

Bibliometric-Analysis.R

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2023-08-04

```
# Bibliometric Analysis (Findings Part 1)
```

```
#### Setting up the script ####
```

```
# Loading required packages
```

```
library(bibliometrix)
```

```
## Please note that our software is open source and available for use, distributed under the MIT license
```

```
## When it is used in a publication, we ask that authors properly cite the following reference:
```

```
##
```

```
## Aria, M. & Cuccurullo, C. (2017) bibliometrix: An R-tool for comprehensive science mapping analysis,
```

```
## Journal of Informetrics, 11(4), pp 959-975, Elsevier.
```

```
##
```

```
## Failure to properly cite the software is considered a violation of the license.
```

```
##
```

```
## For information and bug reports:
```

```
## - Take a look at https://www.bibliometrix.org
```

```
## - Send an email to info@bibliometrix.org
```

```
## - Write a post on https://github.com/massimoaria/bibliometrix/issues
```

```
##
```

```
## Help us to keep Bibliometrix and Biblioshiny free to download and use by contributing with a small d
```

```
##
```

```
##
```

```
## To start with the Biblioshiny app, please digit:
```

```
## biblioshiny()
```

```
library(igraph)
```

```
##
```

```
## Attaching package: 'igraph'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
## decompose, spectrum
```

```
## The following object is masked from 'package:base':
```

```
##
```

```
## union
```

```
library(stringr)
```

```
library(dplyr)
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:igraph':
```

```
##
```

```

##      as_data_frame, groups, union
## The following objects are masked from 'package:stats':
##
##      filter, lag
## The following objects are masked from 'package:base':
##
##      intersect, setdiff, setequal, union
library(xtable) # converts tables to latex tables
library(kableExtra)

##
## Attaching package: 'kableExtra'
## The following object is masked from 'package:dplyr':
##
##      group_rows
# Convert cleaned data to data frame
M <- convert2df(file="savedrecs.txt", dbsource="wos",format="plaintext")

##
## Converting your wos collection into a bibliographic dataframe
##
## Done!
##
##
## Generating affiliation field tag AU_UN from C1: Done!
# Reformat the row names (names of the review articles)
#input_string <- rownames(M)
#output_string <- gsub("([~],+)\|s[~],*\|s(\|d+).*", "\\|1 \|2", input_string)
#output_string[26] = "SHARIFI EI 2016"
#output_string[27] = "SHARIFI RSER 2016"
#rownames(M) = output_string

#### Descriptive Statistics ####

# Main bibliometric results of the articles
results = biblioAnalysis(M)
findings = summary(results,verbose=T)

# Most cited references
CR = citations(M,field="article",sep=";")
a=data.frame(cbind(CR$Cited[1:20])) # top 20 most cited references
colnames(a)[1] = "Number of Times Cited"
a %>%
  kbl(caption = "Top 20 Most Cited References") %>%
  kable_classic(full_width = F, html_font = "Cambria")

# xtable(a,auto=TRUE) #This produces Latex table

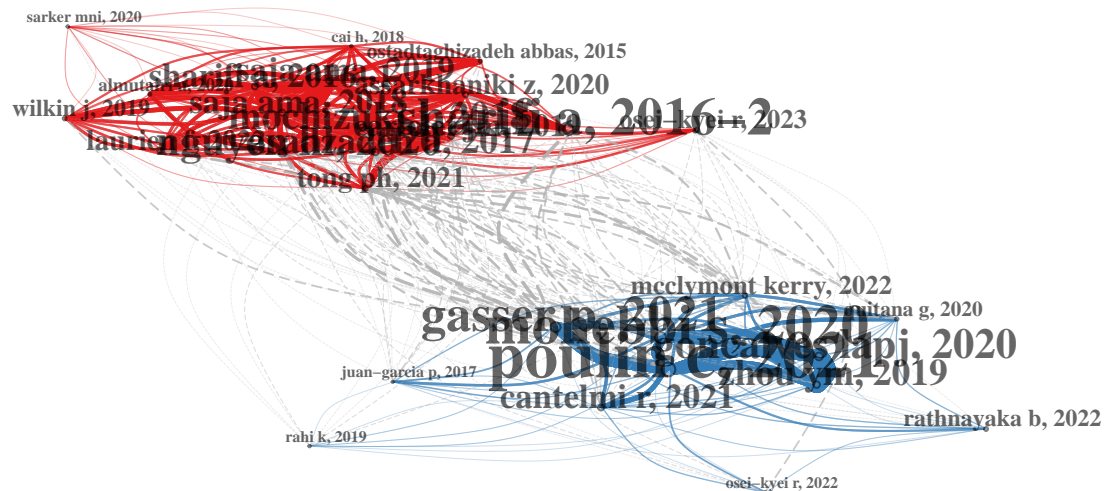
#### Bibliographic Coupling Network ####
NetMatrix <- biblioNetwork(M, analysis = "coupling", network = "references", sep = ";")
net=networkPlot(NetMatrix, Title = "Bibliographic Coupling Network", type = "auto", size.cex=TRUE, size=

