

A Comparison of Algorithms for Hypertext Notes Network Linearisation

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2. Hypertext linearisation

There are two main occasions where hypertext linearisation can be of use in supporting a reader browsing through a hypertext database and in providing a version of a hypertext linearisation for -

2.1 Supporting browsing

A reader browsing through a hypertext may not always wish to choose or be able to follow an informed choice of working paths to follow. A hypertext display typically shows the names of the successor nodes, but will not tell the reader whether a link is important, nor whether it leads to a familiar or new part of the network, or just to a dead end. A linearisation author could provide a default linearisation path for the hypertext reader, selected according to criteria such as the reader's interests, for example a path containing closely related topics, the reader's abilities, for example a path containing an explanation and directions linked to the reader's knowledge, or a given point of view, for example a path chosen to represent one side of an argument.

2.2 Hypertext to linear text

There are many reasons to print out a copy of a hypertext database, such as to create a record for completeness, to provide a paper copy in a report or dissertation, or to produce a book version of a hypertext reference work. The printed version may still indicate the links between nodes as conventional text references, but the hypertext must be linearised to fit the format of a printed work, with pages ordered into sequential order. A simple author tool, such as a system that orders the nodes in order of time of creation, or in alphabetical order, may be acceptable for some purposes, but others, such as producing a printed reference book with a preferred linearisation, may require a more sophisticated linearisation.

Linear texts have served as the main means of storing and transferring knowledge over the past five thousand years and there is no sign that printed books, libraries and journals are about to be superseded. For most purposes, there is no need to present the reader with a hypertext, writers want to be in control of the order in which a text is read, and readers are content to be carried along by a narrative flow.

One use of hypertext is, instead of giving a choice to a reader, to assist the author of a conventional text, by providing links to Buzan's [] or notes networks. These [] act as interrelated representations between ideas, associations and draft text. A productive way for a writer to generate new material is to follow ideas through associations, setting down each new idea as a note, and linking the notes together to form a visual map of the topics to be covered. The use of notes networks as an aid to study and writing has been advocated by Buzan [] and Ulfan [] and others. Creativity comes in part from exploration of conceptual space of related ideas. Boden []

An interconnected network of notes representing ideas or topics can be a very creative and explicit.

Xerox NoteCards [] is a computer based presentation of a notes network. A user can brainstorm ideas, write each one on a suitable card, and print the cards together rapidly on the screen. Two or more people can work with the same set of

during the creation of the network. Information potentially available to the author includes node names, node contents, node names, connectivity, time of creation and spatial layout.

It should be deterministic

The order of linearisation should be fully determined by the information contained in the network, not by the processing priorities of the program language.

It should produce a linearisation which is acceptable to a human writer.

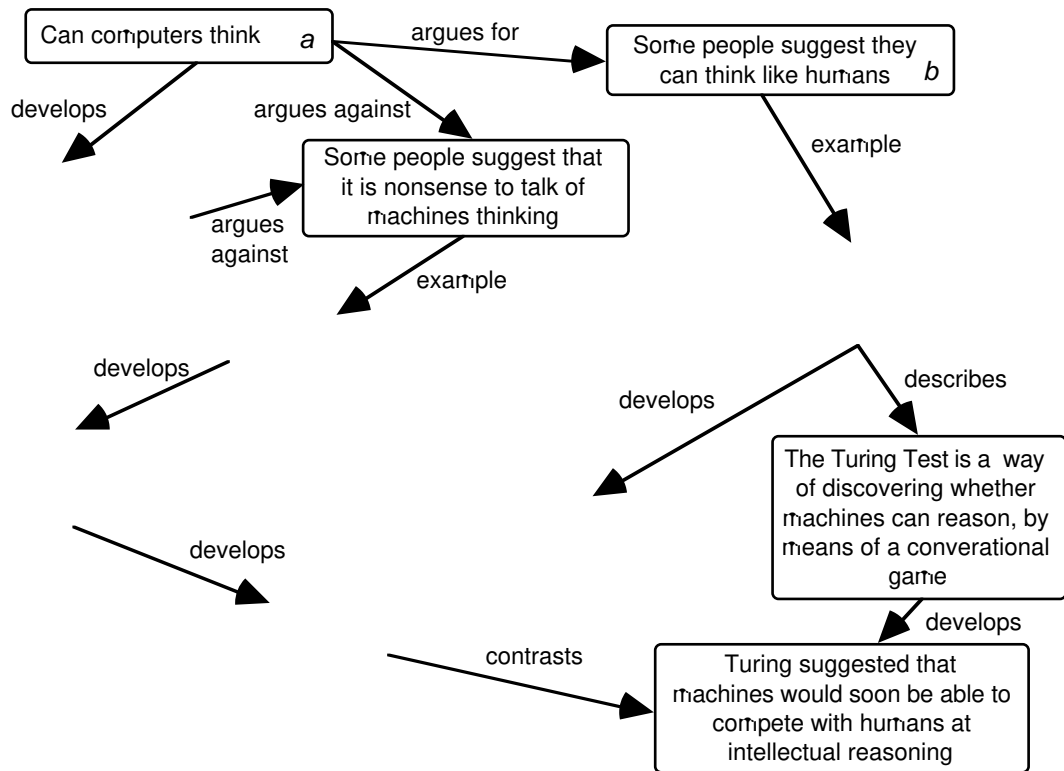
A linearisation author differs from a search author in that the criterion for success is not to reach a specified goal, but to produce an order of nodes which meets the expectations of a reader. A coherent/easy-to-read narrative leads the reader onwards by following trails of association that resonate with the reader's experience. Thus, it follows that a linearisation author can only be judged by subjective criteria.

