## Topic Modeling Classification of Publications on CSR and Technology Transfer

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#### The dataset

- ➤ On June 6, 2022, the Web of Science (WoS) database was searched by selecting the following keywords and Boolean commands:
  - ► Title: ("technology transfer" OR "corporate social responsibility") AND ("management" OR "public policy") OR
  - ► Abstract: ("entrepreneur" OR "firm" OR "responsible innovation")
- ➤ Thus, 529 records (**publications** or **articles**) were retrieved, attributed in 70 fields (columns), among which (11 are shown in a sample of the dataset in next slide) were retained for the present study.

### A sample of the dataset

	Туре	Title	Source	Authors	Editors	Year	Author Keywords	Keywords Plus	Abstract	Times Cited, WoS Core	Times Cited, All Databases
481	J	Measuring systematic-risk using implicit beta	Manage. Sci.	SIEGEL, AF	NaN	1995	EXCHANGE OPTION; VOLATILITY AND RETURN IN FINA	NaN	A new technology is proposed for estimating th	18	18
487	J	Corporate social responsibility and financial	Strateg. Manage. J.	McWilliams, A; Siegel, D	NaN	2000	corporate social responsibility; firm performa	R-AND-D; INDUSTRY MATTER; INVESTMENT; EVENT; P	Researchers have reported a positive, negative	1739	1775
8	J	Corporate social responsibility and management	J. Bus. Ethics	Lee, D	NaN	2017	Corporate social responsibility; Disclosure re	SARBANES-OXLEY ACT; STAKEHOLDER MANAGEMENT; EA	This study examines the association between co	32	32
355	J	Management of technology-transfer - can it be	NaN	DEBETTIGNIES, HC	NaN	1978	NaN	NaN	NaN	0	0
468	J	Lessons learned for a more efficient knowledge	Waste Manage. Res.	Bezama, A; Szarka, N; Navia, R; Konrad, O; Lor	NaN	2007	municipal solid waste management; industrial w	TANNERY	The present paper describes the development, p	5	5
510	J	Issues in the use of the event study methodolo	Organ. Res. Methods	McWilliams, A; Siegel, D; Teoh, SH	NaN	1999	NaN	NaN	Organizational researchers are increasingly us	67	68

#### Motivation

- ▶ In bibliographic social network analysis, besides the common cases of citation (co–citation and bibliographic coupling) and co–authorship networks, some authors have also studied studied keyword co–occurrence networks (Maltseva & Batagelj [2018, 2019], Leydesdorff et al. [2008], Groenewegen et al. [2015]).
- ► Typically, WoS catalogues two types of keywords for each archived article (called *publication* from now on):
  - Author Keywords: chosen by the author to best reflect the content of the document, and
  - Keywords Plus: index terms automatically generated from the titles of cited articles.
- ▶ Instead of these, here, we are using a *Machine Learning* and *Natural Language Processing* approach in order to assign the hiden semantic structures of a text body, discovered by a **Topic Model**, as another type of keywords, which are endogenously extracted from the text (or abstract) of publications (independently of exogenous allocations made by authors or archivers).

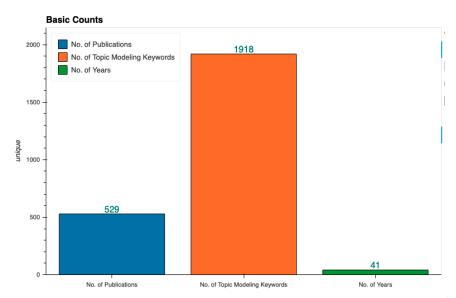
#### Topic Modeling Classification, I

- ► Suppose that we have a collection of documents, each one being composed of nontrivial words or phrases.
- ▶ Here, each document is a publication and we are only considering words in the abstract. However, in a future furthering of our study, we intend to consider the whole text of each publication.
- ► **Topic Modeling** is an unsupervised machine learning technique that proceeds in two stages.
- ▶ In the first stage, after scanning the set of documents, Topic Modeling produces a vocabulary of pre-processed ("cleaned," uncapitalized, lemmatized etc.) words (or phrases).
- ► Here, these words will be called **Topic Modeling keywords**.

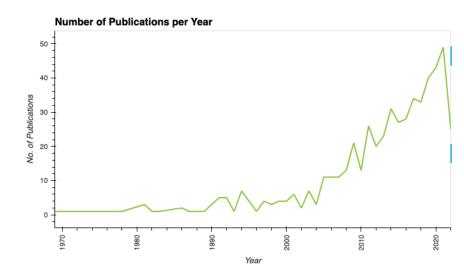
#### Topic Modeling Classification, II

- ▶ In the second stage, using Latent Dirichlet Allocation (LDA), which is a generative statistical model, what Topic Modeling does is the following:
- ▶ Instead of representing a document in its feature space (by considering frequencies of words in each document), it clusters (classifies) all the words of the vocabulary in a *topic space*, consisting of a given number of *Topics*, i.e., groups of words that are associated under a single interpretable theme, and it assess two types of probabilities:
  - ▶ the weights of words in each document to be assigned to each topic, and
  - the strength with which each document exhibits each one of the topics.
- ► Thus, according to the weights that words in a document possess in order to be assigned to each topic, for each document, there exists a dominant topic and a corresponding probability contribution for the document to be assigned to a dominant topic.