```

Table 1: Top 20 Most Cited References

	Number o
BRUNEAU M, 2003, EARTHQ SPECTRA, V19, P733, DOI 10.1193/1.1623497	
CUTTER SL, 2010, J HOMEL SECUR EMERG, V7	
CUTTER SL, 2014, GLOBAL ENVIRON CHANG, V29, P65, DOI 10.1016/J.GLOENVCHA.2014.08.005	
HOLLING C.S., 1973, ANNUAL REV ECOL SYST, V4, P1, DOI 10.1146/ANNUREV.ES.04.110173.000245	
NORRIS FH, 2008, AM J COMMUN PSYCHOL, V41, P127, DOI 10.1007/S10464-007-9156-6	
HOSSEINI S, 2016, RELIAB ENG SYST SAFE, V145, P47, DOI 10.1016/J.RESS.2015.08.006	
CUTTER SL, 2008, GLOBAL ENVIRON CHANG, V18, P598, DOI 10.1016/J.GLOENVCHA.2008.07.013	
SHERRIEB K, 2010, SOC INDIC RES, V99, P227, DOI 10.1007/S11205-010-9576-9	
BURTON CG, 2015, ANN ASSOC AM GEOGR, V105, P67, DOI 10.1080/00045608.2014.960039	
CARPENTER S, 2001, ECOSYSTEMS, V4, P765, DOI 10.1007/S10021-001-0045-9	
CUTTER SL, 2016, NAT HAZARDS, V80, P741, DOI 10.1007/S11069-015-1993-2	
ADGER WN, 2000, PROG HUM GEOG, V24, P347, DOI 10.1191/030913200701540465	
FOLKE C, 2006, GLOBAL ENVIRON CHANG, V16, P253, DOI 10.1016/J.GLOENVCHA.2006.04.002	
FRAZIER TG, 2013, APPL GEOGR, V42, P95, DOI 10.1016/J.APGEOG.2013.05.004	
MEEROW S, 2016, LANDSCAPE URBAN PLAN, V147, P38, DOI 10.1016/J.LANDURBPLAN.2015.11.011	
ORENCIO PM, 2013, INT J DISAST RISK RE, V3, P62, DOI 10.1016/J.IJDRR.2012.11.006	
ROSE A, 2007, ENVIRON HAZARDS-UK, V7, P383, DOI 10.1016/J.ENVHAZ.2007.10.001	
AINUDDIN S, 2012, NAT HAZARDS, V63, P909, DOI 10.1007/S11069-012-0201-X	
ALEXANDER DE, 2013, NAT HAZARD EARTH SYS, V13, P2707, DOI 10.5194/NHESS-13-2707-2013	
JOERIN J, 2014, DISASTERS, V38, P540, DOI 10.1111/DISA.12058	

Bibliographic Coupling Network



```
# dotted lines means that there are still connections exist between papers, and the colored lines repre
summary(net) # Look at the net object
```

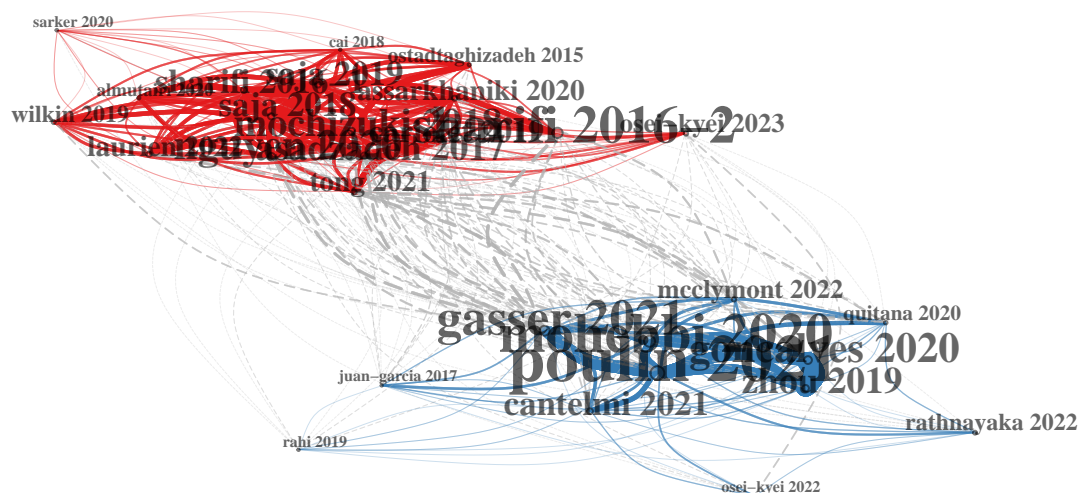
```
##           Length Class      Mode
## graph      29      igraph      list
## graph_pajek 29      igraph      list
## cluster_obj  2      communities list
## cluster_res  5      data.frame list
## community_obj 2      communities list
## layout      58      -none-      numeric
## S            0      -none-      NULL
## nodeDegree   2      data.frame list
## params       2      data.frame list
```

```
bibnet = net$graph # extract the igraph object and name it bibnet
V(bibnet)$label == V(bibnet)$name # make sure it is all true.
```

```
## [1] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [24] TRUE TRUE TRUE TRUE TRUE TRUE
```

```
nodenames = gsub(" [A-Za-z]+", "", V(bibnet)$name) # transform the names/labels and make it a bit nice
V(bibnet)$label = nodenames
V(bibnet)$name = nodenames
plot(bibnet) # plot it!
```

