

# Visualizing Spoken Interaction

Jens Allwood  
Christopher Ahlberg

**Abstract:** This paper introduces and explores the use of IVEE, a technique for visualization and dynamic database queries applied to a database of transcriptions from various types of linguistic interaction. The use of IVEE is illustrated by presentation of data on vocabulary richness, verbal dominance, hesitance, uncertainty and liveliness,

## 1. Visualizing and exploring complex datasets with dynamic queries

### 1.1 Introduction - dynamic queries

Dynamic queries is a concept for information exploration and database querying based on visualization. A dynamic query is defined as "a user controlled animated visualization of the results from a database query."

Users manipulate query widgets of various kinds to construct database queries, such as rangesliders, toggles, and alphasliders (Ahlberg, 1994, Figure 1), to incrementally update a visualization of the current query result. The visualization has an update rate of less than 100 ms, allowing for users immediate control of the result set.

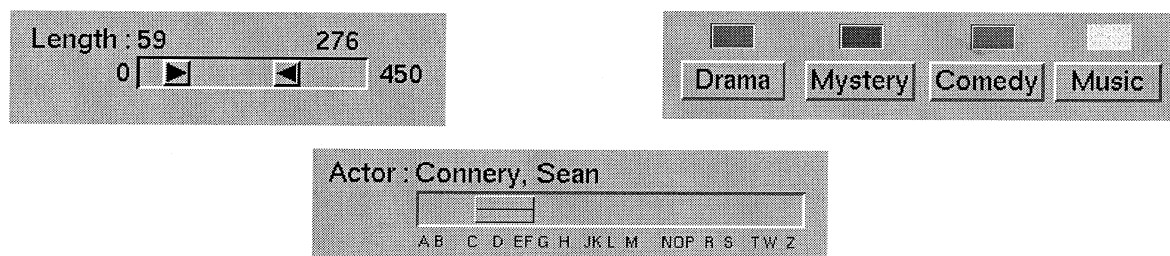


Figure 1. Query widgets in dynamic queries.

The benefit of static visualizations of data sets has long been known, visualizations of for example demographic, scientific, and economic data sets are commonplace today (Tufté, 1983). Static visualizations can be created with tools such as Excel, Lotus 1-2-3, and various statistical packages such as SAS and Statview. User controlled animated visualizations with a screen update rate of less than 100 ms have the additional benefit of allowing for easy exploration of relations between individual query result sets.

Users might increase the value of a query parameter to detect how the increase affects the size of the result set. The relevance of the result set can be judged from for example the color of the elements in a visualization, or from where the result set shows up in the visualization (indicating for example geographic proximity or some other high level semantic property). Users might also manipulate query widgets to explore trends and patterns in data, or to detect anomalies.

Visual representations of databases also allows for presentation of many more database elements in single screens. This in turn, allows for overviews in a way which is not possible with traditional textbased data exploration systems. A visualization might present thousands of elements simultaneously, while a text based system is limited to the magnitude of tens of elements.

The cognitive load while performing information exploration tasks is usually high. Finding patterns, exploring trends, and detecting anomalies is cognitively difficult. Allowing users to incrementally control the animated display of query result sets in visualizations, minimizes the mental effort needed to grasp relations between queries and query results and relations between individual query results. A lot of the benefits from animated visualizations arise from the fact that we can perform many perceptual tasks such as detection of trends, patterns, and anomalies in pictures with very little conscious mental effort. Not until recently has the use of these our remarkable perceptual abilities found its way into information retrieval systems (Ahlberg & Shneiderman, 1994; Robertson, Card & Mackinlay, 1993).

We have built a number of interesting prototype dynamic queries applications for experimental purposes, such as a dynamic periodic table Ahlberg et.al, 1992 and a dynamic homefinder Williamson & Shneiderman, 1992. The prototypes used domain specific visualizations such as the periodic table of elements and geographic maps.

## **1.2 IVEE: An Information Visualization & Exploration Environment**

The Information Visualization and Exploration Environment (IVEE) (Ahlberg & Wistrand, 1995) is an attempt at a system for creating visualization environments for nearly any database. IVEE can automatically create dynamic queries environments holding query widgets and visualizations. IVEE imports a relation with named attributes, given on a straight forward text format. The data in the relation is examined and classified into datatypes (integers, reals, and strings) and size (number of different values held in the attribute). Based on this examination, IVEE selects query widgets for each attribute. Existing widgets are: rangesliders, alphasliders 3, and toggles (see figure above).

## **2.3 Query widgets in IVEE**

Rangesliders are useful for selecting a ranges criteria for integer attributes, or other attributes with an order relation, and where it is meaningful to specify ranges. Increasing or decreasing the range of an attribute allows for powerful exploration of trends and anomalies.

The alphaslider is a widget for selecting items from long lists of for example strings. The rationale of the alphaslider is its small size, i.e., it allows for selection from lists

of thousands of elements in a small screen area. The alphaslider can not only be used for selection of specific strings, it is also very useful for browsing categorical variables while observing the query result set in a visualization. Finally IVEE utilizes so called toggles when only a few alternatives exist for an attribute and these alternatives should be presented on the screen explicitly.

## 2.4 Visualizations in IVEE

IVEE offers a number of visualizations. The basic visualization is an interactive scatterplot, called a starfield - see Figure 1 for a starfield visualization. Two ordinal variables from a database relation are chosen as the axes in a scatterplot. Users can perform a number of operations on the starfield: zooming, panning, filtering, and selection of details-on-demand. IVEE allows for other visualizations as well, such as geographic maps and other spatial visualizations, and cluster visualizations.

The starfield is simple, yet powerful as it allows for display of many database elements simultaneously. Each element is represented by a small square, which can be coded by color, brightness, shape, size, etc. The spatial location of an element also effectively encodes two properties of an element. The starfield can be used for quick judgements of result relevance, exploration of trends, detection of anomalies, etc.

An obvious advantage of the starfield compared to many other databases visualizations is that it is very easy to create and works for many databases. The only criteria is that two ordinal variables that are meaningful to visualize should exist in the database to be queried.

