



<https://vsr.informatik.tu-chemnitz.de/projects/2019/growth>



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VISH: Does Your Smart Home Dialogue System Also Need Training Data?

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Introduction

Natural Language Understanding in Smart Home

It's too dark in this room



**Automatic Speech
Recognition**

**Speech To
Text**

**Natural Language
Processing**

**Device
Actuations**

Research Challenge

To ensure understanding of the user goal and provide good user experience, a **high-quality** and **sufficiently** large training corpus is required



Common Process

**Record sample
utterances from users**

**Create labelled
training dataset**



**Transcribe
utterances to text**

Train NLU model



Approach

<https://www>

Goal-oriented WoT Interfaces

GWoT is defined as $\text{GWoT} = \langle S, s^0, A, s^*, \gamma \rangle$

S

$S = \{s_1, s_2, \dots, s_n\}$ includes s^0 and s^*
 s_i represented by $\text{Par}_s = \{p_1, p_2, \dots, p_k\}$

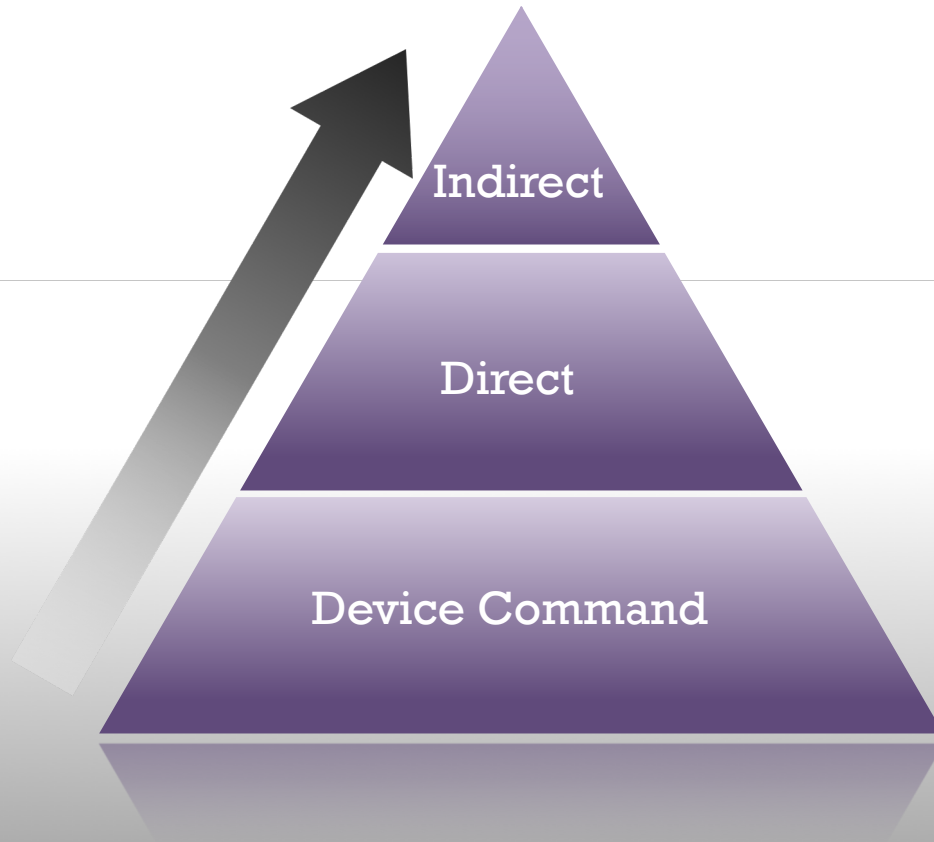
A

$a \in A$ is $a = \langle \text{effects}(a), \text{actuation}(a) \rangle$
 $\text{effects}(a) = \langle p_i, t, v \rangle$
 $\text{effects}(a) \in \{\text{increase}, \text{decrease}, \text{assign}, \text{toggle}\}$
 $\text{actuation}(a)$: set of HTTP / MQTT requests