Bibliographic Coupling Network



```
# summary(bibnet)
```

```
#edgelist = as_edgelist(bibnet) # Look at it in an edgelist.
```

```
net$cluster_res
```

##	vertex	cluster	btw_centrality	clos_centrality	pagerank_centrality
## 1	osei-kyei r, 2023	1	6.805657e+00	0.02000000	0.01826661
## 2	laurien f, 2022	1	9.948300e+00	0.02083333	0.03422040
## 6	tong ph, 2021	1	3.518905e+01	0.02272727	0.03190060
## 8	assarkhaniki z, 2020	1	4.218850e+00	0.02000000	0.03713273
## 9	sarker mni, 2020	1	4.792412e-02	0.01562500	0.01247474
## 11	nguyen hl, 2020	1	2.573417e+01	0.02272727	0.04572943
## 14	almutairi a, 2020	1	9.908996e-01	0.01785714	0.03381351
## 16	cariolet jm, 2019	1	1.204345e+01	0.02083333	0.04365244
## 18	saja ama, 2019	1	3.819307e-01	0.01724138	0.05714217
## 19	wilkin j, 2019	1	1.900392e+00	0.01785714	0.02841970
## 20	cai h, 2018	1	7.882468e-03	0.01562500	0.02960671
## 21	saja ama, 2018	1	5.479310e+00	0.02083333	0.05327095
## 22	mochizuki j, 2018	1	1.260023e+01	0.02083333	0.04441923
## 23	asadzadeh a, 2017	1	1.134465e+01	0.02083333	0.04511708
## 25	sharifi a, 2016-1	1	1.231012e+00	0.01785714	0.04503408
## 26	sharifi a, 2016-2	1	9.815400e+00	0.02000000	0.02676718
## 28	ostadtaghizadeh abbas, 2015	1	2.608928e-01	0.01666667	0.02795806
## 3	osei-kyei r, 2022	2	1.328670e+00	0.01587302	0.01036352
## 4	rathnayaka b, 2022	2	4.378781e-01	0.01639344	0.01515358
## 5	poulin c, 2021	2	1.684061e+01	0.02127660	0.06687476
## 7	gasser p, 2021	2	1.048721e+02	0.02564103	0.04388268
## 10	mohebbi s, 2020	2	3.512567e+01	0.02439024	0.05211406
## 12	quitana g, 2020	2	1.328241e+01	0.02127660	0.02269349
## 13	goncalves lapj, 2020	2	7.545409e+00	0.02040816	0.05368759
## 15	zhou ym, 2019	2	3.140653e+00	0.01886792	0.04647882
## 17	rahi k, 2019	2	1.548293e+00	0.01754386	0.01005597
## 24	juan-garcia p, 2017	2	5.203251e+00	0.01818182	0.01672836
## 27	cantelmi r, 2021	2	6.003096e+00	0.02040816	0.02612079
## 29	mcclymont kerry, 2022	2	1.267192e+01	0.01960784	0.02092078

As stated in the slides, it is very interesting to see that with Louvain clustering, there are two cl

Have to make the cluster table by hand instead...