## 3. Data and Measures

The interactive visualization methods described above have been tried on a database of spoken interaction consisting of 171 transcriptions involving 120 speakers (59 men and 61 women). Each activity has been subsectioned into between 1 and 26 subsections. There are a total of 302 subsections. The transcriptions contain 26,778 utterances and 216,240 words.

The transcriptions are transcribed in "Modified standard orthography" cf Allwood & Hagman 1994 - a format of transcription for Swedish which is identical to standard orthography except that words which have several pronunciations in spoken language are given separate representations, eg written language *jag* (I) can be written *jag* or *ja* depending on pronunciation.

The transcriptions otherwise indicate speakers, utterances, word and subsections due to change of topic or activity. Over and above standard orthography the transcriptions contain indications of pauses, overlaps between speakers, stressed words or segments of words, inaudible stretches of speech as well as comments on gestures or non-linguistic activity when this is necessary for comprehension of the interaction.

On the basis of the data directly available, a number of measures derived from this data can be defined cf Allwood & Hagman, 1994. Below we will exemplify how dynamic queries provide a new way of exploring the data with these measures.

The measures we will consider are:

1) Four measures of **vocabulary richness** (Van Hout, cf. Broeder, Extra and Van Hout, 1986)

- **Type - token ratio** = types/tokens
- **Herdan** =  $\log(\text{types})/\log(\text{tokens})$
- **Guiraud** =  $\text{types}/\sqrt{\text{tokens}}$
- **Theoretical Vocabulary** = The expected number of types for a transcription reduced to a given number of tokens

2) We will also consider the following complex measures defined in Allwood & Hagman 1994.

- **Verbal dominance** = % of tokens/expected % tokens

This measure takes the actual % produced words for a certain speaker and divides it by the % words speakers would have if all speakers had the same share. For 5 speakers this would mean that each speaker spoke 20% of the total produced words. This makes the measure comparable across transcriptions with different numbers of speakers.

- **Hesitance** = (pauses/tokens) + % OCM

This measure correlates hesitance with share of pausing and share of OCM. OCM abbreviates Own Communication Management and includes hesitation morphemes and other devices for change and management of one's own speech

- **Uncertainty** = % OCM + % epistemic tokens

Uncertainty is by this measure related to share of OCM and share of epistemic tokens (primarily verbs and adverbs with epistemic attitudinal content, eg think, believe, perhaps, maybe, probably etc).

- **Liveliness.** =  $2 \times \% \text{ overlapped words} +$   
 $\% \text{ overlaps utterances} +$   
 $2 \times \% \text{ stressed tokens}$   
 $2 \times (\text{pauses/tokens}) - \text{MLU (mean length per utterance)}$

This measure views overlaps and stress as factors adding to liveliness while pauses and long MLU are seen as diminishing liveliness.

#### 4. Some examples of spoken interaction visualized

##### 4.1 Vocabulary richness

A desideratum among researchers who develop measures of vocabulary richness is that the measure should be independent of text length and sampling contingencies. The measure should give a fairly stable result independently of number of tokens.

The IVEE visualizer now gives us a nice and direct way of exploring the four measures defined above, with regard to this desideratum. Compare the four figures below for the four measures.

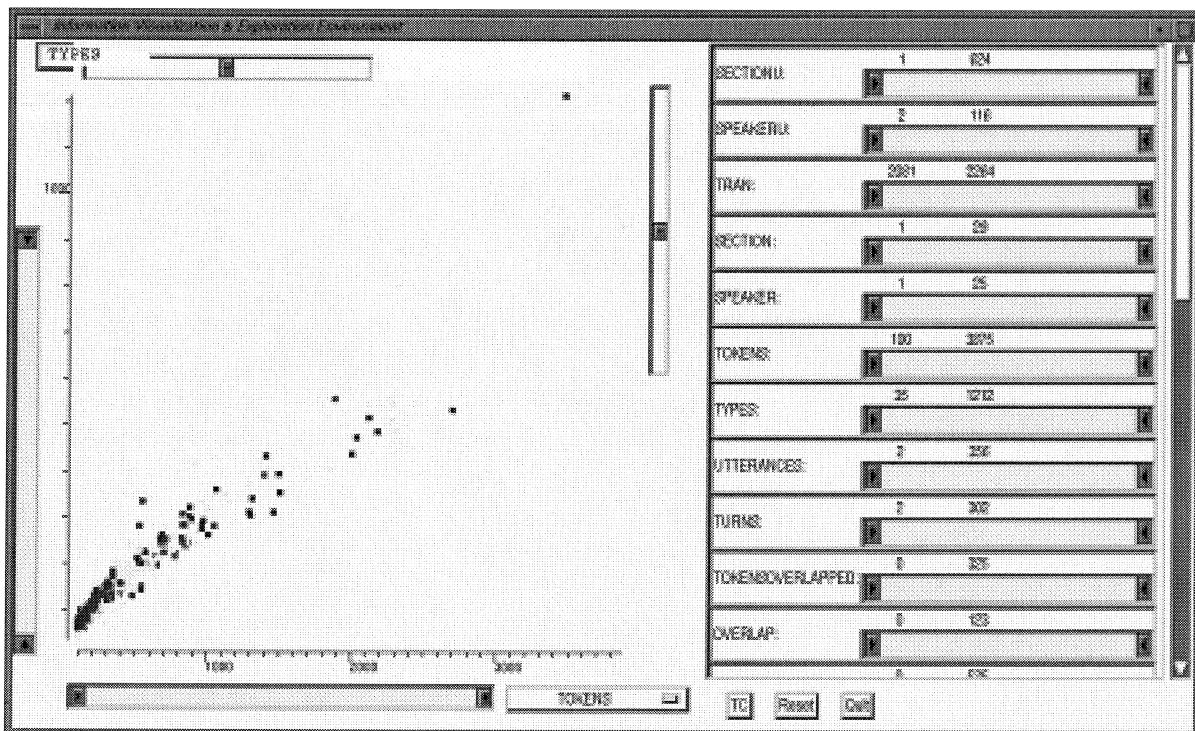


Figure 2. The type token ratio

As we can see, the measure is not linear which would mean that it was independent of number of tokens. Rather it tends to give lower values for higher number of tokens.