3. Descriptions of the algorithms

The experiments described in this paper compare two algorithms for hypertext linearisation. Both algorithms satisfy the requirements above, apart from the deterministic criterion. The best first algorithm uses four heuristics to resolve the priority of nodes, and other information such as the time of node creation can be employed to ensure the algorithm is deterministic. They treat a hypertext as a directed, labeled graph. They ignore the content of the nodes and so can also be used to linearise a network containing non-textual elements, but the use of the priority of the nodes type. The algorithms have both been implemented in Prolog.

Both algorithms can be implemented to run in time $O(nl)$ where n is the number of nodes in the network and l is the mean number of nodes from each node. Although l will depend on

node are included before the ones - its extension was not presented for the experiments described below



-Find all untraversed nodes from each node in LINEA I ED - and move each node from the rap -

-Merge the nodes with O EN [lowest value] to the front -

a-If there are two or more candidate nodes with the same value [then put them on O EN in order of the size of sub rap read from the nearest sub rap to the front - the size of the sub rap is calculated on the pruned rap [with nodes already on LINEA I ED removed -

b-If there are two or more candidate nodes with the same value [and same size of sub rap [then put them on O EN in order of the value of the nearest node from which the nodes depart lowest value to the front -

c-If there are two or more candidate nodes with the same value [size of sub rap [and value of node] then put them on O EN in order of the distance of the node from the start node furthest from the start node to the front -

d-If there are two or more nodes at the same distance from the start node [and one or more already on O EN [then put any new nodes in front of the ones already on O EN -

e-If there are still two or more candidate nodes [then put them on O EN in some order determined by information contained in the network such as the time the node was created -

-If O EN is empty and not all nodes have been removed from the rap [then reverse all the remaining nodes in the rap -Go to 1.5 -

-If O EN is empty then stop -

-Move the node at the front of O EN -

-Calculate the node from which the node departs the FOC NODE and the node to which the node points the CCE O NODE -

-If the CCE O NODE is already on LINEA I ED then stop -

-Add the CCE O NODE to LINEA I ED in position immediately after the FOC NODE -

-Find all untraversed nodes from the CCE O NODE -

-Go to -

Figure 4. The Best First algorithm

The heuristic is designed to favour the choice of priority nodes which lead to or from farthest and therefore likely to be important sub parts of the network - Heuristic a requires the size of the sub rap from a node to be computed [but the computation can be bounded without significantly affecting the operation of the algorithm - Line allows for networks where some nodes cannot be reached due to the direction of the nodes - It is needed because occasionally a subset connected a cluster of nodes to the main network with a node in the reverse direction - Add all reachable nodes to the near set and then reverse all the remaining nodes as the effect of including the remaining nodes in the near set [but at low priority -

The best first algorithm overcomes the particular problems of the greedy algorithm [producing a near set of a d b c e f for the network in Figure 1 [and a d b c e f for the

network in Figure 4. It also has the advantage of forming the LINEA I ED string in order of priority so that by varying a cut off value for the priority it can filter out parts of the hypertext network retaining only those nodes on a path.

