

# **A Comparison of Algorithms for Hypertext Notes Network Linearisation**

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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.***

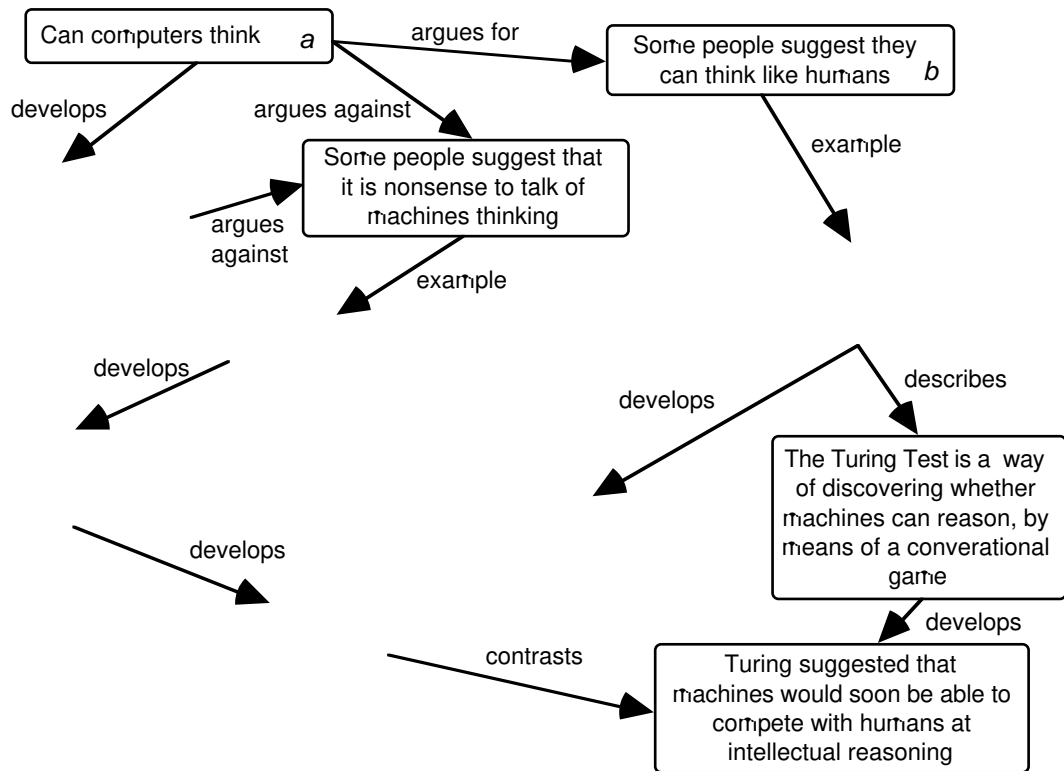
A linearisation author differs from a search author in that the criterion for success is not to reach a specified goal, but to produce an order of nodes which meets the expectations of a reader. A coherent/easy narrative leads the reader onwards by following trails of association that resonate with the 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 node 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 - 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 15 -  
 -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 15 -

**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 breadth first 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 set 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

Figure 5 shows an example of the algorithm's operation. Figure 6 shows a sample network produced by a writer on the topic of 'Can computers think?' and Figure 7 builds up the nearest text in the order shown in Figure 5.

```

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

```

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

The nearest text corresponding to the final order of nodes is 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.hsr mC.S4 T TL Tw So4444\_c . . Tc Tw\_f fiTL Tc, TL Tc,



A sort of creates the nearest student order s own n F ure [ and the final order of nodes produces the near 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.

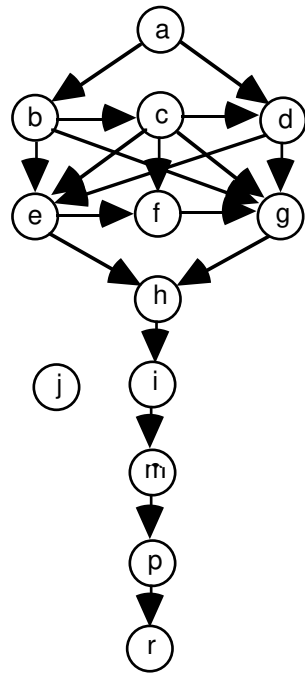
Machine thought is impossible in principle.

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

The order of a sort of provides a more plausible framework for a near text [ with the writer could then finesse 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. However, 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, and description, but new episode, facets, and 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 counterbalanced.

### 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	5 4 5	4 5 5 5 5 5	4 5 4 5 4 5	4 5 4 5 4 5

Table 2 shows the mean scores for the unaided evaluator and the scores produced by the least cost algorithm. The error rates for the least cost algorithm were not significantly different from those of the unaided evaluator. There was a significant correlation between the scores of the unaided evaluator and the least cost algorithm.

The error rates were rated by the evaluator. The best score for any of the near-sat-on algorithms was for the best first algorithm applied to the labeled description hypertext. The randomly ordered texts were also rated. The lowest score for any of the algorithms was for the best algorithm applied to the unlabeled sequence hypertext.