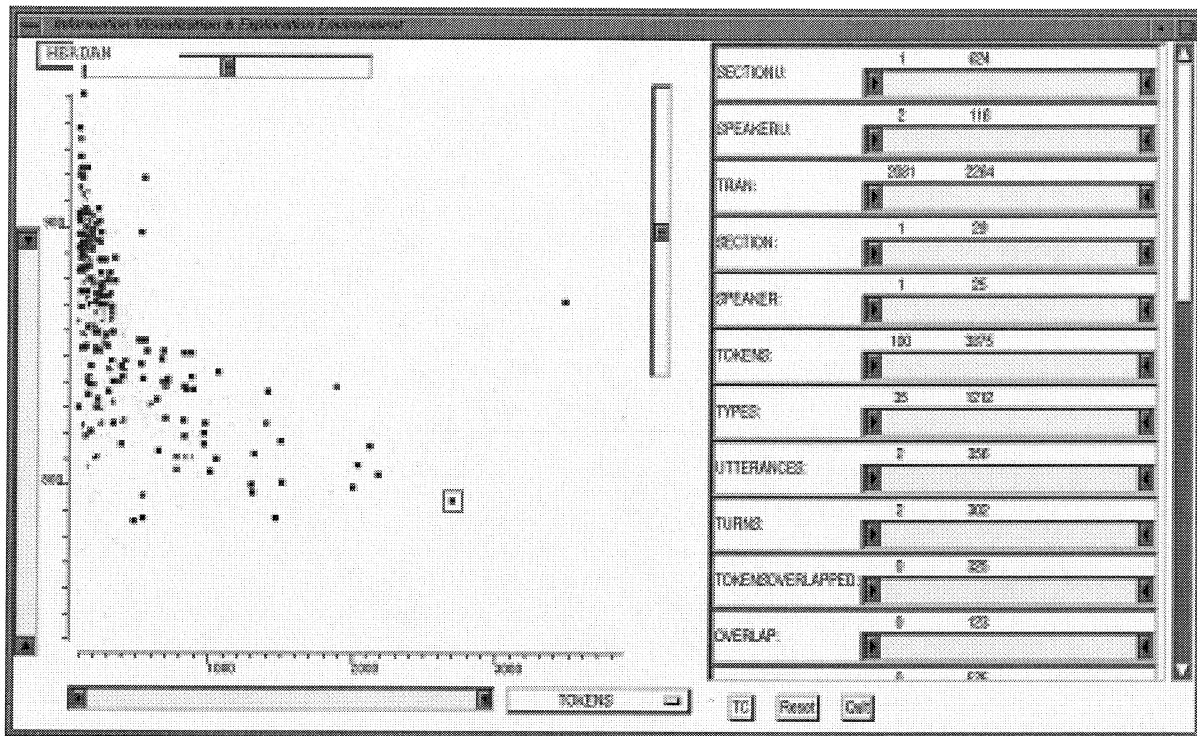


Figure 3. Herdan

The visualization shows that Herdan's measure of vocabulary richness has the same problem as the type/token ratio. A large number of tokens tends to give a low value.

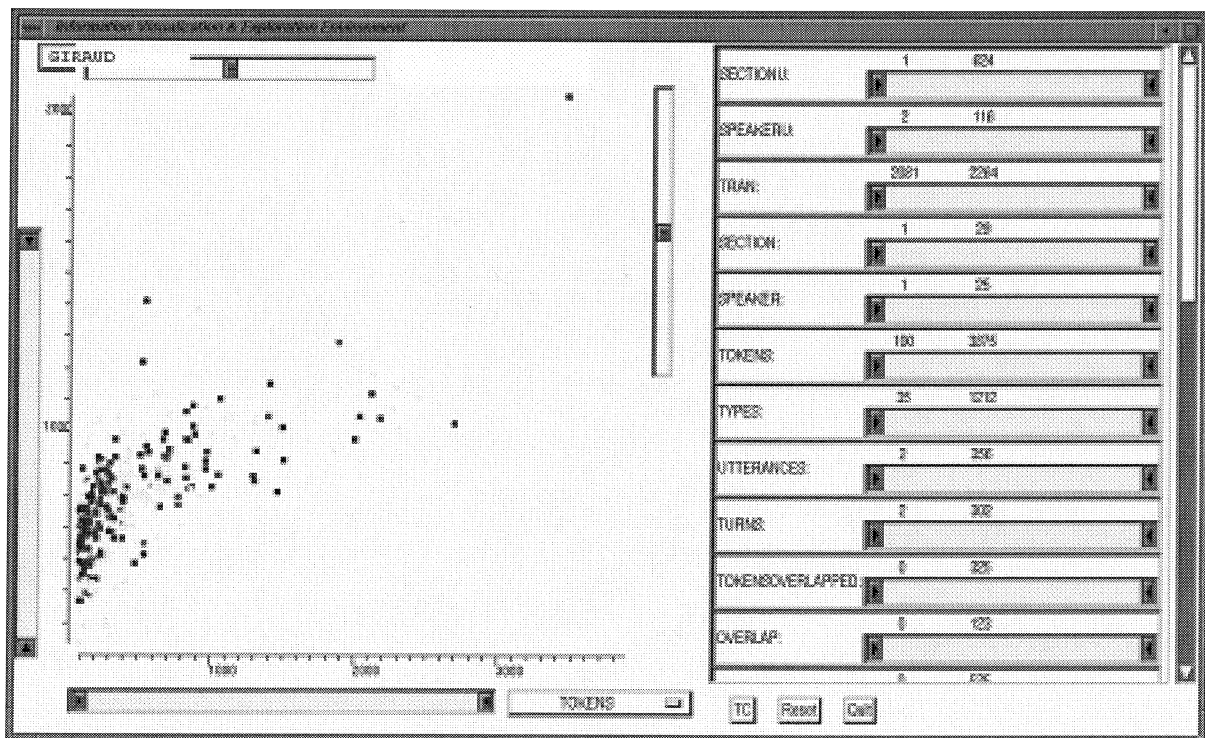


Figure 4. Guiraud

For Guirand the visualization shows that the problem is the opposite. A high number of tokens tends to give a high vocabulary richness score.