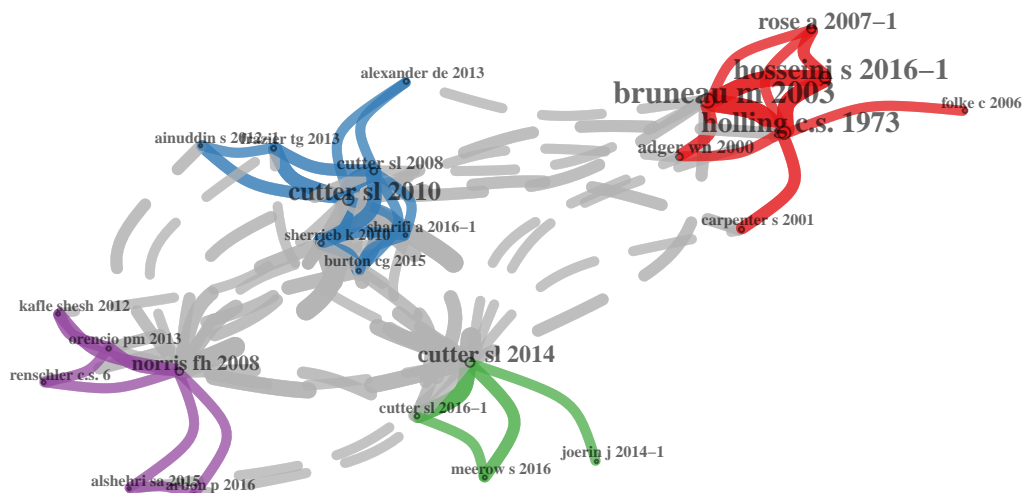
Co-citation network

Shows how the different cited references relate to each other

```
NetMatrix <- biblioNetwork(M, analysis = "co-citation", network = "references", sep = ";")
```

```
net=networkPlot(NetMatrix, Title = "Co-citation Network", type = "auto", size.cex=TRUE, size=3, remove.l
```

Co-citation Network



```
net$cluster_res
```

##	vertex	cluster	btw_centrality	clos_centrality	pagerank_centrality
## 1	bruneau m 2003	1	44.8222280	0.011235955	0.05851356
## 4	holling c.s. 1973	1	39.7402720	0.011111111	0.07967746
## 6	hosseini s 2016-1	1	0.0000000	0.003278689	0.04354785
## 11	carpenter s 2001	1	0.0000000	0.010416667	0.01820406
## 13	adger wn 2000	1	0.0000000	0.010416667	0.02621088
## 14	folke c 2006	1	0.0000000	0.003125000	0.01466059
## 18	rose a 2007-1	1	0.0000000	0.003278689	0.03299030
## 2	cutter sl 2010	2	78.8206711	0.012048193	0.09169578
## 8	cutter sl 2008	2	1.3737669	0.010752688	0.06462451
## 9	sherrieb k 2010	2	0.9230769	0.010416667	0.04883882
## 10	burton cg 2015	2	0.8076923	0.010416667	0.04279540
## 15	frazier tg 2013	2	0.0000000	0.010309278	0.03494226
## 19	ainuddin s 2012-1	2	0.0000000	0.009009009	0.02346809
## 20	alexander de 2013	2	0.0000000	0.008928571	0.02102661
## 24	sharifi a 2016-1	2	0.0000000	0.010309278	0.03648655
## 3	cutter sl 2014	3	153.3429290	0.012987013	0.06211060
## 12	cutter sl 2016-1	3	4.2796528	0.010526316	0.03566218
## 16	meerow s 2016	3	0.0000000	0.003278689	0.02725530
## 21	joerin j 2014-1	3	0.0000000	0.003257329	0.01556413
## 5	norris fh 2008	4	37.8897110	0.011363636	0.07135302
## 7	alshehri sa 2015	4	0.0000000	0.010000000	0.02712497
## 17	orencio pm 2013	4	0.0000000	0.010416667	0.04110449
## 22	kafle shesh 2012	4	0.0000000	0.009433962	0.02932166

```
## 23 renschler c.s. 6      4      0.0000000  0.009433962  0.02569594
## 25 arbon p 2016        4      0.0000000  0.010000000  0.02712497

# This network shows nodes that has at least 5 edges. This filter was chosen in order to reveal the mos
# In a way, this network is the flip side of bibliographic coupling network.

#### Author Keywords Analysis ####
NetMatrix <- biblioNetwork(M, analysis = "co-occurrences", network = "author_keywords", sep = ";")
net=networkPlot(NetMatrix, Title = "Author Keywords Co-occurrence Network", type = "auto", size.cex=TRUE)
```

Author Keywords Co-occurrence Network



```
net$cluster_res

##          vertex cluster btw_centrality clos_centrality pagerank_central
## 1          resilience      1    3199.6250000    0.0017857143    0.047391
## 2 critical infrastructure      1    395.2083333    0.0015479876    0.024199
## 15          disasters      1     24.5000000    0.0014880952    0.011758
## 38 criticality assessment      1     0.0000000    0.0014164306    0.008363
## 39 damage reduction      1     0.0000000    0.0014164306    0.008363
## 44 drinking water      1     0.0000000    0.0006887052    0.007114
## 49 flood risk management      1     0.0000000    0.0006802721    0.004820
## 59 literature review      1     0.0000000    0.0006887052    0.007114
## 60 management      1     0.0000000    0.0006896552    0.007646
## 62 metrics      1     0.0000000    0.0006887052    0.007114
## 65 natural hazard      1     0.0000000    0.0006887052    0.007114
## 69 performance measures      1     0.0000000    0.0006887052    0.007114
## 82 sendai framework      1     0.0000000    0.0014164306    0.008363
## 83 sewer systems      1     0.0000000    0.0006896552    0.007646
```