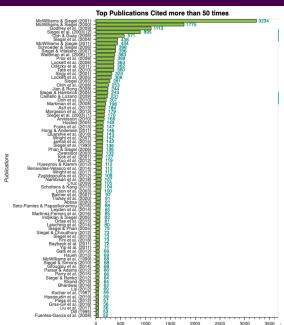
#### The dataset of publications from 1969 to 2022



### Publications per Year

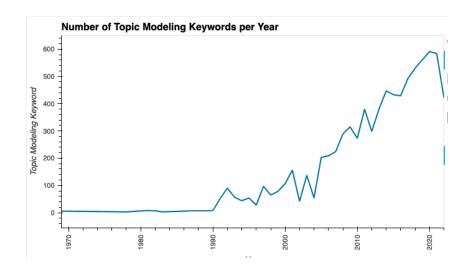


## Top Publications in Times Cited

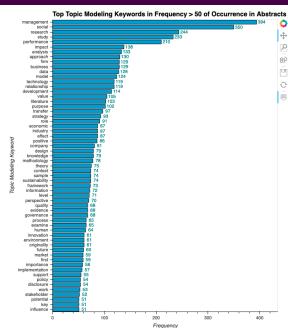


900

#### Topic Modeling Keywords per Year

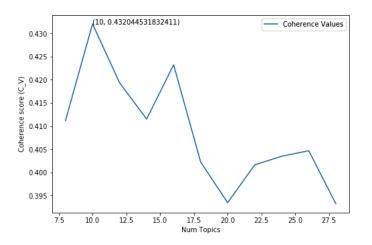


## Top Topic Modeling Keywords



200

# Why 10 Topics? Maximization of the CV coherence score



#### Wordclouds of Topics and Topic Modeling Keywords

Topic Modeling of the Web of Science bibliographic dataset on CSR and Technology Transfer

Topic 1 (Governance) bonding methodology governance fee awareness batterv environment tax forecast accounting listing security evidence action sophistication tirm Topic 4 (Management Models) case analysis development system project program transfer management technology research paper process study model ਸਲਾਦ innovation Topic 7 (Work Organization) change management Topic 10 (Firm Performance) management value performance ownership image impression 1 r Meffect

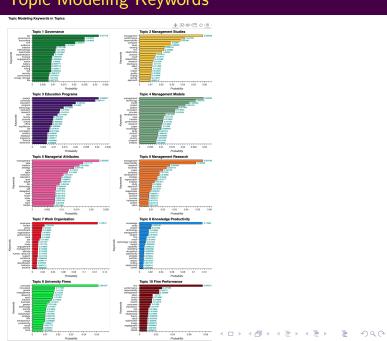
Topic 2 (Management Studies) effect practice research managementimpact firm stakeholder approach responsibility relationshipfinding disclosure \_activity performance company studv earning result quality Topic 5 (Managerial Attributes) self management Topic 8 (Knowledge Productivity) retailer student

## knowledge result project

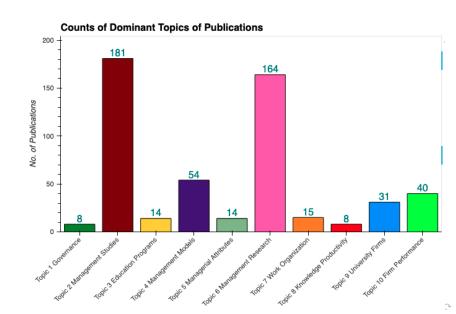
management recognition
marketing market
relationship firm opportunity
study technology\_transfer
capability technology
finding

Topic 3 (Education Programs) Topic 6 (Management Research) researchapproach paper strategy study <sup>system</sup>analysis Topic 9 (University Firms) business innovation industrystakeholder technologygrowth study university management activity sportfirmresearch

## Weights of Topic Modeling Keywords

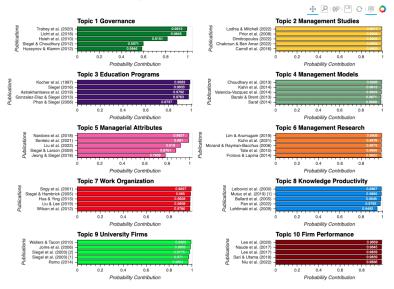


## Dominant Topics of Publications

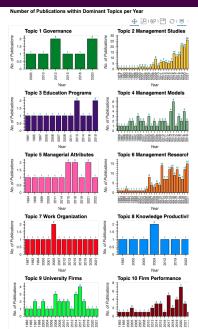


#### Top Dominant Topic Publications in Probability Contribution

Top Dominant Topic Publications in Probability Contribution



#### Number of Publications within Dominant Topics per Year



#### Top Dominant Topic Publications in Times Cited

#### Top Dominant Topic Publications in Times Cited