3.4 The algorithms in operation

To give an example of the algorithm's operation Figure 5 shows a string of nodes network produced by a writer on the topic of 'Can computers think?' or 'Artificially intelligent machines?' near the start of the order shown in Figure 4.

```

a
a h
a h g
a h g i
a h g i j
a h g i j k
a h g i j k f
a h g i j k f b
a h g i j k f b c
a h g i j k f b c d
a h g i j k f b c d e

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Figure 5. Order of nodes produced by the hillclimbing algorithm for the network in Figure 4

The near sed text correspond to the following order of nodes as follows

Can computers think

Computers may be able to think in non human ways.

Some people suggest that it is nonsense to talk of machines thinking.

filling the. h s r m C. S 4 T TL Tw So c . . Tc Tw f fi TL Tc , TL Tc ,

A graph creates the nearest shortest order sequence and the final order of nodes produces the nearest text below

Can computers think

Computers may be able to think in non human ways.

Some people suggest computers can think like humans.

Turing suggested an operational definition of thinking.

The Turing Test is a way of discovering whether machines can reason, by means of a conversational game.

This is reminiscent of behaviourist psychology.

Some people suggest that it is nonsense to talk of machines thinking.

Searle argues that machines do not have intentionality.

Machines have syntax but no semantics.

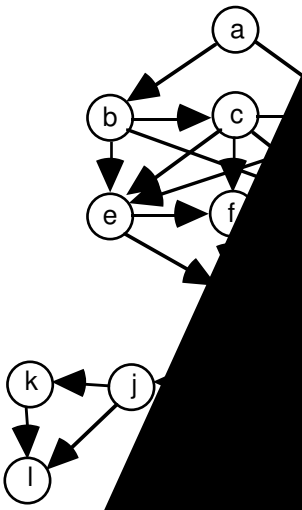
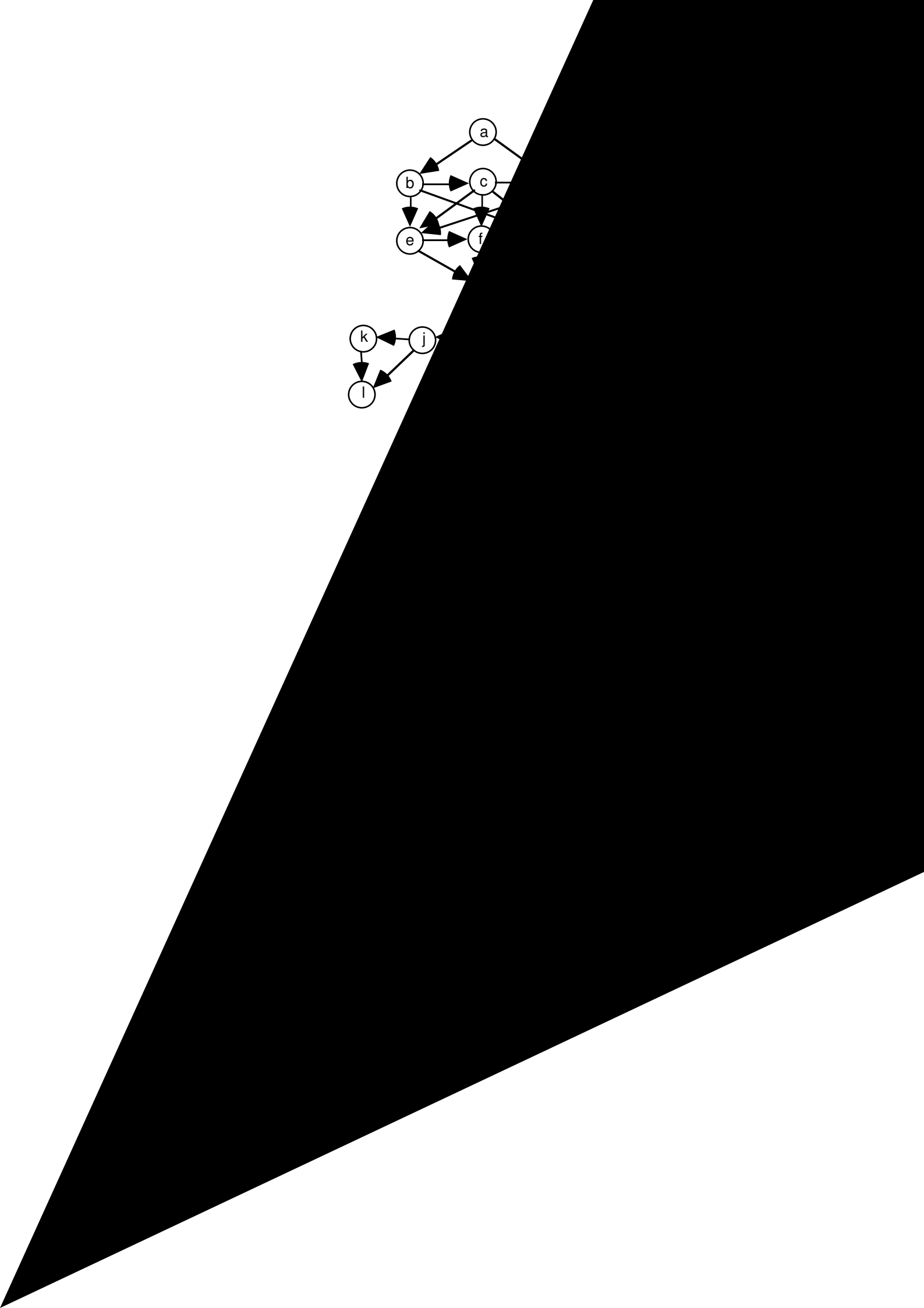
Machine thought is impossible in principle.

Turing suggested that machines would soon be able to compete with humans at intellectual reasoning.

The order of a graph provides a plausible framework for a nearest text which a writer could then flesh out with connecting phrases to create a first draft

Can computers think?

Computers may be able to think in non human ways **but** some people suggest computers can't think. Turing suggested an operational definition of thinking. Turing's test is a way of discovering whether machines can reason by means of a conversational game. This is reminiscent of behaviourist psychology. **However**, some people suggest that it is nonsense to talk of machines thinking. Searle argues that machines do not have intentionality. Machines have syntax but no semantics.



There is no generally agreed set of basic types and hypertext systems will provide pre-specified ranges from IBI with the types intended for development. Harasz and Conrath to the EX/NE/ER <...> series with over a hundred different types. The set of types chosen for the experiment was intended to be sufficient to be handled by the experimental subjects but large enough to cover the main types of conceptual relations for the text types used in the experiment. The subjects were instructed to say during the experiment if they required any further types and at the end of each subject was asked to suggest further types which they might have found useful. No subject indicated that an additional type would be necessary to complete the task. The subjects suggested types which they might have found useful. Each of the three people suggested two or more types. These were /underlines/ /context/ /co-descriptions/ but /new episodes/ /facilities/ /subjects/.

3.1.4 Design

A repeated measures design was used with each subject producing a hypertext for each of the four texts. The order of texts was counter-balanced.

3.1.5 Procedure