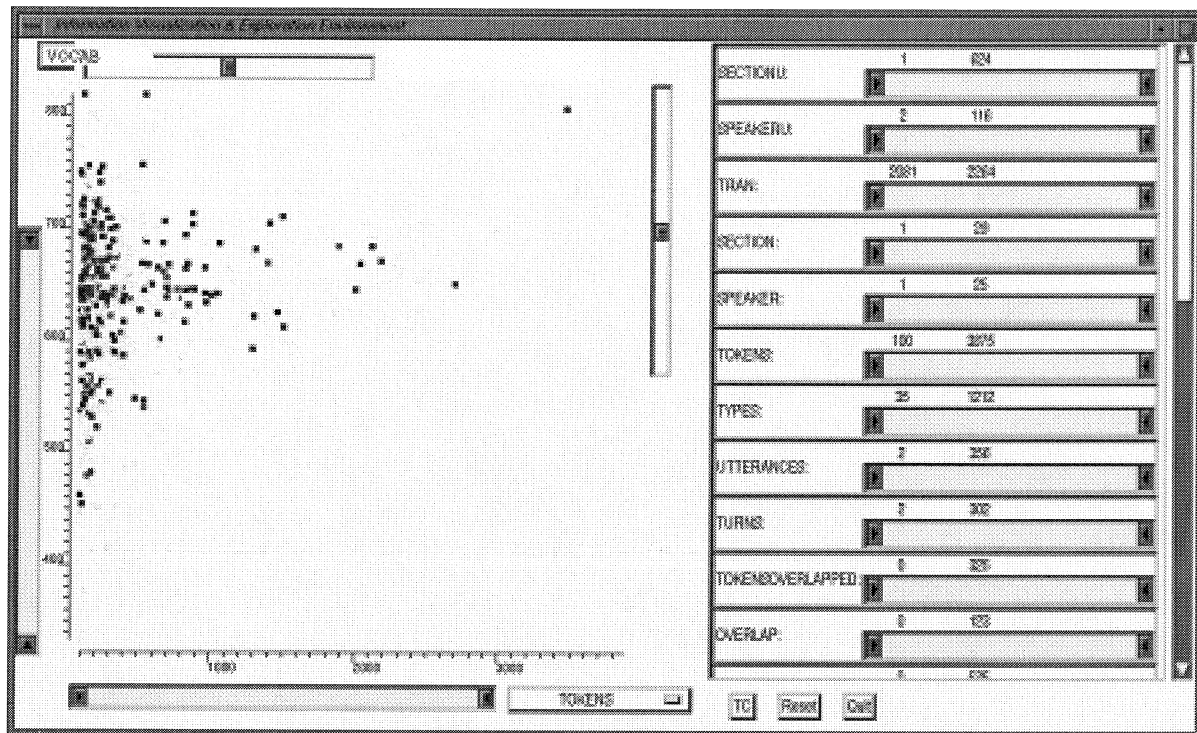


Figure 5. Theoretical Vocabulary (Van Hout)

The theoretical vocabulary measure gives the desired stability. High numbers of tokens do not affect the measure in either an upwards or downward direction.

#### 4.2 Activity dynamics and personal variation

The IVEE alphaslider query device also gives us a way to sequentially inspect the dynamics and personal variability in different activities. In figures 6, 7 and 8 we show how liveliness can vary in three different activities. The dots represent different speakers and the location of the dot represents a liveliness measure for a specific speaker in a specific subsection of an activity.

