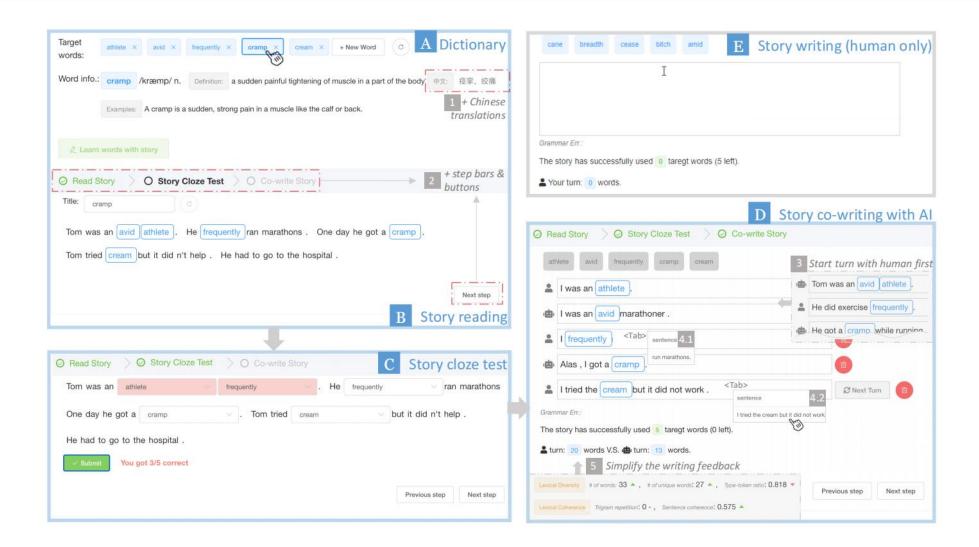


Storyfier: Exploring Vocabulary Learning Support with T ext Generation Models

Zhenhui Peng, Xingbo Wang, Qiushi Han, Junkai Zhu, Xiaojuan Ma, Huamin Qu

ACM UIST





conquer the shortage of existed vocabulary learning tools with context

Motivaion



• Existed tools fall short in 2 aspects:



Only leverage existing materials



lack of flexibility



Use language receptively



lack of prodctive task to master usage of words

Design & develop phases



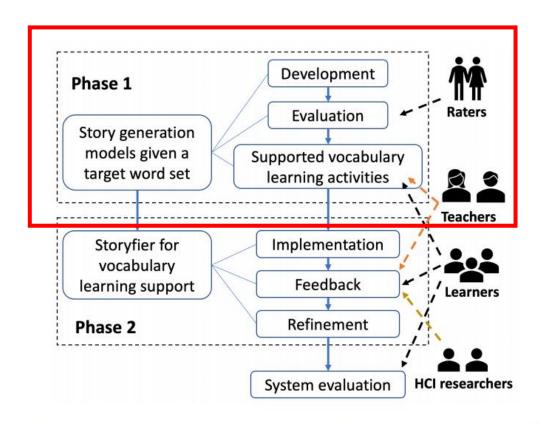


Figure 1: Our two-phases design and development process of *Storyfier* with teachers, learners, and HCI researchers.

In the whole 2-phases develop procedure, phase 1 mainly consists of 3 steps:

- develop story generation model;
- **validation** of the model;
- explore learning activities supported by this m odel.

develop a story generation model for target words context generation

Model development



Target: controllable model which can generate stories with target word sets

Attributes	Values	
# of stories	101,661	
# of words	4,640,319	
Average story length	45.65	
Average sentence length	7.80	
Average readability	57.14	
Coverage of CET-4 words	89.52%	



tuple:{ story title, target words, sto
ry sentences}



validate the quality of generated stories

Model validation



- Validating object: 20 human-machine story pairs
 - 20 stories from ROCstory with varied difficulty levels of contained words
 - 20 stories generated by model based on corresponding story title and contained words
- Validation methods: technical and human

Automatically evaluated

7). 7)	Grammar	Type-token ratio	0	Sentence coherence
Human	1.00	0.75	0.01	0.42
Machine	1.00	0.77	0.01	0.43

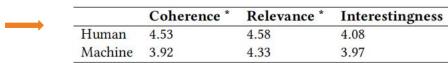
Adequate

Overall

4.26

4.01

Evaluated by English skilled people



conduct interview to explore learning activities that can be supported by the model