Each subject was shown a list of types and the experimenter explained the meaning of each of the types. The subject was also shown an example hypertext. First, the subject was then given the set of cards containing the text chunks for the first example text as well as the text itself. The text was available for reference during the experiment. The subject was asked to stick the cards onto a whiteboard and to use a board marker to draw in relations. Each type would have an arrow indicating direction and a label chosen from the set of available types. The subject was encouraged to use whichever strategy seemed natural to construct the hypertext. The subjects placed all the cards on the board and then drew in the types. Others added types after placing each card. Subjects were allowed as much time as they wished to carry out the task. The experimenter recorded the layout of the hypertext on paper as the subject created it. When the subject was satisfied that the hypertext was complete the experimenter removed it from the board and gave the subject the next set of cards. The experiment ended when the subject had created four hypertexts.

	Hillclimbing Labelled					Best First Labelled					Hillclimbing Unlabelled					Best First Unlabelled				
	seq	cmp	cau	des	mn	seq	cmp	cau	des	mn	seq	cmp	cau	des	mn	seq	cmp	cau	des	mn
S				4		4										4		4		4
S	4						4		4				4			4				
S																				
S						4										4			4	
S	4		4			4	4	4								4				
S	4		4			4	4		4							4				
S									4											4
S	4		4																	
S						4										4				
S							4													
S						4														
S																				
MN																				

Table 2. The evaluator's scores for the linearised text produced by the algorithms.

L f_d, f_p, f_r surpr f_c vec extnatuns of t | stys e f_n desc pt ve^l styp a newspaper repor

Table 2 shows the mean scores for the usual evaluator and the scores produced by the least cost algorithm. The error produced scores in the range 0 to 1 (these have been normalized to ease comparison in the table). There was a significant correlation between the scores r_s

The error texts were rated by the evaluator. The best score for any of the near sat on algorithms was 0.9 for the best first algorithm applied to the labeled Description yper-text. The randomly ordered texts were also rated. The lowest score for any of the algorithms was 0.2 for the best algorithm applied to the unlabeled sequence yper-text.

The mean score for the best first algorithm applied to the yper-texts with labeled nodes [and the corresponding mean score for the best algorithm] is 0.9. A t-test $N=10$ [adjusted for ties] shows the difference between the algorithms to be significant at $p=0.01$ one-tailed.

The scores for best first algorithm applied to the yper-texts with the labeled nodes and to the yper-texts with the unlabeled nodes are almost identical.

4.3 Discussion

The good correlation between the ratings of the usual evaluator and the least cost restoration scores indicates that computerized restoration of the structure of a near-sent text from error texts is a useful means of measuring the effectiveness of near-sent and could provide a comparative test of new near-sent algorithms. Any new algorithm could be applied to the yper-texts used in this experiment and the near-sent could be readily compared with those of the two algorithms tested here.

As expected [the best first algorithm] was significantly more effective at near-sent the yper-texts than the best algorithm [and the mean score of 0.9 above the slowest or an sent] slowest at disordering = even suggests that it produced texts with sufficient or an sent on to be useful as first drafts. The low score of 0.2 for the Causation yper-text suggests that automatic near-sent may be less useful for yper-texts which contain a number of distinct but related topics. One approach may be to group the nodes into topics and then apply a version of the algorithm which keeps to the text on the same topic.

