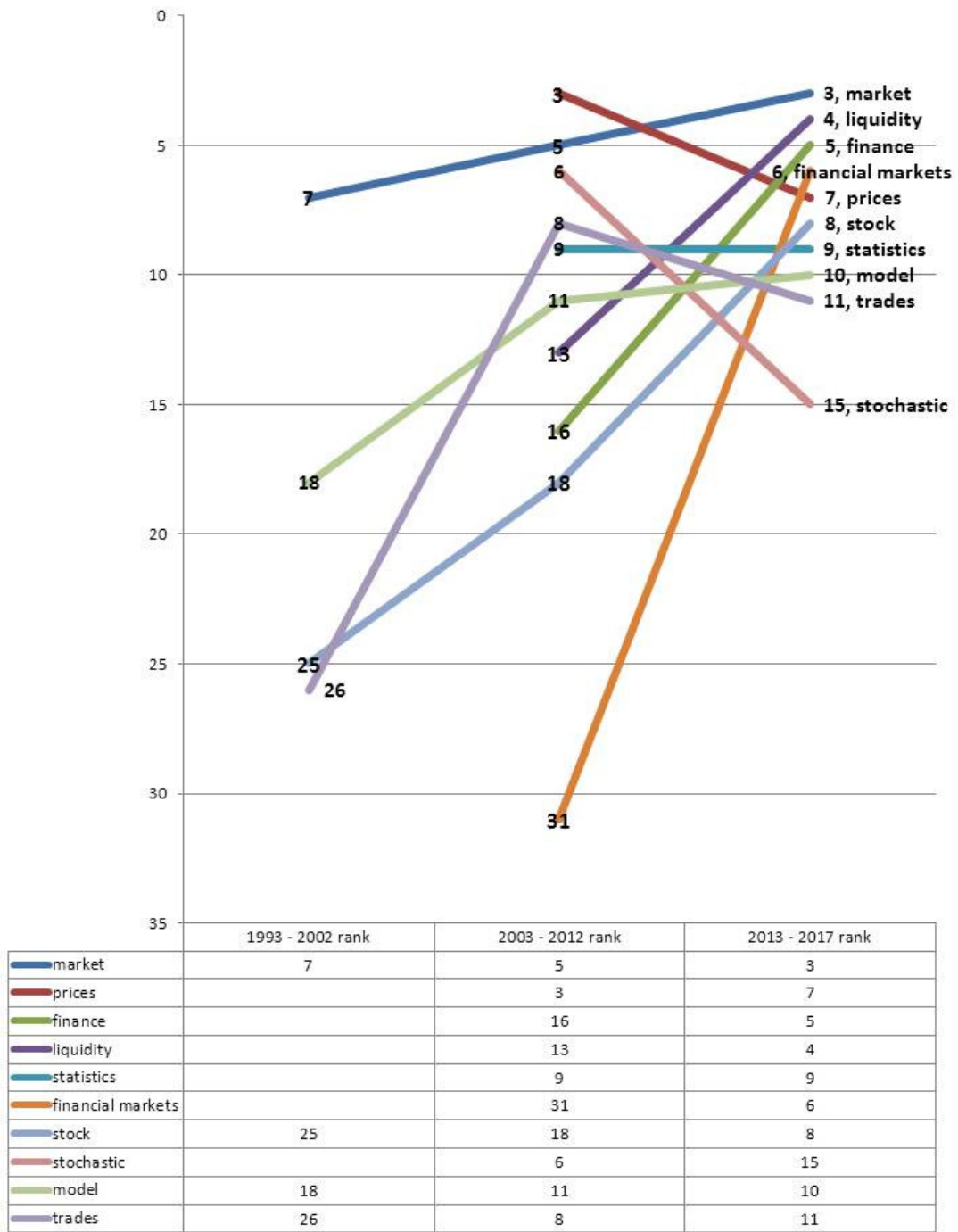


Fig. 6 Changes in important keywords over time

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4. CONCLUSION

In this article, we used social network analysis (SNA) technique to give a comprehensive understanding of HFT research during 1993 to 2017. We obtain some clear and reasonable results which can provide useful insights to better understand evolution and emerging trends in HFT research.