Explore learning activities





- Explore possible activities by conducting **semi-structured interviews**
- **interviewees:** skilled English **teacher** and well-experienced **students** in using vocabulary learning tools

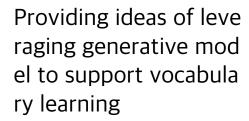








Interact with rough interf ace to try generative mo del and brainstorm



- Story reading
- Cloze test
- Turn-taking writing

Design & develop phases



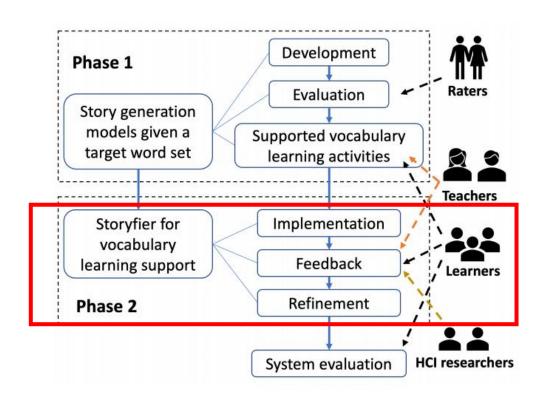


Figure 1: Our two-phases design and development process of *Storyfier* with teachers, learners, and HCI researchers.

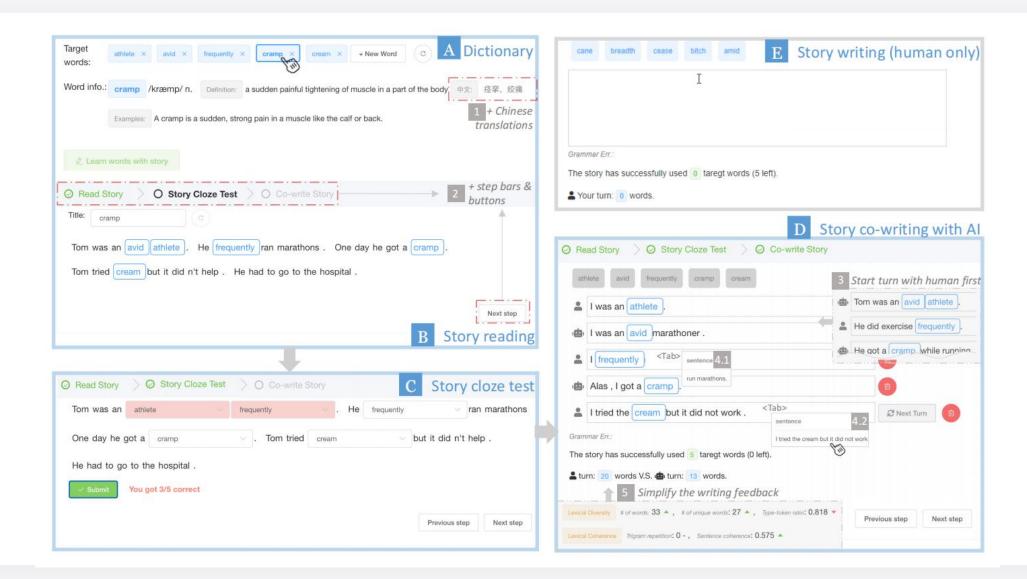
In the whole 2-phases develop procedure, phase 2 mainly consists of 3 steps:

- system implementation;
- tests for feedback;
- refinement according to feedback.

design process about how to implement storyfier

System implementation





Test for feedback & refinement



• **Target**: not to evaluate system effectiveness but to **improve** with quick feedback

Usability test with **12 ES L learners**

- Storyfier is more use ful than baseline with out context;
- Some have difficulti
 es in writing next se
 ntence in turn-taking
 writing.

