# Discerning Industrial Networks, Clusters and Competences – An Alternative View Using Web Mining Techniques

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### Outline

- Problem areas in industrial cluster research
- Use of the web to find industrial activity
- Comparison with conventional methods
- Shared competencies and overlaps between clusters
- Finding industrial networks using the web
- Analysis of linkages
- Concluding remarks

- Problem areas in industrial cluster research
  - The general problem of finding enough companies and other contributors to an economic milieu
  - The definition of an economic entity
  - Shortcomings of Standard Industry Classification (SIC) codes to describe company activity because:
    - > SICs have difficulty capturing 'new' industries
    - > They are often simply inaccurate
    - > Many firms have multiple lines of business

The net result is that many actors in a practical industrial cluster study are not found when using company lists with SICs as the sole indicator of activity

Why not use the web to find out 'what goes on' in industry in a defined locale?

In this context the Research Question becomes:

Can the use of the web as an information resource add anything useful to more conventional methods of researching industrial clusters and networks?

- Use of the web to find industrial activity
  - Acquire a large number of URLs of appropriate firms and supporting actors in a defined locale
    - Credit reference database lists/Chamber of commerce/trade associations/other
  - Full text spidering of all the URLs with data then written to a linked database
- Metrics
  - Region North East of England
  - o 14000 URLs used
  - 177432 keywords

- Comparison with conventional methods
  - Used a subset of 8518 company records of known provenance (from DNB)
  - Compared the results of full text spidering with fields containing SIC based data
  - Not surprisingly whole text searching in a closed set of websites shown to be far superior for finding activity than by searching on SIC based activity alone (Table 1)

### • Table 1 - Test of keyword frequency by number of firms

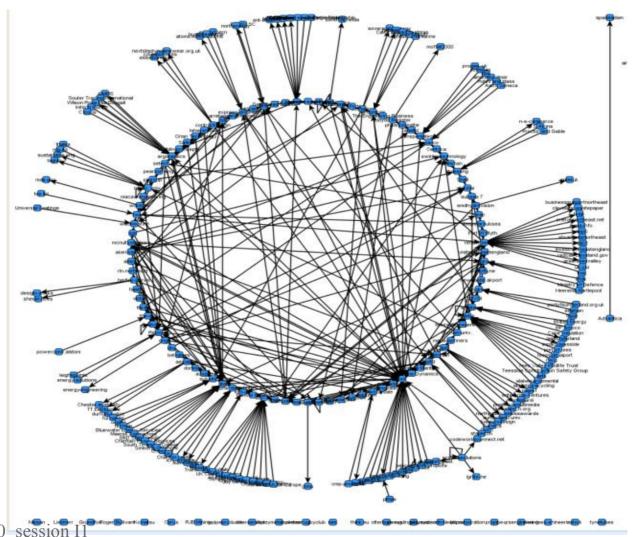
Word	Count by DNB 'Main Activity'	Count by any DNB 'SIC text'	Count by Keyword
Motorsport	0	0	11
Motor racing	0	0	9
Motorcycle	3	24	38
Automotive	10	8	104
Defence	0	0	48
Military	2	0	25
Offshore	5	0	105
Subsea	1	0	15
Yacht	2	0	13
Sail	1	0	9
boat	6	8	36
TOTALS	30	40	413

- Overlaps between clusters and shared competencies
  - Looked at specific groups of engineering firms
    - > 1502 in Marine, Offshore, Subsea, Defence and Environmental tech.
  - o Firms had 'membership' of more than one group
  - o Commonalities of capability across sectors
    - manufacture of medium and large sized artefacts such as for example ships, armaments, rigs and subsea ploughs.
    - Particular capability in engineering for more hostile environments

- Finding industrial networks using the web (1)
  - Assumption made that a link exists if, on a company website:
    - > a firm puts a reference to another organisation on its own website as a clickable link (found by a short search program)
    - there are visual clues to other organisations on a company website e.g. a client list (carried out by inspection)
    - > an external firm references the URL being looked at, so called in-links (using an in-link finder such as www.linkpopularity.com)

- Finding industrial networks using the web (2)
  - o Base URLs taken from a subsea engineering grouping identified from earlier work i.e. organisations with the word 'subsea' anywhere on their website
  - Connectivity only within the N.E. region
  - Only looked at linkages and direction of knowledge or trades, no strength inferred
  - Some companies had large numbers of in-links therefore some selective filtering was applied
  - Other quality checks applied manually
  - o 3 types of links merged into single matrix