The mean score for the best first algorithm applied to the hypertexts with labeled nodes [and the corresponding mean score for the best algorithm] was significantly higher than the mean score for the unlabeled algorithm. A t-test indicated that the difference between the algorithms was significant at the 0.05 level.

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

### 4.3 Discussion

The good correlation between the ratings of the unaided evaluator and the least cost restoration scores indicates that the restoration distance of a near-sat-on text from the original text is a useful means of measuring the effectiveness of near-sat-on and could provide a comparative test of new near-sat-on algorithms. Any new algorithm could be applied to the hypertexts used in this experiment and the near-sat-ons could be readily compared with those of the two algorithms tested here.

As expected, the best first algorithm was significantly more effective at near-sat-on the hypertexts than the unlabeled algorithm, and the mean score of 0.57 above the slowest or an sat-on to be useful as first drafts. The low score of 0.22 for the Causation hypertext suggests that automatic near-sat-on may be less useful for hypertexts 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 reduction in error rates about node types did not alter the effectiveness of the near-sat-on. There are a number of possible explanations for this result. The subjects may not have had enough practice in creating hypertexts to be able to put appropriate labels to the nodes or the error rates of node types may not be adequate to indicate the conceptual nodes in the text or the algorithm may not have good use of node 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-sat-on texts as input to a learning algorithm which determines the optimal property for each node type.

In this experiment the subjects created hypertexts 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-sat-on algorithms are in selecting nodes and traversing them in an appropriate order. But creating a hypertext as part of writing is not quite the same activity. A writer in producing a notes network follows a trail of mental associations with no textual cues for guidance. A hypertext 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-sat-on. 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 hypertext notes networks generated as part of a writing activity. It differs from experiment one in that the subjects are generating their own hypertexts on a given topic. The assumption is that the hypertext acts as a means of externalising and controlling the writer to represent a pattern of mental associations between topics. The two nearest adjacent topics were compared against an adjacent pair carried out by the authors of the hypertexts and a random ordering of nodes. The nearest adjacencies were scored by two blind assessors on a five point scale for textual or adjacentness.

**Hypothesis 1:** The nearest adjacent texts will have a higher rating than the best method of automatic nearest adjacency.

**Hypothesis 2:** The automatic nearest adjacent texts will have a higher rating than the random orderings.

**Hypothesis 3:** The best first adjacent words with labeled nouns will produce a higher rating than the best adjacent words with labeled nouns.

**Hypothesis 4:** The best first adjacent words with labeled nouns will produce a higher rating than the best first adjacent words with no noun 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 20 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 hypertexts were as for experiment one.

### 4.4.4 Design

Each subject produced one hypertext.

### 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 hypertext. 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

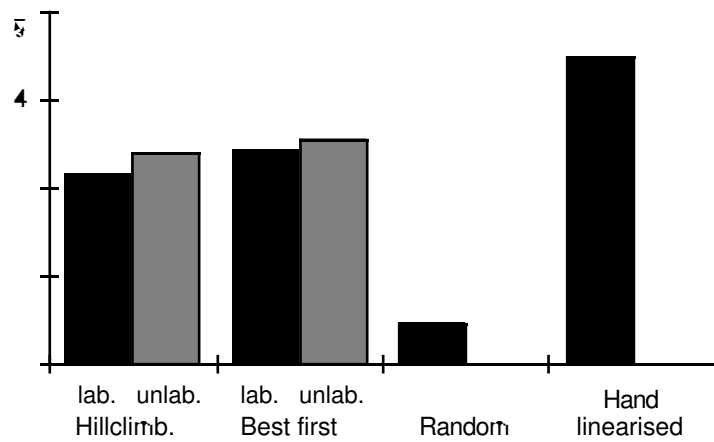
Table 4 shows the scores after re-marking of the two evaluators for each of the 12 texts produced by the near-synonym or text random order and and near sat on.

	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		5				4	4
S		4	4			4					5	5
S4	4	4		4	4	4		4			5	4
S <sub>5</sub>	4				4						4	
S	4		5		4		4	4			4	4
S			4				4	5			4	4
S				5				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, 5.0; best first labelled, 5.0; hill climbing unlabelled, 5.0; best first unlabelled, 5.0; random, 5.0; and near-synonym. As the automatic marking of near-synonym texts gave lower ratings than the texts near-synonym and and are better than the random order scores, the differences between the scores are significant in both cases at  $p < 0.05$  (Mann-Whitney U-test). 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.





**Figure 8. Mean scores for the linearised texts**

#### **4.6 Discussion**

Both algorithms produce near-solutions in the range between /software/ and /online software/ and /acceptable/ and /online/. It indicates that automatic near-solutions could provide a useful bridge between an idea 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 achieve good use of the network information to determine saturation and to evaluate the use of the method as part of a writing environment

## References

- Boden M. *The Creative Mind: Myths and Mechanisms* London: Heinemann and Nicholson
- Bolter J-D, Joyce M and Fitt J-B. *Storyspace: Hypertext Writing Environment*.