Results indicate that HFT research has been strongly influenced by “market”, “prices”, “finance”, “liquidity”, “statistics”, “financial markets”, “stock”, “stochastic”, “model” and “trades”, which represent some established research themes. They are major focuses and the bridges connecting to other research themes in HFT. “Market” revealed to be the most important keyword by betweenness centrality measuring for all three periods, because it has received consistent upward attention. “Stock” even received sharply upward attention since 1993 until 2017. “Liquidity” and “finance” and “financial markets” are emerging theme since 2003 year due to they were not appeared in period of 1993 to 2002. “Model” and “trades” were paid growing attention from 1993 through 2012 period, while 2013 to 2017 were not. The reason may be that “model” and “trades” are viewed as common sense already in HFT research until recent years. “Prices” and “stochastic” emerged since 2003 year, but they were paid less attention from 2013 through 2017 period. The above analysis provides an overview of HFT research and it concludes that market performance related keywords, which represent some established research themes, have become the major focus in HFT research. It also changes rapidly to embrace new themes.

This research is just a preliminary and still has limitations need to be addressed. The main limitation of SNA technique is that it is just one of the tools that can be used to understand evolution and emerging trends in HFT research. It is just one piece of the puzzle. Subject matter experts are needed to provide a context for the research. On the other hand, this study tries to explore the evolution and emerging trends in HFT papers published in world leading journals but the Web of Science database may not completely cover the scientific research of HFT.

In the future, comparative research with other method in the same HFT field could also be explored because different methods may have very different research emphases which would also be worthy of further exploration to extend HFT research theme.

This study utilizes the advantage of SNA technique and such keywords analysis might be helpful to stimulate further research or identify some fruitful future research opportunities.

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Appendix A
Betweenness centrality measuring for the first sub-period (1993-2002)

		1993 - 2002	
No.	rank	Keywords	
8	1	futures	91.474
1	2	arbitrage	69.006
11	3	index	68.29
27	4	volume	68.29
26	5	volatility	46.118
5	6	crashes	9.371
14	7	market	9.371
2	8	bid-ask spread	0
3	9	components	0
4	10	costs	0
6	11	distributions	0
7	12	equilibrium	0
9	13	High-frequency trading	0
10	14	hypothesis	0
12	15	information	0
13	16	margin requirements	0
15	17	Market microstructure	0
16	18	model	0
17	19	performance	0
18	20	profitability	0
19	21	returns	0
20	22	risk	0
21	23	securities	0
22	24	speculative prices	0
23	25	stock	0
24	26	trades	0
25	27	variance	0

Appendix B
Betweenness centrality measuring for the second sub-period (2003-2012)

2003 - 2012				2003 - 2012			
No.	rank	Keywords		No.	rank	Keywords	
5	1	Algorithms	2723.209	119	46	risk	21.541
68	2	High-frequency trading	1947.709	44	47	economics	21.067
113	3	prices	1943.382	62	48	futures	19.741
153	4	volatility	1592.97	13	49	bid-ask spread	18.193
86	5	market	1466.067	149	50	Value-at-risk	15.594
129	6	Stochastic	1124.416	75	51	Intraday	12.218
136	7	systems	1108.412	15	52	Boosting	2.133
144	8	trades	844.66	1	53	1st passage	0
128	9	Statistics	844.355	2	54	Active measurement	0
131	10	strategies	776.203	3	55	Adaptive trader-agents	0
92	11	model	714.488	4	56	Agent-based modelling	0
141	12	time	550.082	6	57	amorphous solids	0
84	13	Liquidity	393.819	7	58	anomalous diffusion	0
73	14	information	387.353	8	59	Approximation	0
12	15	behavior	379.06	10	60	Asynchronous data	0
55	16	Finance	330.972	14	61	Binary classification	0
87	17	Market microstructure	305.516	16	62	C33	0
130	18	stock	276.927	17	63	C41	0
41	19	distributions	198.327	18	64	C50	0
137	20	Technical analysis	190.978	19	65	cascades	0
95	21	news	186.846	20	66	choice	0
80	22	Latency	164.73	21	67	classification	0
9	23	arbitrage	163.332	22	68	Cloud computing	0
45	24	Efficiency	135.079	23	69	Codes of conduct	0
46	25	empirical-analysis	134.262	24	70	Codings	0
100	26	Order flow	130.235	25	71	Commodity hardware	0
58	27	Foreign exchange	128.547	26	72	Common factor	0
76	28	investment	128.284	27	73	Commonality	0
118	29	returns	114.102	28	74	competition	0
83	30	limit order book	109.02	29	75	component analysis	0
56	31	Financial markets	97.433	30	76	components	0
107	32	performance	87.288	31	77	continuous double auction	0
71	33	index	56.19	32	78	costs	0
115	34	profitability	56.01	33	79	covariance	0
36	35	decision	55.881	34	80	crashes	0
52	36	experience	54.663	35	81	Data stream processing	0
50	37	Exchange rate	52.486	37	82	Detrending	0
70	38	impact	50.131	38	83	diffusion	0
98	39	optimization	45.131	39	84	disposition	0
121	40	securities	34.477	40	85	Distributed processing	0
116	41	rate dynamics	31.575	42	86	dynamics	0
154	42	volume	30.481	43	87	EaaS	0
11	43	Automation	26.278	47	88	equilibrium	0
112	44	Prediction	26.052	48	89	error-correction	0
94	45	neural-networks	22.656	49	90	evolution	0

