

# Structured Speech Input for Clinical Data Collection

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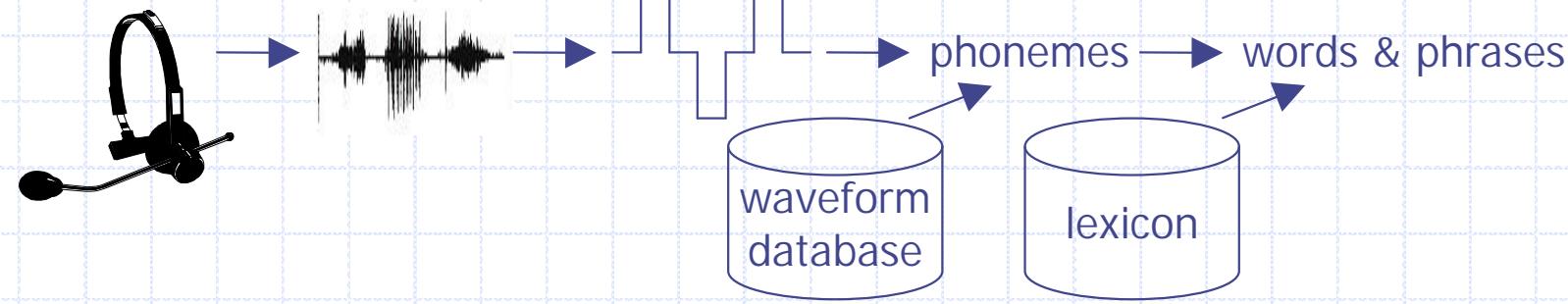
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# Introduction

- ◆ A common application of speech recognition software is to dictate clinical narratives and reports.
- ◆ However, free-form dictation has several problems.
  - ☞ Inefficient operation.
  - ☞ High error rate.
  - ☞ Low user acceptance.
  - ☞ Can not analyze narratives by computer automation.
- ◆ We address these limitations with the following.
  - ☞ Structured speech input to increase efficiency and reduce speech recognition errors.
  - ☞ Nomenclature types and validation rules to support a wide range of clinical studies.
  - ☞ Strongly typed observations in an event-oriented data model for distribution and analysis according to study protocol.

# How Speech Recognition Works

- ◆ Speak into the microphone.
- ◆ Convert signal to digital waveform.
- ◆ Convert digital information to phonemes.
  - ☞ A phoneme is the smallest unit of speech.
  - ☞ Compare to database of waveforms for all known phonemes.
- ◆ Assemble phonemes into words and phrases.
  - ☞ Use lexicon and stochastic model.



# Recent Speech Recognition Trends

- ◆ Large vocabulary.
  - ☞ Systems can recognize up to 40,000 words or more.
- ◆ Speaker independence.
  - ☞ Little or no training is required to teach the system to recognize your voice.
- ◆ Continuous recognition.
  - ☞ You can speak in a natural rhythm with no pauses between words or phrases.
- ◆ Major speech recognition products.
  - ☞ IBM ViaVoice ([www.ibm.com/software/speech/](http://www.ibm.com/software/speech/)).
  - ☞ Dragon NaturallySpeaking ([www.lhs.com/naturallyspeaking/](http://www.lhs.com/naturallyspeaking/)).

# Clinical Dictation with Speech Recognition

- ◆ Dictate medical narratives directly into a computer.

- ☛ Used in Radiology, Pathology, and Endoscopy.

- ◆ Advantages.

- ☛ Clinicians speak in a natural manner.
  - ☛ Narratives do not need to be transcribed by clerical personnel.
  - ☛ Decrease turnaround time.

- ◆ Disadvantages.

- ☛ Computer can only recognize words, but not their meanings.
  - ☛ Clinical narratives can not be interpreted or analyzed.
  - ☛ Free-form text dictation has a high error rate.

# Research Objectives

- ◆ Use structured speech input to increase efficiency and minimize speech recognition errors.
- ◆ Constrain speech input through a set of nomenclature types and validation rules that can be customized for a wide range of studies.
- ◆ Store speech input as a series of coded and strongly typed observations to facilitate data analysis and interpretation.

# Materials

- ◆ Software prototype.

- ❖ Pentium-based computer running Windows 98.
  - ❖ Dragon NaturallySpeaking and the Microsoft Speech API.
  - ❖ Microsoft Visual Basic and Microsoft Access.
  - ❖ Roughly 1,500 lines of code.

- ◆ Event-oriented data model.