Workshop with **English teachers**

- Maximize 3 activitie s's value by chaining into a flow;
- Add main translation

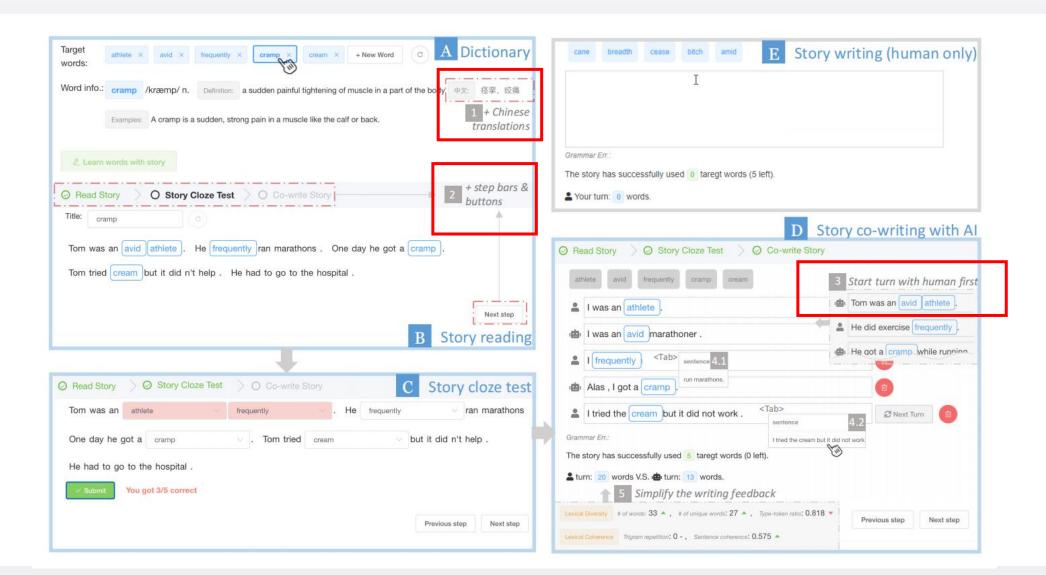
Workshop with **HCI res** earchers

- Use clear widgets in learning flow;
- **Remove metrics** to reduce distraction

refinement after test and workshops

System refinement



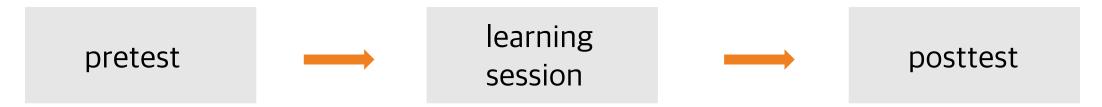


verify the impact of Storyfier on vocabulary learning

Experiment



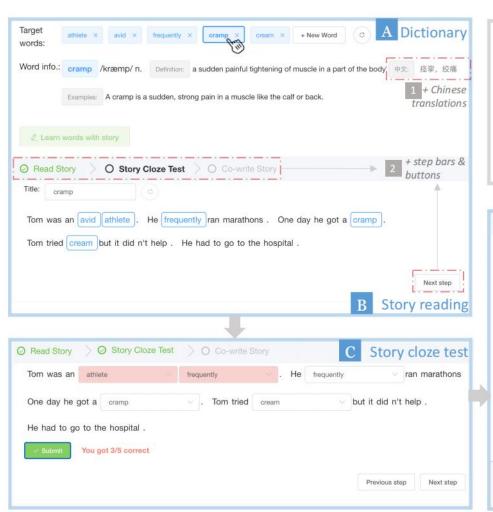
- **Subjects**: 28 ESL Chinese students (1/12 have not passed CET-4)
- **Conditions**: 2 (with vs. without AI features) x 2 (read-only vs. read-cloze-write activities)
- Research Questions:
 - Learning outcome
 - Experience
 - Perceptions towards Storyfier
- Procedure:



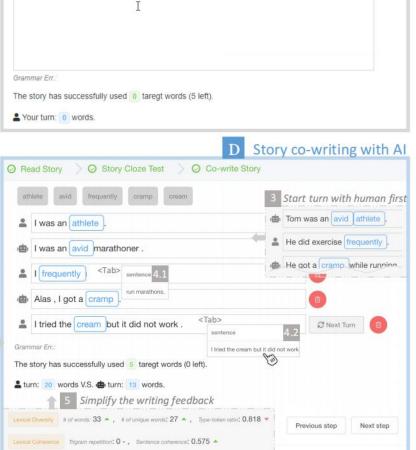
verify the impact of Storyfier on vocabulary learning

Experiment









- Read-only
 - with AI A+B
 - without Al A

- Read-cloze-write
 - with AI A+B+C
 - without Al A+B+D

Learning out come

Analyses and results



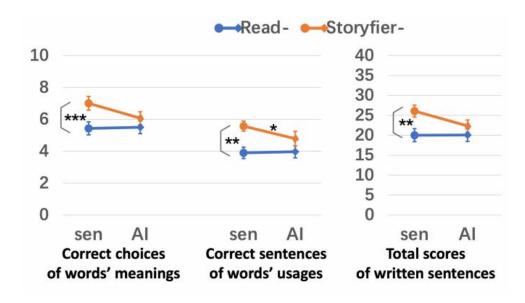


Figure 5: RQ1 results regarding numbers of correct choices on target words' meanings, numbers of sentences that correctly use target words, and total scores of the written sentences in each condition. ***: p < 0.001, **: p < 0.01, *: p < 0.05.