The unexpected finding was that the relationship between the number of nodes and the effectiveness of the near-sent. There are a number of possible explanations for this result. The subjects may not have had enough practice in creating yper-texts to be able to put appropriate labels to the nodes or the range of node types may not be adequate to indicate the conceptual nodes in the text or the algorithm may not have good use of the information due to poor heuristic or the properties assigned to the node types may be inappropriate. It would be possible to investigate the latter explanation by varying the results produced by the computer evaluation of the near-sent texts as input to a learning algorithm which determines the optimal property for each node type.

In this experiment the subjects created yper-texts from published near-texts by applying the results of reading comprehension to the expected referential nodes embedded in the text. The experiment measured how effective the near-sent algorithms are in selecting nodes and traversing them in an appropriate order. But creating a yper-text as part of writing is not quite the same activity. A writer in producing a notes network follows a trail of internal associations with no textual cues for guidance. A yper-text produced during the writing process may have nodes and categories deep conceptual relations rather than surface textual ones and may thus be more difficult for an algorithm to near-sent. The second experiment tests this possibility.

4.4 Experiment 2

4.4.1 Rationale

The aim of the experiment is to test the ability of subjects on yperertext notes networks generated as part of a writing activity. It differs from experiment one in that the subjects are generating their own yperertexts on a given topic. The assumption is that the yperertext acts as a means of externalising and controlling the writer to represent a pattern of mental associations between topics. The two near-saturation articles were compared against an near-saturation carried out by the authors of the yperertexts and a random ordering of nodes. The near-saturations were scored by two blind assessors on a five-point scale for textual organisation.

Hypothesis 1: The near-saturated texts will have a higher rating than the best method of automatic near-saturation.

Hypothesis 2: The automatic near-saturated texts will have a higher rating than the random orderings.

Hypothesis 3: The best first article with labeled nodes will produce a higher rating than the unlabeled first article with labeled nodes.

Hypothesis 4: The best first article with labeled nodes will produce a higher rating than the best first article with no node information.

4.4.2 Subjects

The subjects were the same as for experiment one.

4.4.3 Materials

The subjects were given the same card of node types as for experiment one. Each subject was given a stock of blank index file cards on which to write the text. The materials for creating the yperertexts were as for experiment one.

4.4.4 Design

Each subject produced one yperertext.

4.4.5 Procedure

The subjects were given a list of three topics and asked to choose one topic on which they would create a yperertext. The topics were 'How to choose a suitable holiday', 'Would I sell my car and cycle to work?' and 'The role of Britain in Europe'. Four of the subjects chose the 'Holiday' topic, seven chose the 'Bicycle' topic, and one chose the 'Europe' topic.