- ❖ Supports a list of observations for each subject on a study.

- ◆ Validation rules.

- ❖ Ensure the completeness and accuracy.
  - ❖ Follow a simple “if-then” format based on Boolean criteria.
  - ❖ Used to identify missing data, verify normal ranges, and identify trends in longitudinal data.

# Methods

## ◆ Evaluation.

- ✉ Prototype supported microscopic pathology data collection.
- ✉ Compared structured vs. unstructured speech input.
- ✉ Twenty pathologists in a within-groups design counterbalanced on structure and task order.
- ✉ Measured speed, accuracy, and acceptance.

## ◆ Extrapolation.

- ✉ Added configurable nomenclature types and validation rules.
- ✉ Added an event-oriented data model.
- ✉ Five subjects attempted to apply the system to a wide range of clinical studies.

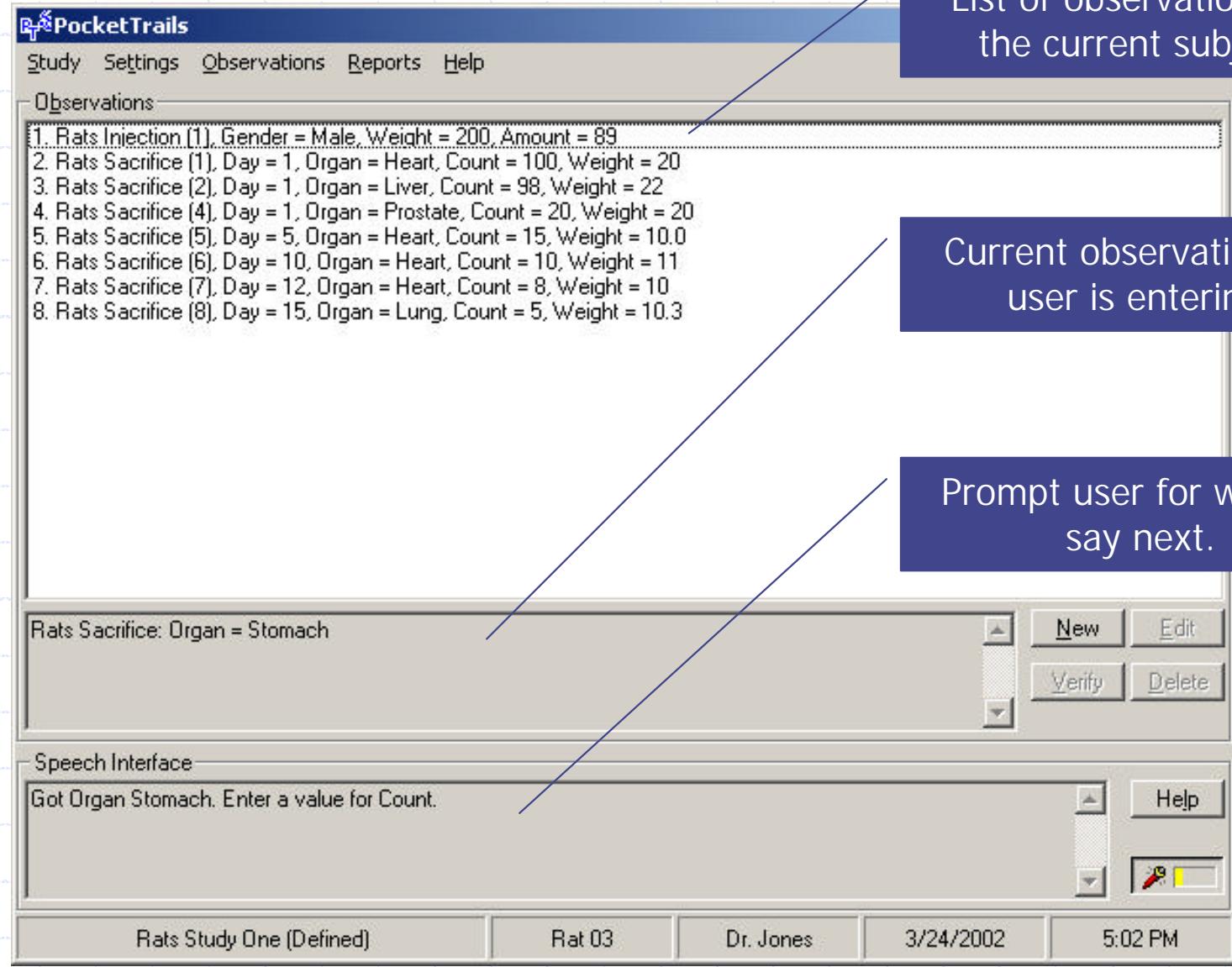
# Results

- ◆ Evaluation of microscopic pathology prototype.
  - ✉ A 22% decrease in task completion time ( $p < .001$ ).
  - ✉ A 36% reduction in speech recognition errors ( $p < .01$ ).
    - ✉ Reduction due to the application of user interface principles, not improved speech recognition algorithms.
  - ✉ A 2.4% improvement in user acceptance ( $p < .05$ ).
  