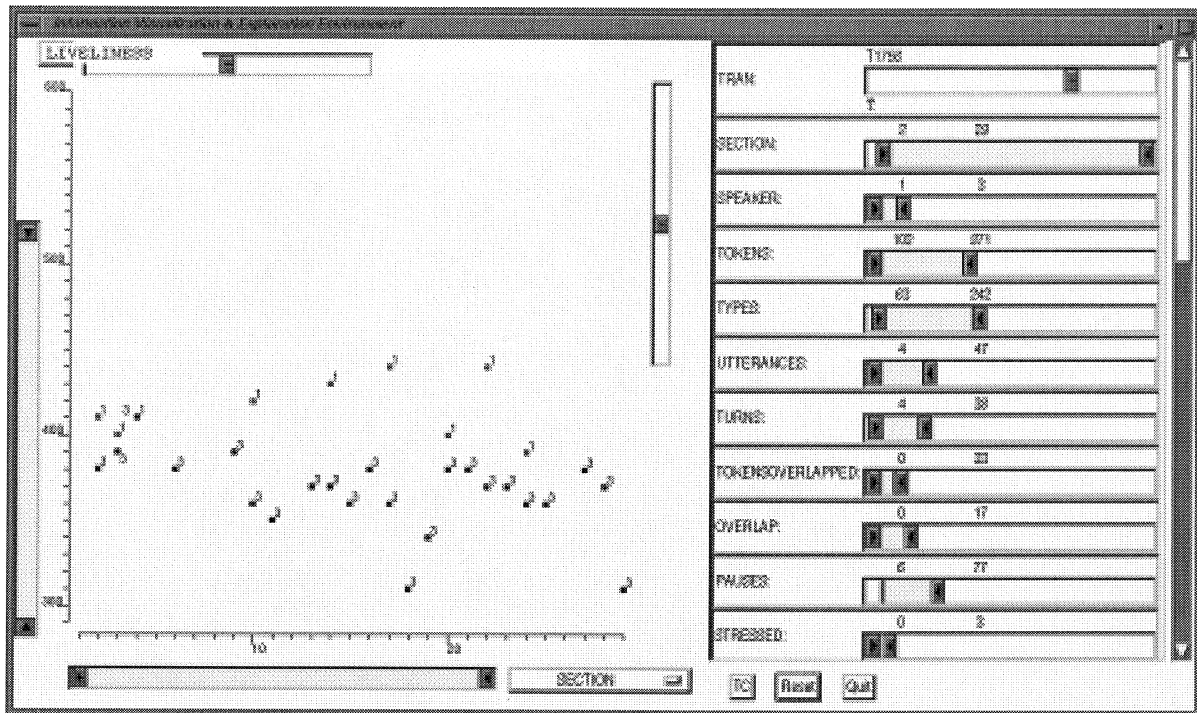


Figure 6. Liveliness variation in an activity with 2 speakers and 28 sections

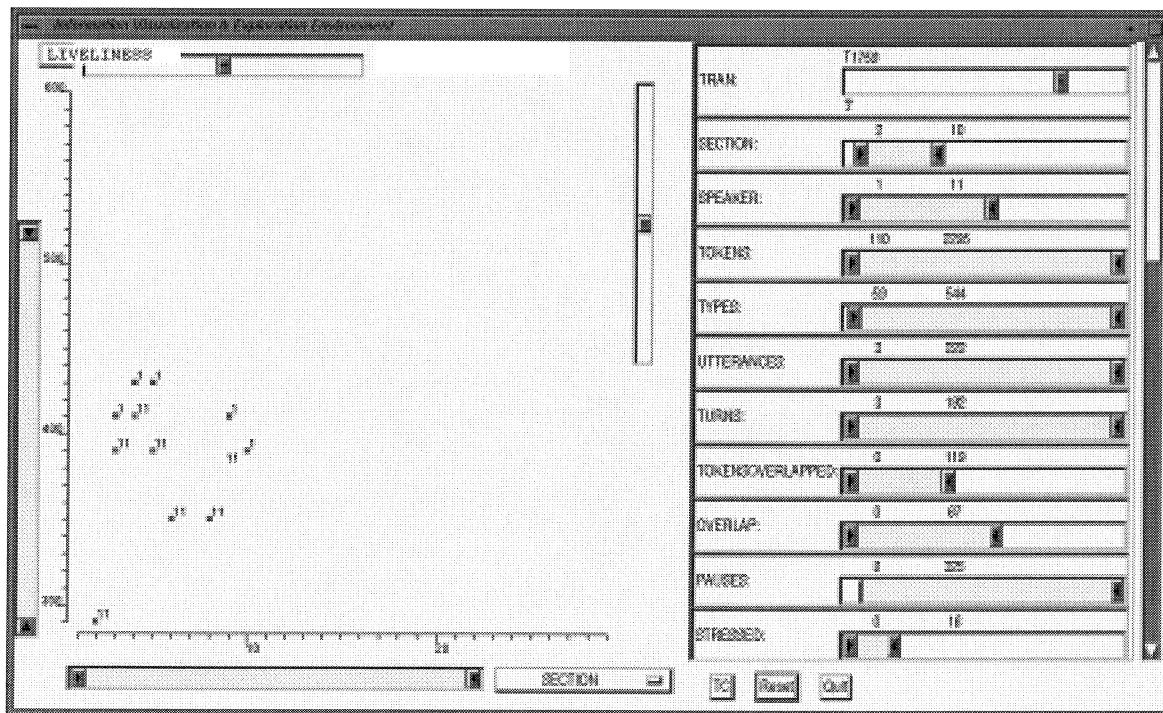


Figure 7. Liveliness variation in an activity with 2 speakers and 9 sections



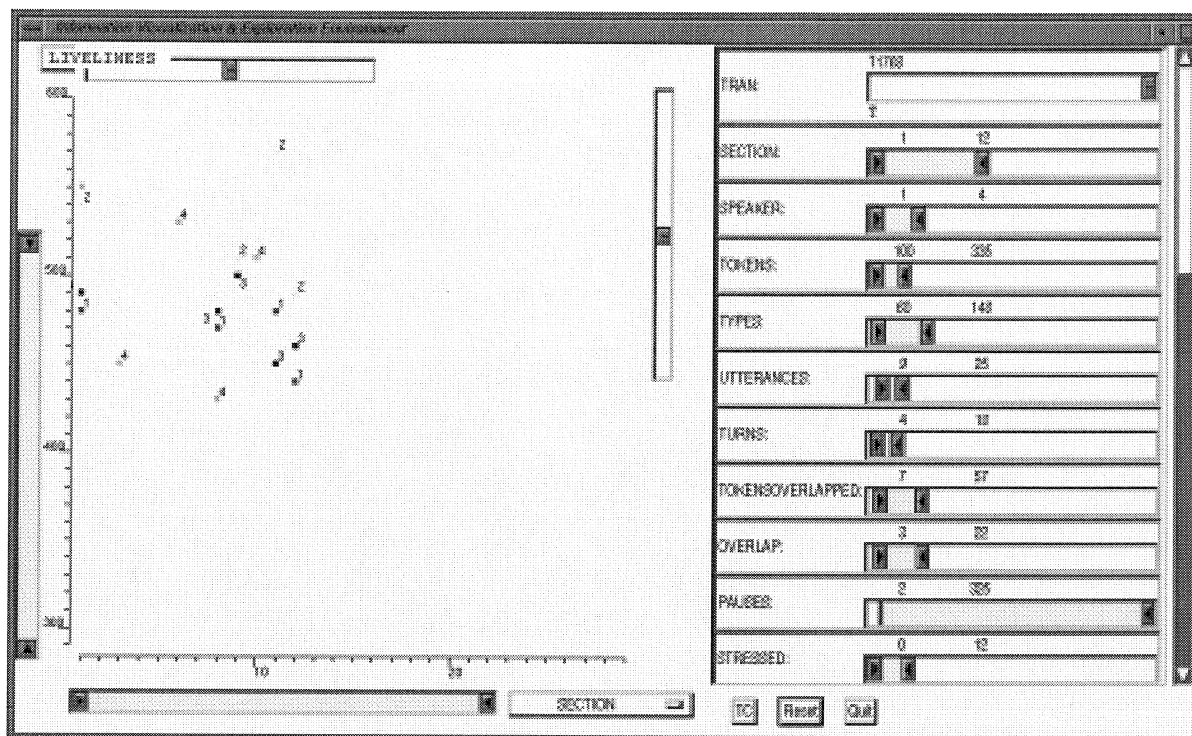


Figure 8. Liveliness variation in an activity with 4 speakers and 12 sections

Using IVEE, all 171 transcriptions in the database can be interactively and rapidly inspected in a similar way, enabling us to pick out transcriptions which show interesting configurations. In the three activities exemplified, we can see how, in figure 6, speaker 1 fairly consistently, in every subsection, is livelier than speaker 3. The personal variability of the two speakers is thus fairly low. This is also true of the activity represented in figure 7 but not of the activity represented in figure 8 where especially speaker 4 shows quite a bit of variation in liveliness depending on which section he is in.

