# **Interactive Learning of Dialog Scenarios from Examples** Daiga Deksne and Raivis Skadiņš (Tilde, Latvia; Faculty of Computing, University of Latvia, Latvia)

## Problem

- manually.
- examples.
- dialog corpus for end-to-end goal-oriented dialog system training.
- the next virtual assistant (VA) action from the conversation history.

### Experiments

• English DSTC dialog corpus used for choosing the optimal architecture of the model.

| Nr | Vectorization    | Dimensions | Entities | Entities set in previous turns | Test<br>accuracy |  |
|----|------------------|------------|----------|--------------------------------|------------------|--|
| 1  | wiki.en.bin      | 300        | 194      | no                             | 0.8395           |  |
| 2  | news.wiki.en.bin | 300        | 194      | no                             | 0.8303           |  |
| 3  | <b>BERT-Base</b> | 768        | 194      | no                             | 0.8209           |  |
| 4  | Intents          | 42         | 194      | no                             | 0.8360           |  |
| 5  | wiki.en.bin      | 300        | 20       | no                             | 0.8191           |  |
| 6  | news.wiki.en.bin | 300        | 249      | no                             | 0.8291           |  |
| 7  | news.wiki.en.bin | 300        | 498      | yes                            | 0.8388           |  |
| 8  | wiki.en.bin      | 300        | 498      | yes                            | 0.8398           |  |
| 9  | Intents          | 42         | 498      | yes                            | 0.8443           |  |

160 Latvian dialogs in transport domain collected using the developed platform.

| Nr | Number of dialogs | Average accuracy | Standard deviation |
|----|-------------------|------------------|--------------------|
| 1  | 33                | 0.8576           | 0.0582             |
| 2  | 52                | 0.8415           | 0.0469             |
| 3  | 68                | 0.8536           | 0.0610             |
| 4  | 82                | 0.8712           | 0.0617             |
| 5  | 97                | 0.8919           | 0.0351             |
| 6  | 122               | 0.8492           | 0.0378             |
| 7  | 128               | 0.8468           | 0.0309             |
| 8  | 160               | 0.8617           | 0.0231             |

• Currently, machine learning techniques are used to train the models for intent detection and entity recognition, but dialog scenarios are usually created

• We explore the methods that will allow VA dialog scenarios to be learned from

• This paper reports on the development of a toolkit that enables collecting

• The toolkit includes the neural network model that interactively learns to predict



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