## 87	standards	1	0.0000000	0.0014224751	0.004231
## 93	system analysis	1	0.0000000	0.0006887052	0.007114
## 96	systems approach	1	0.0000000	0.0006802721	0.004820
## 98	transportation system	1	0.0000000	0.0006849315	0.006396
## 100	urban community	1	0.4666667	0.0014265335	0.004430
## 103	urban networks	1	0.0000000	0.0006849315	0.006396
## 104	urban transportation modes	1	0.0000000	0.0006849315	0.006396
## 106	wastewater	1	0.0000000	0.0006896552	0.007646
## 107	wrrf	1	0.0000000	0.0006896552	0.007646
## 3	community resilience	2	553.3000000	0.0015455951	0.017788
## 6	disaster management	2	657.4333333	0.0016260163	0.015445
## 7	disaster resilience	2	278.8000000	0.0013755158	0.015452
## 16	resilience assessment	2	2.0000000	0.0006882312	0.013831
## 26	big data	2	0.0000000	0.0015337423	0.006533
## 33	composite indicators building (cib)	2	0.0000000	0.0006146281	0.006529
## 51	governance	2	0.0000000	0.0015337423	0.006533
## 53	information visualisation	2	0.0000000	0.0006548788	0.008164
## 58	knowledge representation	2	0.0000000	0.0006548788	0.008164
## 67	operationalizing	2	0.0000000	0.0006146281	0.006529
## 74	resilience indicator	2	0.0000000	0.0006697924	0.006839
## 78	resilience modelling	2	0.0000000	0.0006548788	0.008164
## 80	risk management	2	0.0000000	0.0015337423	0.006533
## 94	systematic overview	2	0.0000000	0.0006548788	0.008164
## 95	systematic survey	2	0.0000000	0.0006146281	0.006529
## 4	criteria	3	482.8690476	0.0016366612	0.016406
## 8	indicators	3	138.5071429	0.0016103060	0.014994
## 11	assessment	3	321.0000000	0.0015576324	0.009588
## 21	adaptive capacity	3	0.0000000	0.0015337423	0.007990
## 23	assessment tool	3	0.0000000	0.0014619883	0.006209
## 25	awareness	3	0.0000000	0.0015337423	0.007990
## 29	climate change mitigation and adaptation	3	0.0000000	0.0015337423	0.008913
## 46	empirical studies	3	0.0000000	0.0015337423	0.007990
## 68	organizational resilience	3	0.0000000	0.0015337423	0.007990
## 70	principles	3	0.0000000	0.0015337423	0.008913
## 88	sustainability	3	0.0000000	0.0015337423	0.008913
## 99	uncertainties	3	0.0000000	0.0014619883	0.006209
## 102	urban energy	3	0.0000000	0.0015337423	0.008913
## 5	measurement	4	195.9071429	0.0016260163	0.015813
## 20	adaptive and transformative capacity	4	0.0000000	0.0015337423	0.006475
## 37	coping	4	0.0000000	0.0015337423	0.006475
## 41	databases	4	0.0000000	0.0015337423	0.009643
## 45	earthquakes	4	0.0000000	0.0015337423	0.009643
## 48	enhancement	4	0.0000000	0.0015337423	0.009643
## 52	hurricanes	4	0.0000000	0.0015337423	0.009643
## 79	risk drivers	4	0.0000000	0.0015337423	0.006475
## 81	roads	4	0.0000000	0.0015337423	0.009643
## 97	transportation	4	0.0000000	0.0015337423	0.009643
## 9	resilience indicators	5	603.0333333	0.0015384615	0.015415
## 13	disaster risk management	5	98.2000000	0.0007272727	0.014750
## 18	urban resilience	5	56.0000000	0.0013698630	0.012382
## 22	assessment framework	5	0.0000000	0.0006720430	0.008588
## 24	assessment tools	5	0.0000000	0.0006993007	0.007275
## 30	climate-related disasters	5	0.0000000	0.0013623978	0.006075
## 31	coastal communities	5	0.0000000	0.0006720430	0.008588