- ◆ Extrapolation of prototype to various applications.
  - ✉ Collect animal drug toxicology data.
  - ✉ Collect aquatic pathology data.
  - ✉ Compute the dose deposition of tumor antiferritin in tumor versus normal tissue.
  - ✉ Collect survey data for a Pfiesteria cohort study.
  - ✉ Collect patient vitals.

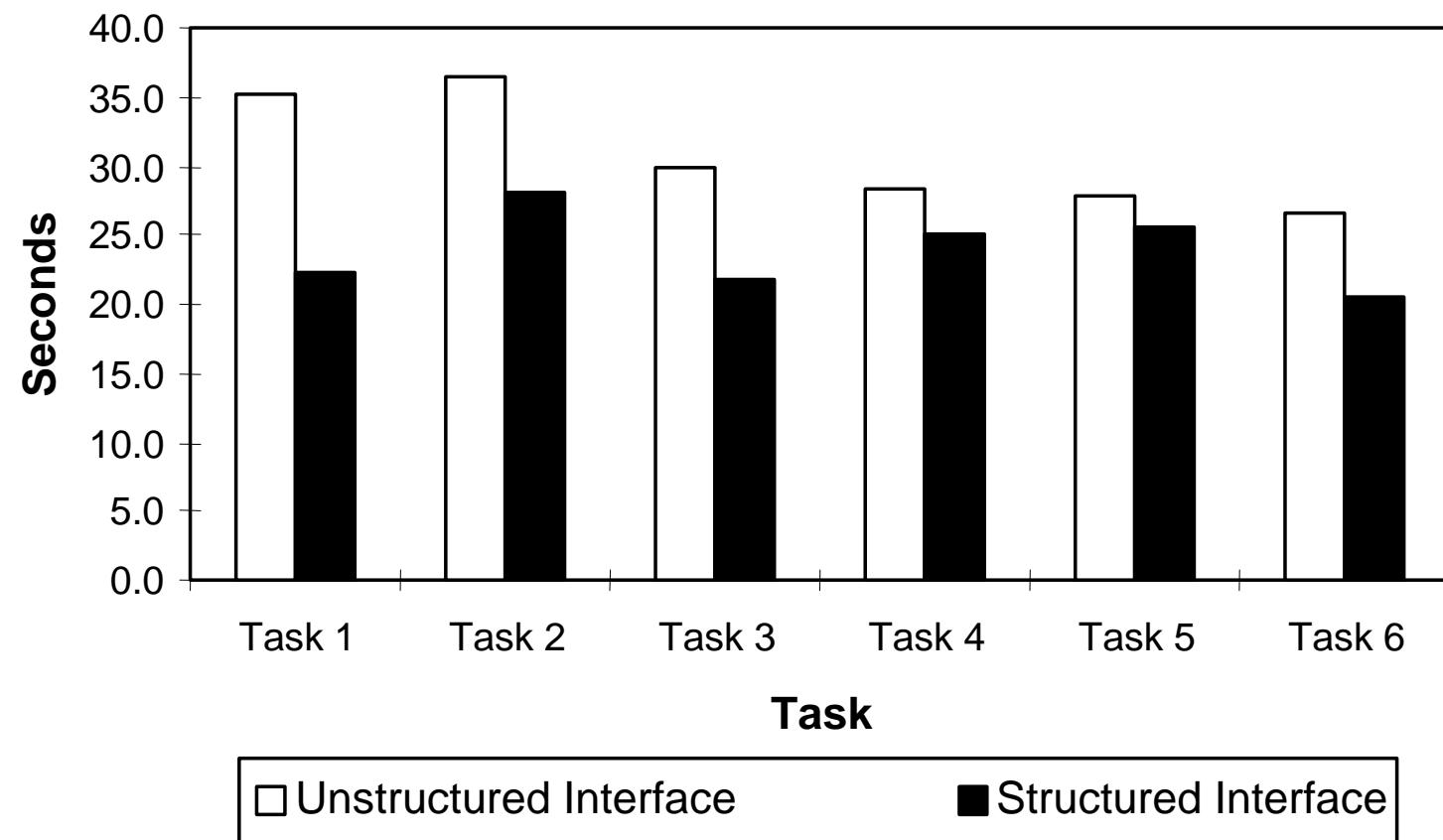
# Conclusions

- ◆ Successfully developed a prototype speech-driven system for hands-free clinical data collection.
- ◆ The prototype address the limitations of free-form clinical dictation using the following.
  - ☞ Structured speech input increase efficiency, reduce speech recognition errors, and increase user acceptance.
  - ☞ Nomenclature types and validation rules to support a wide range of clinical studies.
  - ☞ Strongly typed observations in an event-oriented data model for distribution and analysis according to study protocol.