Appendix C
Betweenness centrality measuring for the third sub-period (2013-2017)

2013 - 2017				2013 - 2017			
No.	rank	Keywords		No.	rank	Keywords	
163	1	High-frequency trading	15349.117	255	46	options	322.945
10	2	Algorithms	9033.231	277	47	power	320.466
217	3	market	8454.144	248	48	news	298.916
211	4	Liquidity	7070.412	365	49	volume	244.794
143	5	Finance	6733.11	362	50	variance	210.176
144	6	Financial markets	5455.364	246	51	neural-networks	206.152
280	7	prices	5301.453	7	52	Agent-based modelling	185.942
333	8	stock	5137.49	279	53	Prediction	181.464
331	9	Statistics	4706.375	62	54	competition	168.874
234	10	model	3940.322	171	55	impact	158.898
353	11	trades	3479.517	74	56	Content-based	144.998
111	12	dynamics	3275.315	309	57	selection	135.227
346	13	systems	2731.716	203	58	law	107.017
275	14	portfolio	2696.57	216	59	Manipulation	104.194
332	15	Stochastic	2543.021	120	60	empirical-analysis	97.351
215	16	management	2533.632	305	61	rules	95.875
256	17	Order flow	2276.406	83	62	covariance	85.314
352	18	time	2174.244	181	63	Innovation	85.282
180	19	information	2104.56	127	64	equilibrium	79.572
269	20	performance	1684.337	178	65	individual investors	77.577
254	21	optimization	1560.585	134	66	exchange	75.6
117	22	Efficiency	1499.502	33	67	bid-ask spread	74.359
44	23	capital	1204.612	136	68	execution costs	68.596
297	24	returns	1129.319	6	69	Adverse selection	64.104
31	25	behavior	1043.882	315	70	sharpe ratio	58.984
210	26	limit order book	981.42	348	71	Technical analysis	53.715
349	27	technology	929.715	17	72	ask	52.004
218	28	Market microstructure	885.26	135	73	Exchange rate	48.534
175	29	index	855.817	22	74	Asymmetry	45.24
364	30	volatility	833.319	139	75	facts	42.178
14	31	arbitrage	755.86	94	76	dealer	35.465
299	32	risk	752.568	38	77	book	33.13
308	33	securities	745.66	97	78	decision	31.873
191	34	investment	711.843	132	79	evolution	31.071
155	35	futures	635.551	192	80	issues	27.821
354	36	transactions	618.743	327	81	spread	23.048
334	37	strategies	590.264	106	82	discovery	22.219
202	38	Latency	579.702	133	83	Evolutionary computation	20.034
24	39	Automation	509.524	236	84	Momentum	18.852
103	40	diffusion	495.759	50	85	classification	17.174
116	41	economics	456.811	245	86	networks	16.995
285	42	quality	415.741	190	87	Inventory risk	16.062
321	43	sociology	409.809	8	88	aggressiveness	15.185
82	44	costs	401.178	123	89	entropy	14.797
84	45	crashes	334.673	89	90	cross-section	14.076