In figures 9, 10 and 11, we show the same activities with regard to hesitance.

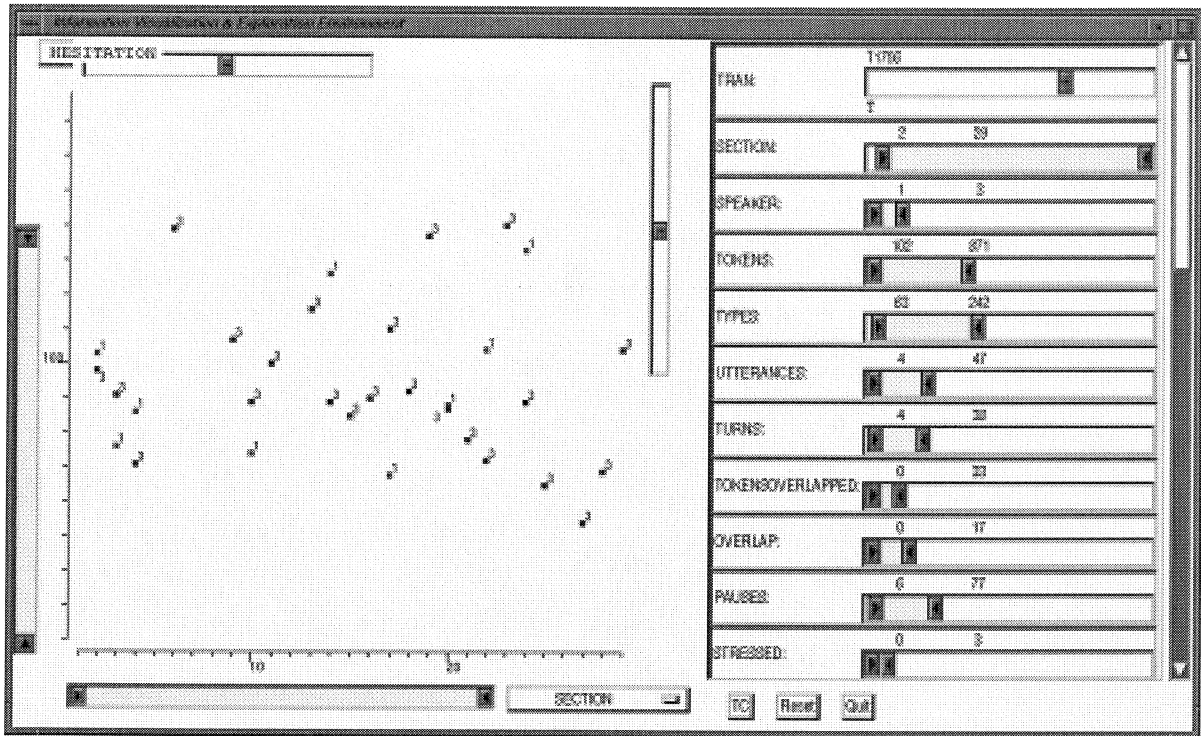


Figure 9. Hesitance in an activity with 2 speakers and 28 subsections

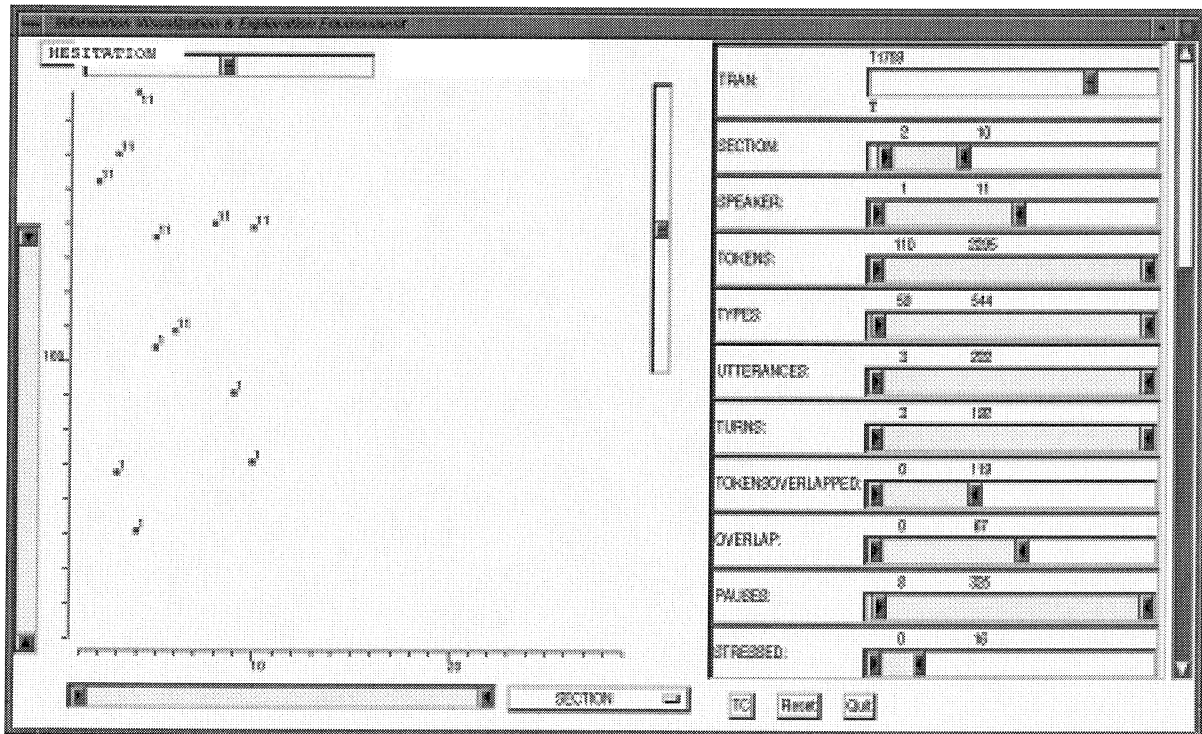


Figure 10. Hesitance in an activity with 2 speakers and 9 subsections

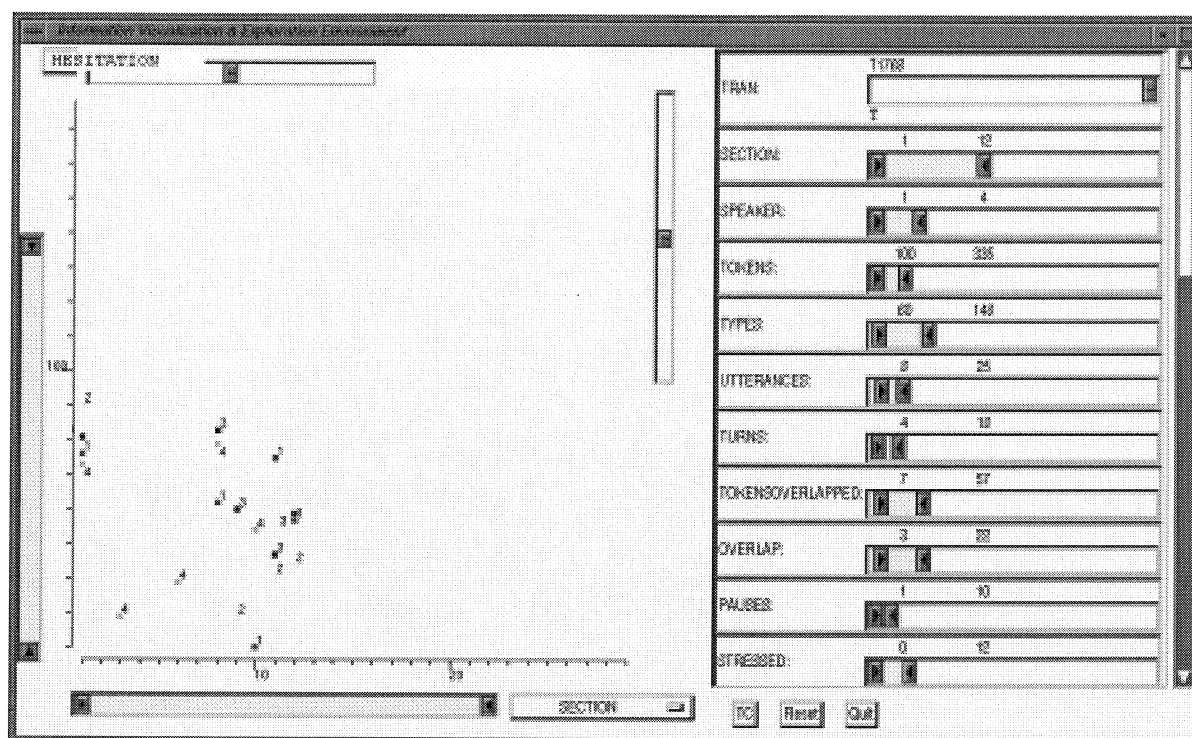


Figure 11. Hesitance in an activity with 4 speakers and 12 subsections

If we compare figure 6 with figure 9, we see that the personal variability of both speaker 1 and speaker 3 is much greater with regard to hesitance than with regard to liveliness. It also looks like values of speaker 1 on an average are slightly lower than those of speaker 3, which is not surprising given that liveliness and hesitance should be somewhat "incompatible". The same inverse relationship can be more clearly seen if we compare figures 7 and 10, where speaker 1 is both consistently livelier and less hesitant than speaker 11. The same tendency can also be seen in a comparison of figures 8 and 11 but since the personal variability of the speakers in this activity seems to be greater, the tendency is weaker.

### 4.3 Complex measures and gender variation

Let us now take a look at how two complex measures, verbal dominance and vocabulary richness, are distributed if we partition the 96 speakers according to gender. The diagrams represent women as black dots and men as grey. Intermediate grey means that the gender of the speaker is unknown. In figure 12 we show the distribution of verbal dominance and in figure 13 the distribution of vocabulary richness as measured by the theoretical vocabulary measure.