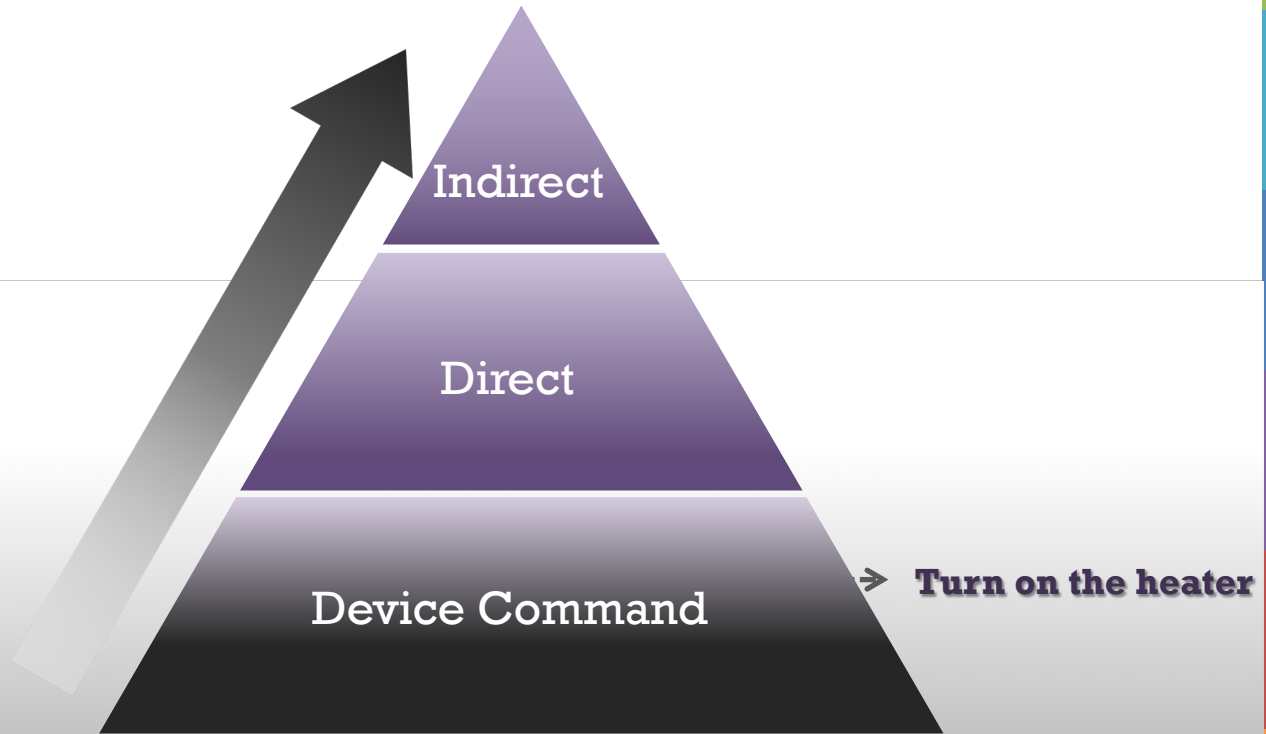
γ

$\gamma: S \times A \rightarrow S$

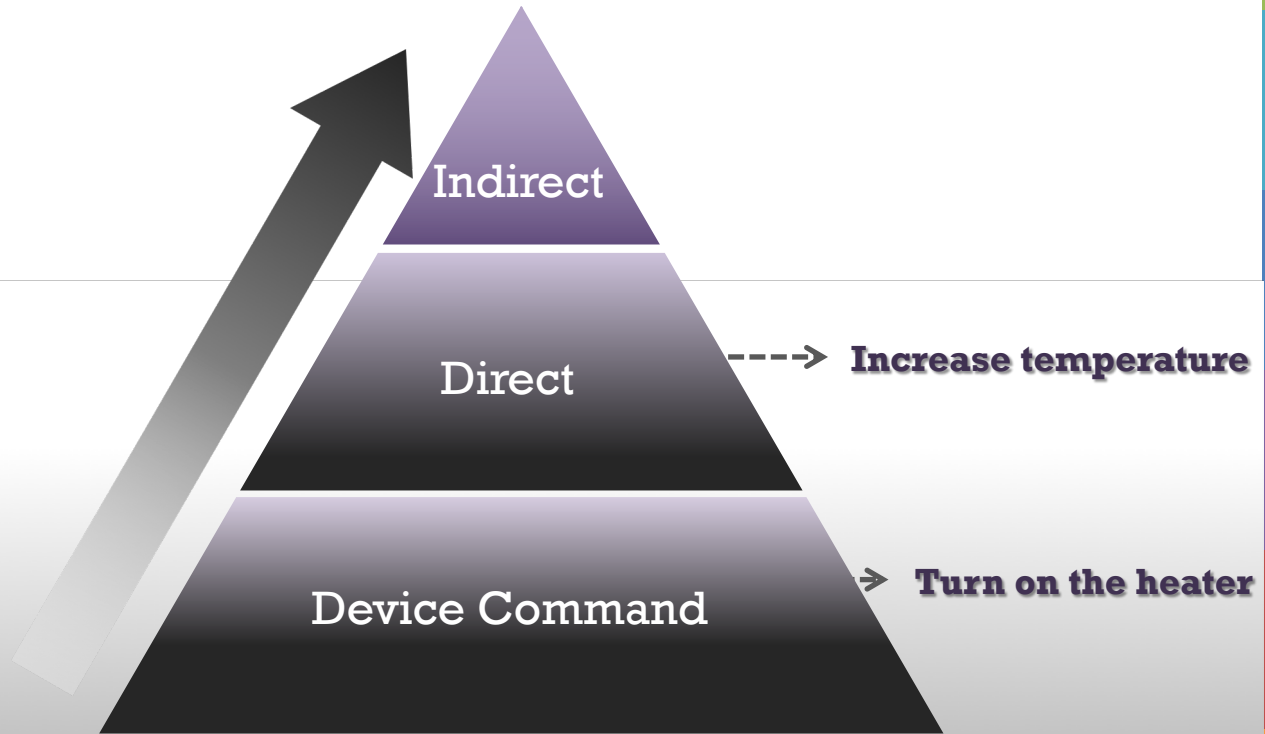
User Goals in Smart Home



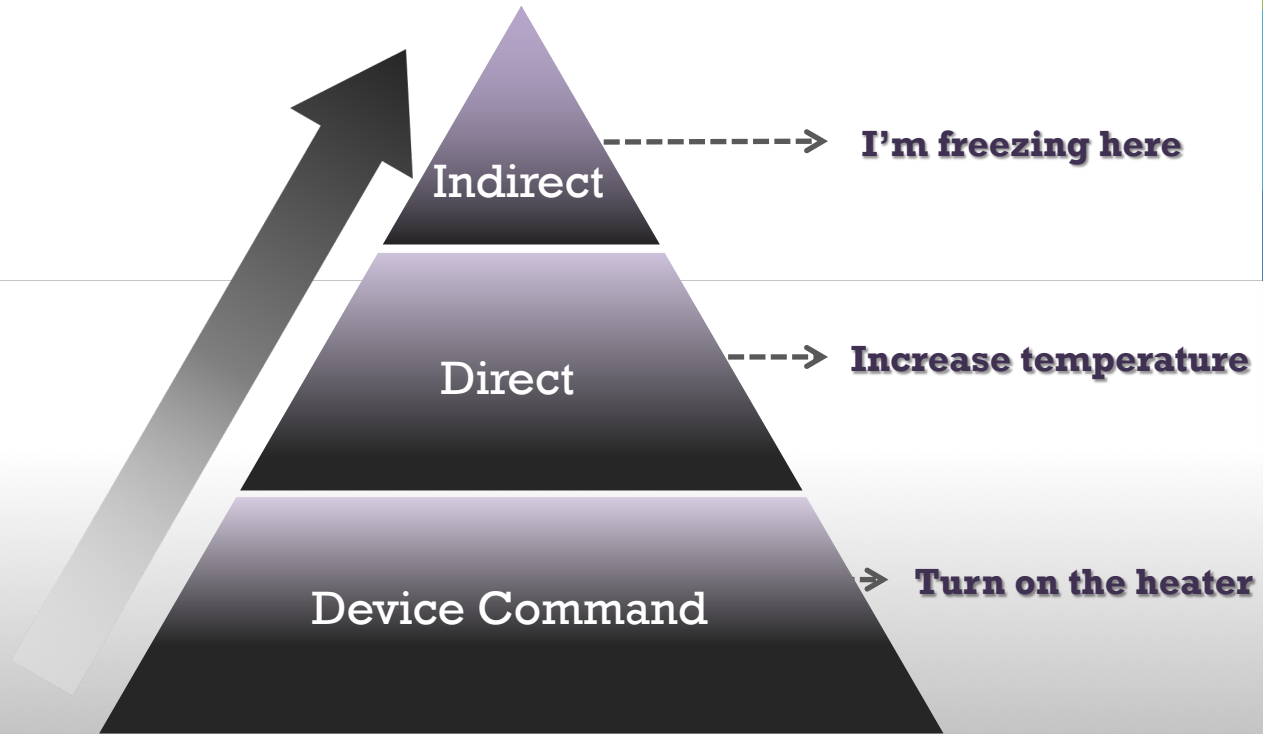