## 32	coastal hazards	5	0.0000000	0.0006720430	0.008588
## 35	conceptual diversity	5	0.0000000	0.0006993007	0.007275
## 50	gis	5	0.0000000	0.0006451613	0.007306
## 61	map	5	0.0000000	0.0006451613	0.007306
## 63	models indices and domains	5	0.0000000	0.0006993007	0.007275
## 101	urban ecosystems	5	0.0000000	0.0013623978	0.006075
## 10	review	6	287.5333333	0.0015723270	0.020913
## 17	resilience measurement	6	242.4000000	0.0015243902	0.009887
## 27	biophysical resilience functions	6	0.0000000	0.0015337423	0.006921
## 28	climate	6	0.0000000	0.0006747638	0.009276
## 34	comprehensive risk management	6	0.0000000	0.0006747638	0.009276
## 42	decision support	6	0.0000000	0.0006747638	0.009276
## 47	energy systems	6	0.0000000	0.0015337423	0.006921
## 55	infrastructure resilience	6	0.0000000	0.0013550136	0.006910
## 64	multiple hazard	6	0.0000000	0.0006747638	0.009276
## 73	resilience engineering	6	0.0000000	0.0013550136	0.006910
## 76	resilience management	6	0.0000000	0.0013550136	0.006910
## 91	swoosh resilience curve	6	0.0000000	0.0015337423	0.006921
## 12	community disaster resilience	7	329.1666667	0.0007473842	0.010261
## 14	disaster risk reduction	7	1035.0500000	0.0016286645	0.010333
## 36	connectivity	7	0.0000000	0.0007002801	0.009510
## 40	data	7	0.0000000	0.0007002801	0.009510
## 56	innovation	7	0.0000000	0.0007002801	0.009510
## 84	social network analysis	7	0.0000000	0.0007002801	0.009510
## 85	social network mapping	7	0.0000000	0.0007002801	0.009510
## 86	social networks	7	0.0000000	0.0007002801	0.009510
## 19	adaptation strategies	8	0.0000000	0.0250000000	0.009345
## 43	disaster resilience measurement	8	0.0000000	0.0250000000	0.009345
## 75	resilience indices	8	0.0000000	0.0250000000	0.009345
## 92	synthesis analysis	8	0.0000000	0.0250000000	0.009345
## 105	validation	8	0.0000000	0.0250000000	0.009345
## 54	infrastructure interdependency	9	0.0000000	0.0333333333	0.009345
## 57	integration	9	0.0000000	0.0333333333	0.009345
## 77	resilience metrics	9	0.0000000	0.0333333333	0.009345
## 89	sustainable design strategies	9	0.0000000	0.0333333333	0.009345
## 66	objective indicator	10	0.0000000	0.0333333333	0.009345
## 71	quantitative measurement	10	0.0000000	0.0333333333	0.009345
## 72	resilience dimension	10	0.0000000	0.0333333333	0.009345
## 90	sustainable development goals (sdgs)	10	0.0000000	0.0333333333	0.009345

Using Louvain clustering, 11 clusters were identified. It's interesting to see that that there are th

Abstract Co-occurrence Network, 1-gram

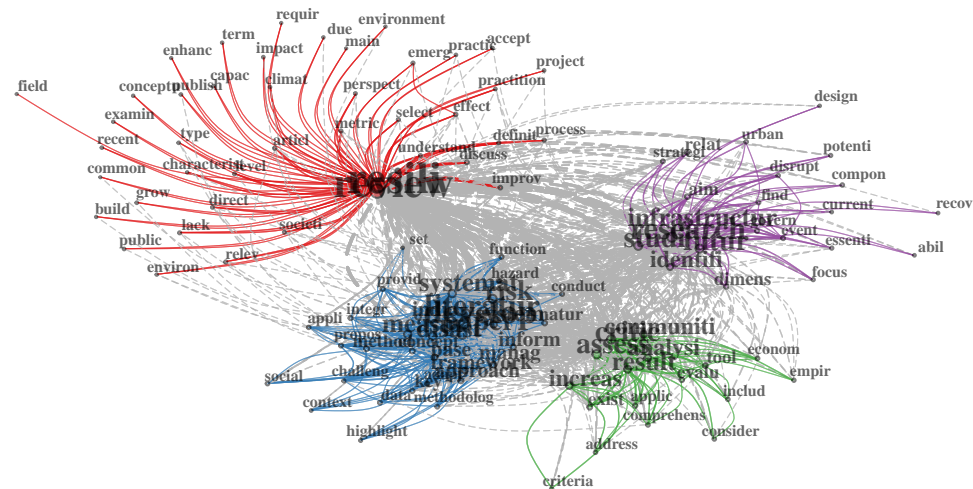
```
abstract_terms = termExtraction(M, Field="AB", ngrams=1, verbose=F, stemming = TRUE)
NetMatrix <- biblioNetwork(abstract_terms, analysis = "co-occurrences", network = "abstracts")
summary(networkStat(NetMatrix))
```

```
##
##
## Main statistics about the network
##
## Size 828
## Density 0.189
## Transitivity 0.479
## Diameter 2
```

```
## Degree Centralization 0.811
## Average path length 1.811
##
```

```
net=networkPlot(NetMatrix, Title = "Abstract Co-occurrence Network (1-gram)", type = "auto", size.cex=TR)
```