- Read-cloze-wirte learning session has better perfor mance comparing to plain reading session;
- Storyfier's **AI features reduce learning gains** on ret ention words' meanings in read-cloze-wirte learning session.

Analyses and results



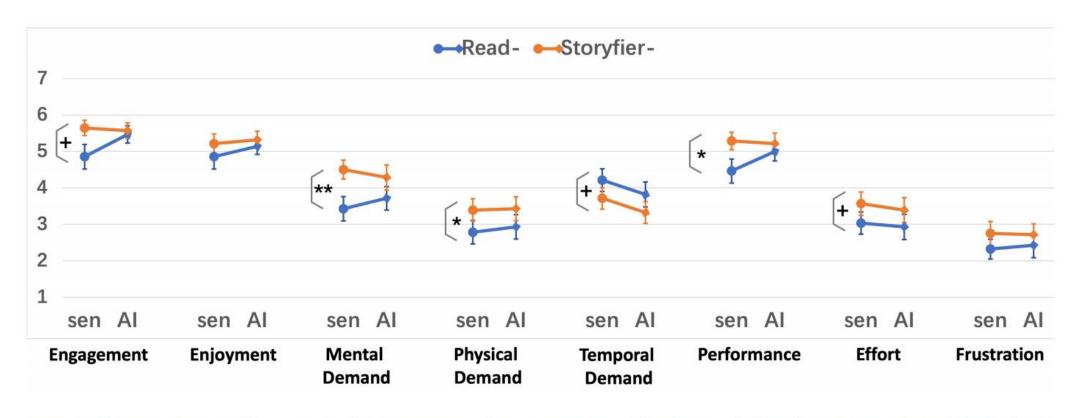


Figure 6: RQ2 results regarding perceived engagement, enjoyment, and workload in vocabulary learning sessions with Read-sen, Read-AI, Storyfier-sen, and Storyfier-AI interfaces. **: p < 0.01, *: p < 0.05, +: p < 0.1.

user perceptions

Analyses and results



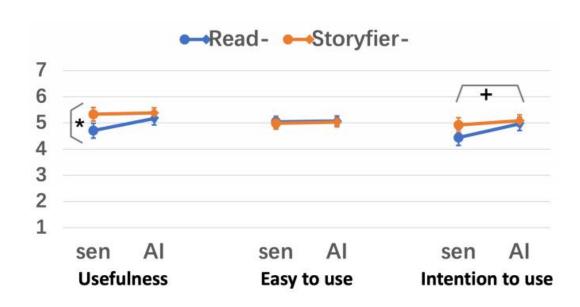


Figure 7: RQ3 results regarding user perceptions with each interface. *: p < 0.05, +: p < 0.1.

• quantitative:

- prefer read-cloze-writing session than read-only;
- find AI-generated stories more useful and feel re ady to use in read-only session.

• qualitative:

- participants generally prefer Storyfier with AI gen erative models;
- model should be further improved regarding coherence, complexity and style of content.

insights from findings

Discussion



generative model:

- not enough coherent when 5 target words are not naturally relevant;
- stories contain unknown words -> content too complex.

learning activities:

read-cloze-writing session provides more learning gains.

impact of genarative model in learning:

- though AI-generated stories do not improve learning gains -> more coherent could help;
- Al support leads to reducesd learning gains in read-cloze-writing sessions -> participants favor **assistance** from generative model in writing.

design condsiderations & limitations

Discussion



design consideration

- integrate mutiple activities to provide in-situ assistance;
- provide feedback relevant to learning goals;
- balance machine and human efforts

limitation

- Only consider main meaning of vocabularies;
- quality of stories (balance trade-off between simplicity and coherence);
- no quantitative evaluation of contribution from cloze and writing seperately.



Thank you