User Goals in Smart Home



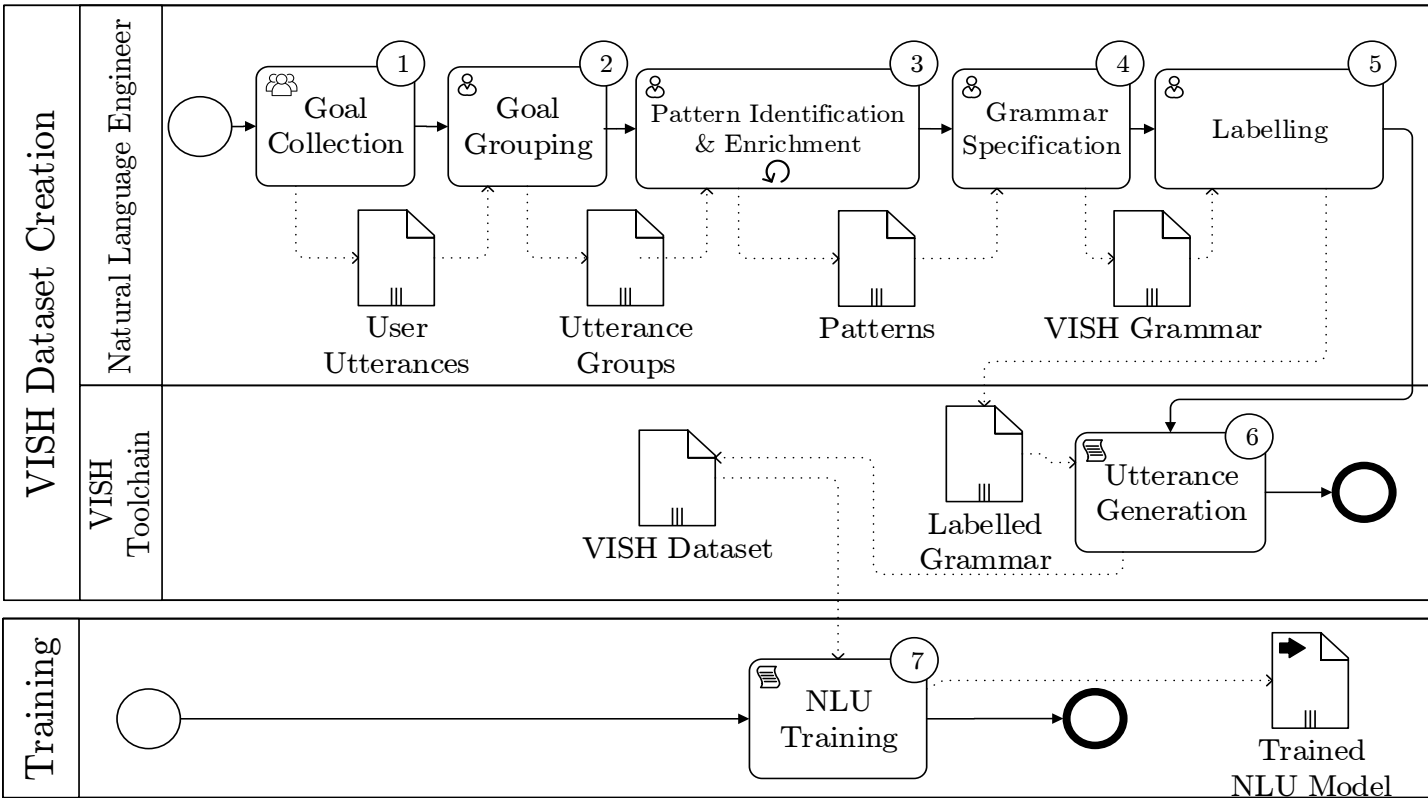
User Goals in Smart Home



User Goals in Smart Home

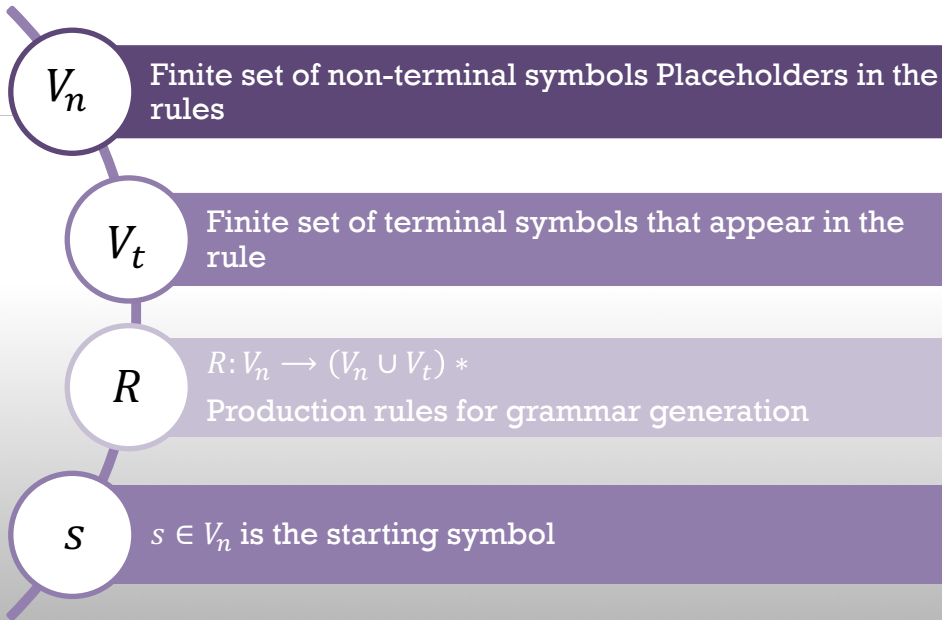