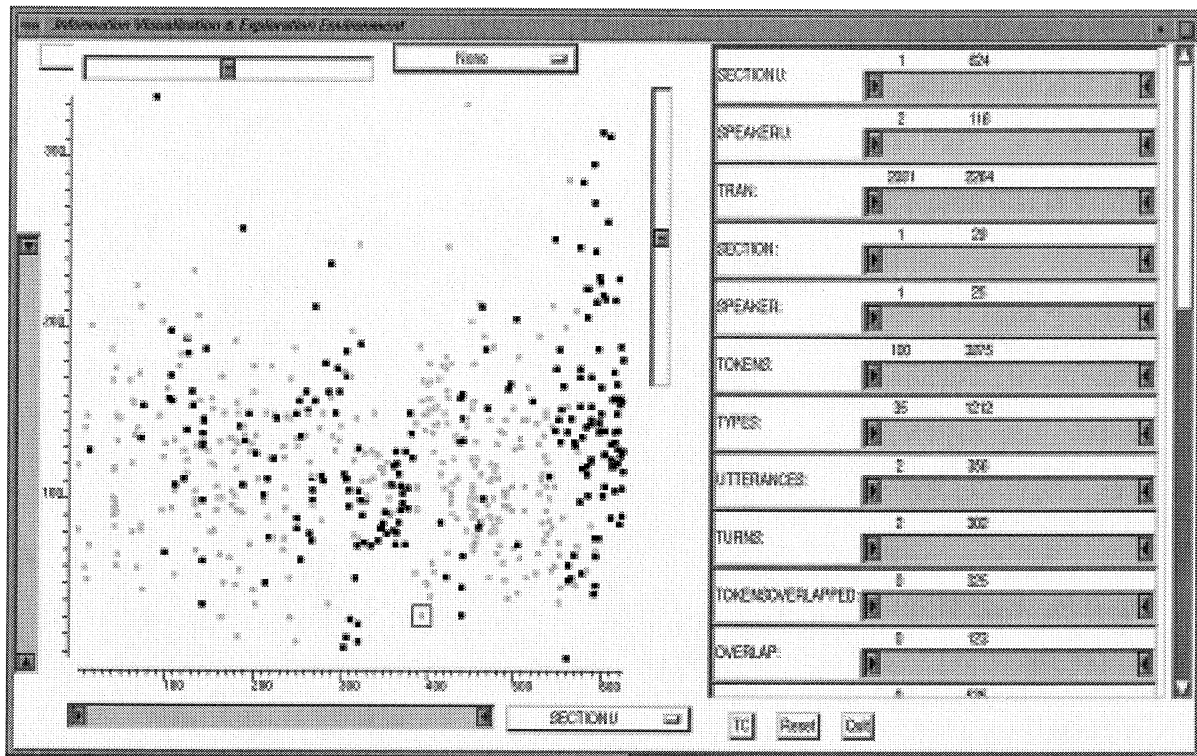


Figure 12. Gender and dominance

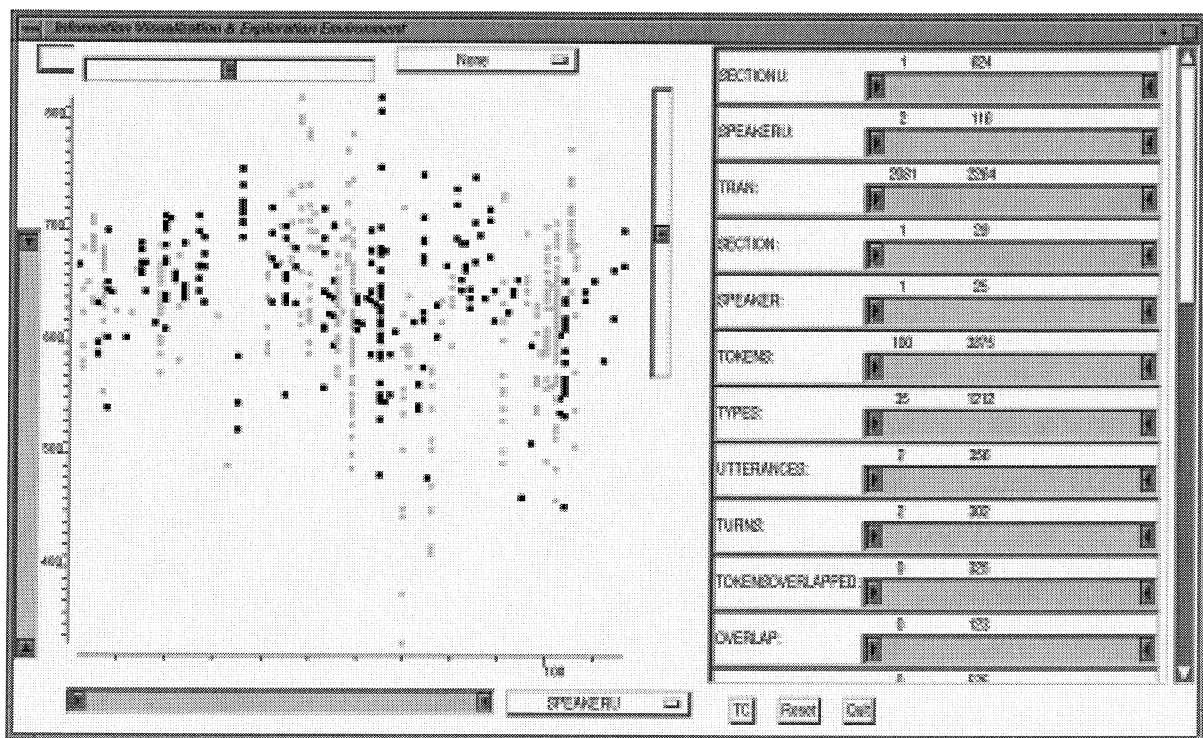


Figure 13. Gender and vocabulary richness

Figure 12 shows that overall men and women are roughly equivalent when it comes to verbal dominance. In this figure, activity variation can be read as a column of

vertical dots. If we look at such volumes we see that there are some activities with only women and some with only men. If we inspect such "single sex" activities, we see that they can show great differences in verbal dominance between different speakers. There is quite a variation in verbal dominance both in "single sex" and "mixed sex" activities. We never seem to occupy equal verbal space. In our data, in fact, the variation in verbal dominance seems to be greatest in women's "single sex" activities, which is perhaps contrary to the impression one gets from works such as Tannen (1991), where women are said to be more focussed on social relations and men on hierarchical relationships.

In figure 13, we can see that also with regard to vocabulary richness, both sexes show a roughly equivalent variation. This is true independently of whether we consider "single sex" or "mixed sex" groups.

## 5. Concluding words

In this paper, we have demonstrated and exemplified how new database search tools involving interactive queries and visualization can be applied to a database consisting of transcriptions from spoken interaction.

We have demonstrated how such tools can be used

- to quickly determine whether proposed complex measures (like vocabulary richness) have certain desired properties (like stability across number of tokens).
- to inspect the dynamics of properties like dominance, hesitance and uncertainty in a number of distinct transcribed and recorded activities.
- to inspect the personal variability of a particular speaker with regard to different properties like those mentioned above.
- to inspect the influence of different individual background variables such as gender on the distribution of a particular property.
- to inspect how different activities vary with respect to properties which have been coded into the database.

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