# Data Collection Prototype

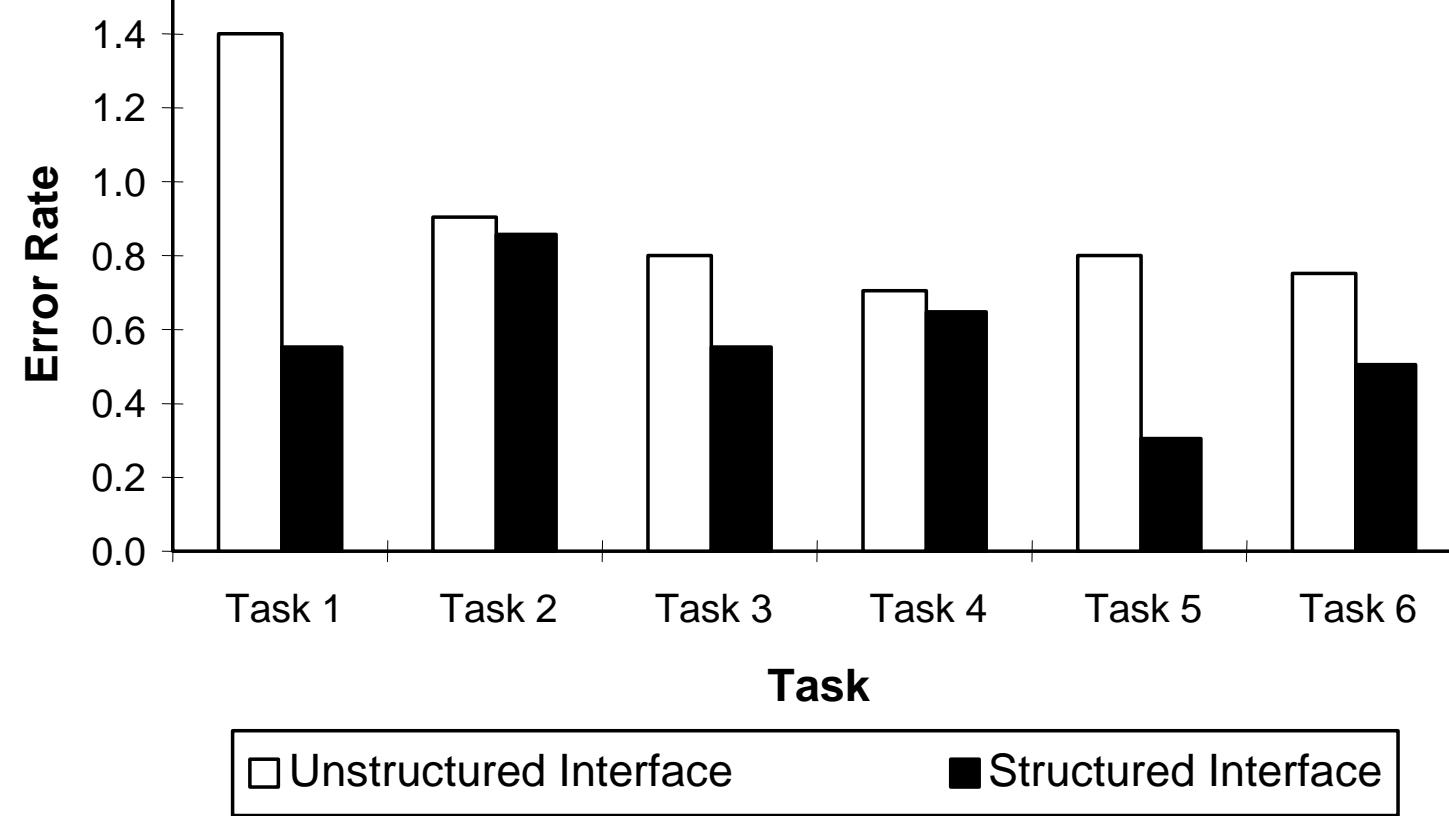


# Mean Task Completion Times



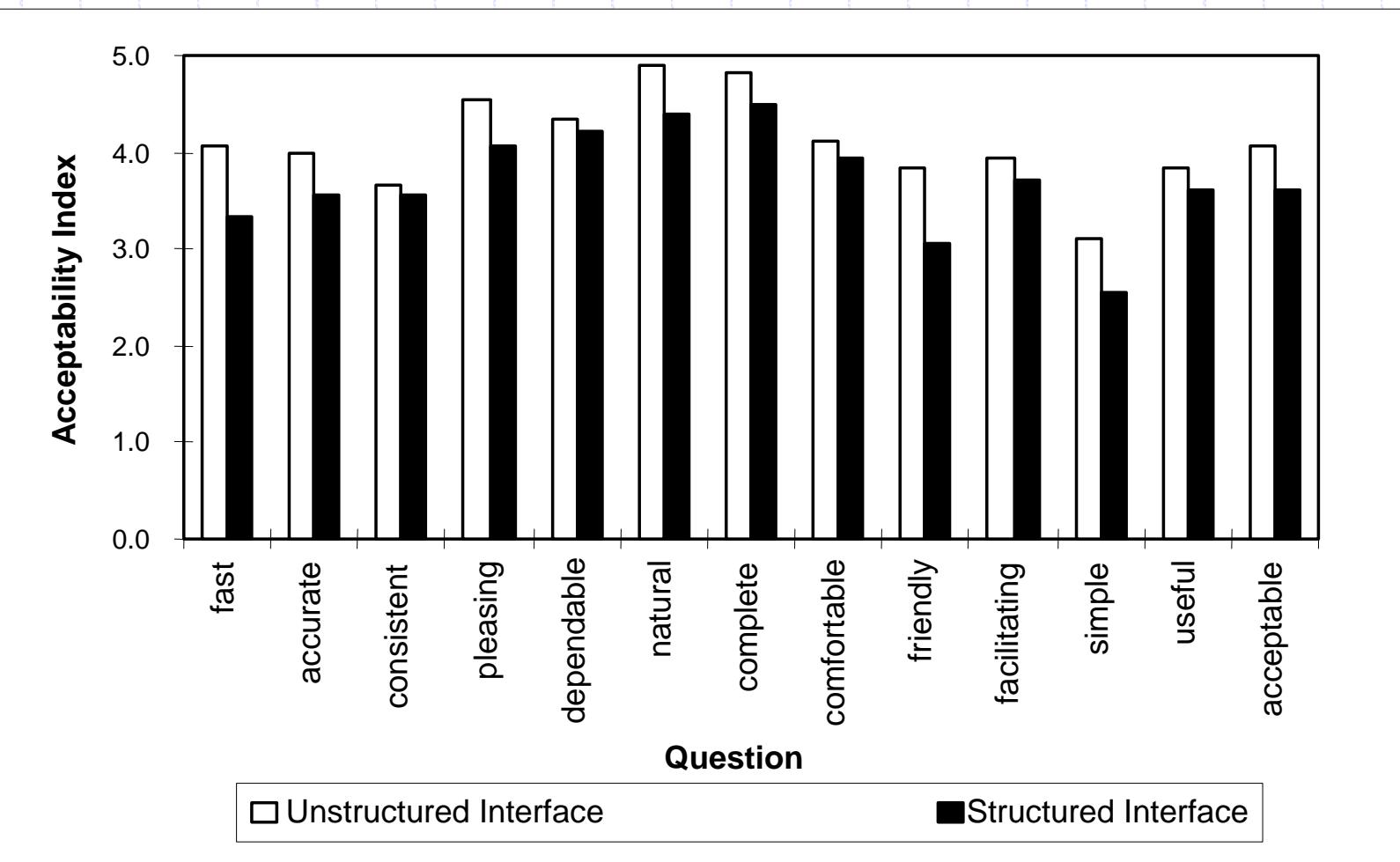
A 22% decrease in task completion time ( $p < .001$ ).

# Mean Speech Error Rates



A 36% reduction in speech recognition errors ( $p < .01$ ).

# Acceptability Index by Question



A 2.4% improvement in user acceptance ( $p < .05$ ).

# References

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