Abstract Co-occurrence Network (1-gram)



```
net$cluster_res
```

##	vertex	cluster	btw_centrality	clos_centrality	pagerank_centrality
## 1	resili	1	1.886113e+03	0.0043859649	0.072272185
## 4	review	1	1.576507e+03	0.0043103448	0.069147247
## 40	articl	1	7.542617e-02	0.0028490028	0.003764447
## 46	climat	1	0.000000e+00	0.0025380711	0.003206164
## 49	improv	1	1.793535e+00	0.0030674847	0.004985659
## 53	capac	1	0.000000e+00	0.0007374631	0.003155818
## 55	definit	1	3.695465e-01	0.0029585799	0.004460501
## 58	metric	1	2.705140e-02	0.0029411765	0.004161237
## 61	characterist	1	0.000000e+00	0.0026809651	0.003596616
## 63	conceptu	1	0.000000e+00	0.0007374631	0.003155818
## 64	direct	1	0.000000e+00	0.0027700831	0.003655978
## 69	level	1	0.000000e+00	0.0024096386	0.003594001
## 70	process	1	1.705548e-01	0.0030487805	0.003905200
## 72	select	1	9.449310e-02	0.0029154519	0.003758533
## 75	discuss	1	4.008754e-01	0.0029940120	0.004664688
## 76	effect	1	3.064362e-01	0.0028901734	0.004984538
## 79	lack	1	0.000000e+00	0.0026525199	0.003638360
## 80	perspect	1	0.000000e+00	0.0027700831	0.003652690

## 82	type	1	4.740834e-02	0.0026455026	0.003255827
## 83	build	1	0.000000e+00	0.0025974026	0.003206024
## 87	emerg	1	0.000000e+00	0.0027548209	0.004349036
## 88	enhanc	1	0.000000e+00	0.0007374631	0.003155818
## 89	environ	1	0.000000e+00	0.0028490028	0.003268921
## 90	examin	1	0.000000e+00	0.0007374631	0.003155818
## 93	practition	1	2.658396e-02	0.0028409091	0.003744023
## 94	relev	1	8.863616e-02	0.0028248588	0.003540993
## 95	understand	1	1.547085e-01	0.0028985507	0.004248204
## 97	common	1	0.000000e+00	0.0024752475	0.003204501
## 98	due	1	0.000000e+00	0.0025906736	0.003219344
## 100	grow	1	0.000000e+00	0.0024875622	0.003221032
## 101	practic	1	0.000000e+00	0.0028490028	0.003289218
## 102	societi	1	7.870220e-02	0.0029069767	0.003659103
## 103	term	1	0.000000e+00	0.0007374631	0.003155818
## 104	accept	1	0.000000e+00	0.0027472527	0.003275143
## 105	environment	1	0.000000e+00	0.0025974026	0.003206024
## 106	field	1	0.000000e+00	0.0007363770	0.002233999
## 107	impact	1	0.000000e+00	0.0025380711	0.003206164
## 108	main	1	0.000000e+00	0.0007374631	0.003155818
## 109	project	1	0.000000e+00	0.0027100271	0.003289124
## 110	public	1	0.000000e+00	0.0026455026	0.003218714
## 111	publish	1	0.000000e+00	0.0024813896	0.003205287
## 112	recent	1	0.000000e+00	0.0007374631	0.003155818
## 114	requir	1	0.000000e+00	0.0007374631	0.003155818
## 2	measur	2	4.976473e+01	0.0033112583	0.023936567
## 5	disast	2	1.851251e+01	0.0033444816	0.022351416
## 6	framework	2	4.911686e+00	0.0032786885	0.011134834
## 7	indic	2	1.555279e+00	0.0032467532	0.013435741
## 10	system	2	2.658428e+01	0.0033444816	0.015565075
## 12	develop	2	5.959670e+02	0.0036764706	0.031411601
## 13	social	2	2.446795e-02	0.0030303030	0.003963786
## 15	risk	2	4.185092e+02	0.0035587189	0.020754278
## 16	paper	2	2.214970e+02	0.0034482759	0.023489080
## 17	approach	2	6.927247e-01	0.0032258065	0.008626303
## 19	literatur	2	1.584770e+02	0.0034364261	0.022428832
## 22	concept	2	1.480119e+00	0.0032467532	0.011970947
## 25	systemat	2	8.803968e+01	0.0034013605	0.017823401
## 27	method	2	7.884331e+00	0.0032362460	0.013165973
## 29	adapt	2	3.096725e-02	0.0030769231	0.010187206
## 30	key	2	6.129000e-01	0.0032362460	0.007487444
## 31	manag	2	6.368753e+00	0.0032573290	0.012865413
## 34	data	2	1.873327e-02	0.0030303030	0.005951352
## 37	base	2	1.295138e+01	0.0033222591	0.017006061
## 39	methodolog	2	1.254333e-01	0.0031055901	0.006745474
## 41	context	2	1.376322e-02	0.0029498525	0.004552277
## 44	propos	2	4.707550e-02	0.0030395137	0.008550923
## 50	inform	2	9.740155e-01	0.0032051282	0.009656746
## 52	appli	2	2.446795e-02	0.0030303030	0.006043911
## 54	conduct	2	4.330643e-01	0.0031645570	0.004664405
## 56	hazard	2	1.504177e-01	0.0031645570	0.006411652
## 57	integr	2	8.684236e-02	0.0031446541	0.006619108
## 59	natur	2	1.065418e+00	0.0032467532	0.006775526
## 60	challeng	2	2.446795e-02	0.0029585799	0.007894663