Each subject was given a stock of twenty blank index cards and was asked to generate short sentences on the topic, writing each sentence on a separate card. The subject was asked to

by the subjects and near sat on were randomly ordered and given to the two independent evaluators. The evaluators marked the texts using the self-evaluative point scale as in the previous experiment. For each of the 12 texts the markers differed by two scale points and for two texts the markers differed by three scale points. The evaluators were asked to re-mark these texts without reference to the previous scores. The re-marking produced six cases where the evaluators differed by more than one scale point.

4.5 Results of Experiment 2

Table 4 shows the scores after re-marking of the two evaluators for each of the 12 texts produced by the near-synonym or text random order and and near-synonym.

	Hillclimbing Labelled		Best First Labelled		Hillclimbing Unlabelled		Best First Unlabelled		Random		Hand Linearised	
	Eval	Eval	Eval	Eval	Eval	Eval	Eval	Eval	Eval	Eval	Eval	Eval
S	4		4		5	4	5	4	4		5	4
S	4	4	4		5						4	4
S		4	4			4					5	5
S4	4	4		4	4	4		4			5	4
S ₅	4				4						4	
S	4		5		4		4	4			4	4
S			4				4	5			4	4
S				5							5	5
S	4	4		4		4		4			5	5
S		4	5	4	4	4	4	5			5	5
S						4	4	4			4	5
Mean			5		4		5		5		4.5	
Mean			4				4		4		4.4	

Table 4. Scores of the two evaluators for the linearised texts.

The correlation between the scores of the two evaluators is significant at $p < 0.05$. The mean scores of the two markers for each of the text types (see Figure 5) are: labelled, 5.5; best first labelled, 5.5; hill-climbing unlabelled, 5.5; best first unlabelled, 5.5; random, 5.5; and near-synonym, 5.5. As the automatic near-synonym texts have lower ratings than the texts near-synonym and are better than the random order texts, the differences between the scores are significant in both cases at $p < 0.05$. Coxon (1991) [one-tailed] the two scores produced by the best first marker are better than those for the hill-climbing one but the differences are not significant. As for experiment one, there is no significant difference between the scores for the best first marker with labelled information and without it.

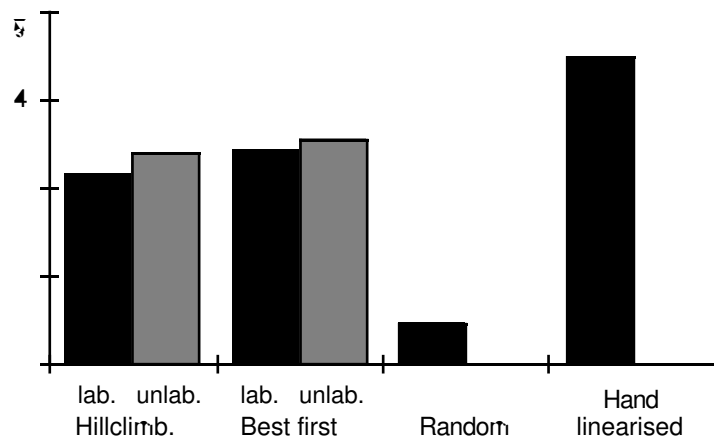


Figure 8. Mean scores for the linearised texts

4.6 Discussion

Both algorithms produce near-solutions in the range between /software at or an sed software at d sor an sed=and /acceptably or an sed=It indicates that automatic near-solution could provide a useful bridge between an ideas or an user and a text editor as part of a writing environment. The near-text files we need to be edited, but the algorithms [t

reverse direction for example the causes would be given a priority for its reverse direction corresponding to the effects caused by relations was the favoured method but it was not preferred due to a lack of evidence to indicate the efficacy of networks

6. Conclusions

We have described a robust generation method for notes networks near saturation which has been presented as part of a writing environment which combines an ideas or anser with a documented theory the experiments suggest that the best first near saturation method is acceptable for creating a first draft of a near text from a notes network but that further work is needed to improve use of the network information to determine saturation and to evaluate the use of the method as part of a writing environment

References

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