Figure 1 — Subsea network, circular representation



- Analysis of linkages
  - Simple matrix output analysed using conventional graphing methods – circular representation (Figure 1)
  - Linear network (Figure 2)
  - Sample egonet (Figure 3)
  - Metrics
    - > 282 nodes
    - > 394 intraregional links
    - > Average geodesic distance 4.306
    - ➤ Overall clustering co-efficient 0.059

Figure 2. Subsea Network (isolates removed)

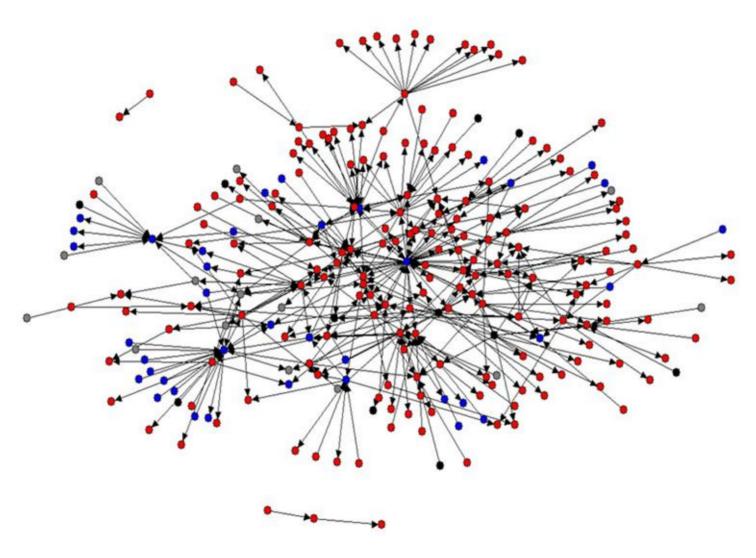
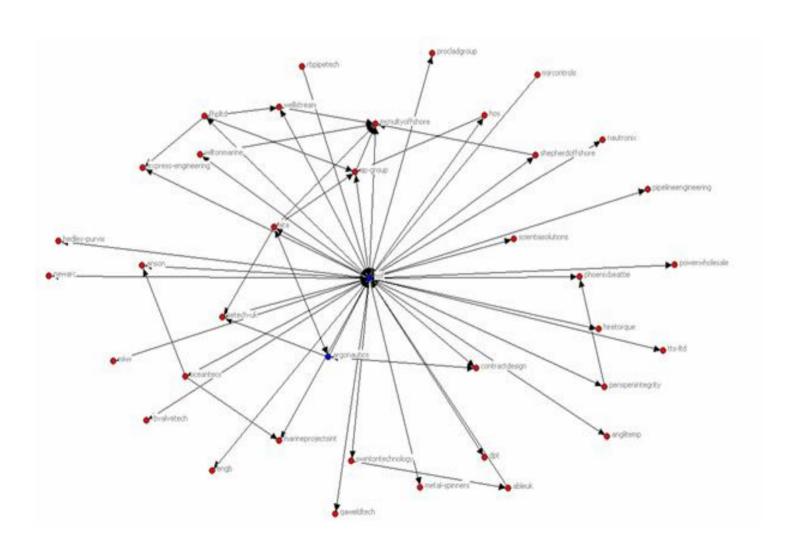


Figure 3. Egonet for a trade association in Oil & Gas sector



- Concluding remarks (1)
  - Comparison made with nets derived from the web and one by snowball sampling using the same firm population
    - > About 50% match although more links found by web based means
  - Method works well for finding:
    - > Specialised groups of firms by keyword searches
    - Linkages in the absence of co-operation from individual companies
    - > Key players and gatekeepers of knowledge

## • Concluding remarks (2)

- Problem areas/more research required
  - > Acquisition of large numbers of seed URLs
    - The presence of directories can confuse and frustrate attempts to automate this process
  - Dependency on the quality of individual websites as regards basic firm information and the number of linkages to or from the site
  - Very large sites take a long time to spider and may not be appropriate
  - Problem of definition of the firm or other contributing or supporting economic entity
  - > The 'nature' of links

# Thanks for your attention

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