## 67	function	2	2.307589e-01	0.0031446541	0.005262398
## 71	provid	2	1.066446e-01	0.0031746032	0.007154765
## 92	highlight	2	1.376322e-02	0.0029498525	0.004033498
## 115	set	2	9.557793e-03	0.0029498525	0.002970970
## 3	assess	3	4.574140e+02	0.0036630037	0.020313587
## 9	communiti	3	3.557327e+01	0.0034722222	0.010707603
## 20	critic	3	4.410049e+02	0.0037735849	0.019934546
## 21	analysi	3	2.870307e+01	0.0034246575	0.014484035
## 28	tool	3	1.148345e+00	0.0032362460	0.006499533
## 32	result	3	2.867380e+01	0.0034602076	0.015415723
## 33	criteria	3	1.376322e-02	0.0030487805	0.003131665
## 35	increas	3	1.583030e+02	0.0034602076	0.012582537
## 36	address	3	2.302738e-01	0.0031746032	0.003830389
## 43	evalu	3	1.155294e+02	0.0033112583	0.007395979
## 47	comprehens	3	1.516501e+00	0.0031746032	0.005977661
## 48	exist	3	2.568085e+00	0.0032258065	0.006822721
## 68	includ	3	1.152854e+00	0.0031055901	0.006133604
## 73	applic	3	2.038524e+00	0.0032786885	0.006573973
## 77	empir	3	7.869119e-02	0.0031948882	0.003801027
## 84	consider	3	1.873327e-02	0.0031446541	0.003941907
## 86	econom	3	2.792986e-01	0.0031948882	0.005067337
## 8	research	4	6.619296e+02	0.0038461538	0.028433757
## 11	studi	4	2.281553e+02	0.0036764706	0.018605965
## 14	urban	4	2.623887e-01	0.0031347962	0.003953533
## 18	infrastructur	4	9.890378e+01	0.0035587189	0.020084957
## 23	futur	4	1.827714e+02	0.0036630037	0.023675894
## 24	identifi	4	1.155874e+01	0.0034013605	0.009538233
## 26	dimens	4	1.288670e+00	0.0032573290	0.006073285
## 38	aim	4	2.997426e+00	0.0033222591	0.010029865
## 42	disrupt	4	1.605709e-02	0.0031055901	0.005753812
## 45	strategi	4	6.442291e-01	0.0031847134	0.004076650
## 51	relat	4	9.016614e-01	0.0031545741	0.004074700
## 62	compon	4	1.376322e-02	0.0030487805	0.003670750
## 65	event	4	5.229286e-01	0.0032573290	0.007423593
## 66	find	4	5.617038e-01	0.0031746032	0.005893936
## 74	current	4	1.376322e-02	0.0029850746	0.003789500
## 78	govern	4	5.396681e-01	0.0031746032	0.004767396
## 81	potenti	4	1.605709e-02	0.0029498525	0.003639136
## 85	design	4	9.557793e-03	0.0029498525	0.002885715
## 91	focus	4	6.594763e-02	0.0030959752	0.003944894
## 96	abil	4	9.557793e-03	0.0029498525	0.002892331
## 99	essenti	4	1.298246e-01	0.0031347962	0.003951226
## 113	recov	4	0.000000e+00	0.0029325513	0.002800149

I'm debating if this should be included... I don't think so, because the previous network already does.

Unused Codes

Abstract Co-occurrence Network, 2-grams

abstract_terms = termExtraction(M, Field="AB", ngrams=2, verbose=F, stemming = TRUE) # need to first

NetMatrix <- biblioNetwork(abstract_terms, analysis = "co-occurrences", network = "abstracts")

net=networkPlot(NetMatrix, Title = "Abstract Co-occurrence Network", type = "auto", size.cex=TRUE, siz

#

Co-word Analysis through Scopus Keyword co-occurrences

```

# NetMatrix <- biblioNetwork(M, analysis = "co-occurrences", network = "keywords", sep = ";")
# net=networkPlot(NeMatrix, normalize="association", n = 50, Title = "Scopus Keywords Co-occurrences",
#
# # Historiograph. Better not to use this, because it is rather broken
# histResults = histNetwork(M, sep=";", min.citations = 1)
# net = histPlot(histResults,size=5,labelsiz=4) # a lot of papers are not shown because there aren't a
# histResults$histData # shows direct citation results
# # bibliometrix::histPlot #shows the source code for histPlot
#
# # Multiple Correspondence Analysis through abstracts
# suppressWarnings(
#   CS <- conceptualStructure(M, method="MCA", field="AB", minDegree=11, clust="auto", stemming=T, labe
# ) #minDegree of 10 works the best
#
# # Metric Multidimensional Scaling through abstracts
# suppressWarnings(
#   CS <- conceptualStructure(M, method="MDS", field="AB", minDegree=18, clust="auto", stemming=T, labe
# ) #minDegree of 10 works the best
#
# # Shiny Interface
# biblioshiny() #to use the Shiny interface.

```