VISH: Voice Interface for Smart Home

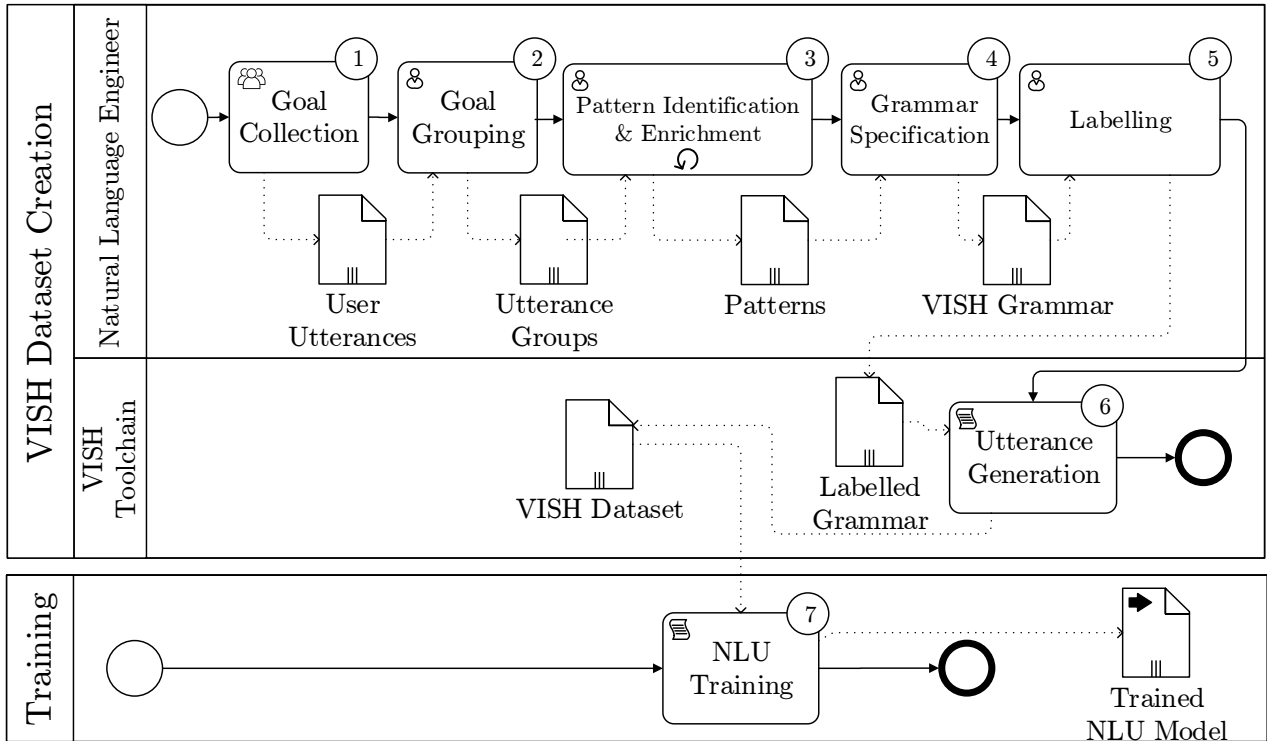


VISH Grammar Specification

VISH Grammar G is defined as $G = \langle V_n, V_t, R, s \rangle$



VISH: Voice Interface for Smart Home



Labelling for Extraction



Set the **DEVICE:thermostat** to 25 degrees

ACTION:Mix a smoothie

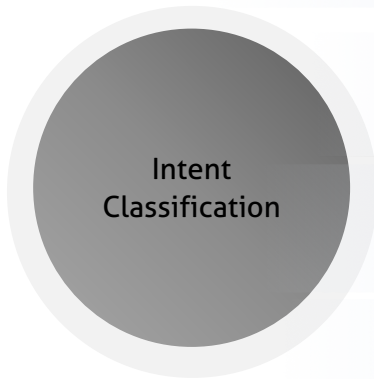
Change the colour to **VAL:blue**

Tell the robot to start **STATE:cleaning**

Turn the **LOCATION:bedroom** lights on

Brew **QUANTITY:three** tablespoons of tea

Labelling for Classification



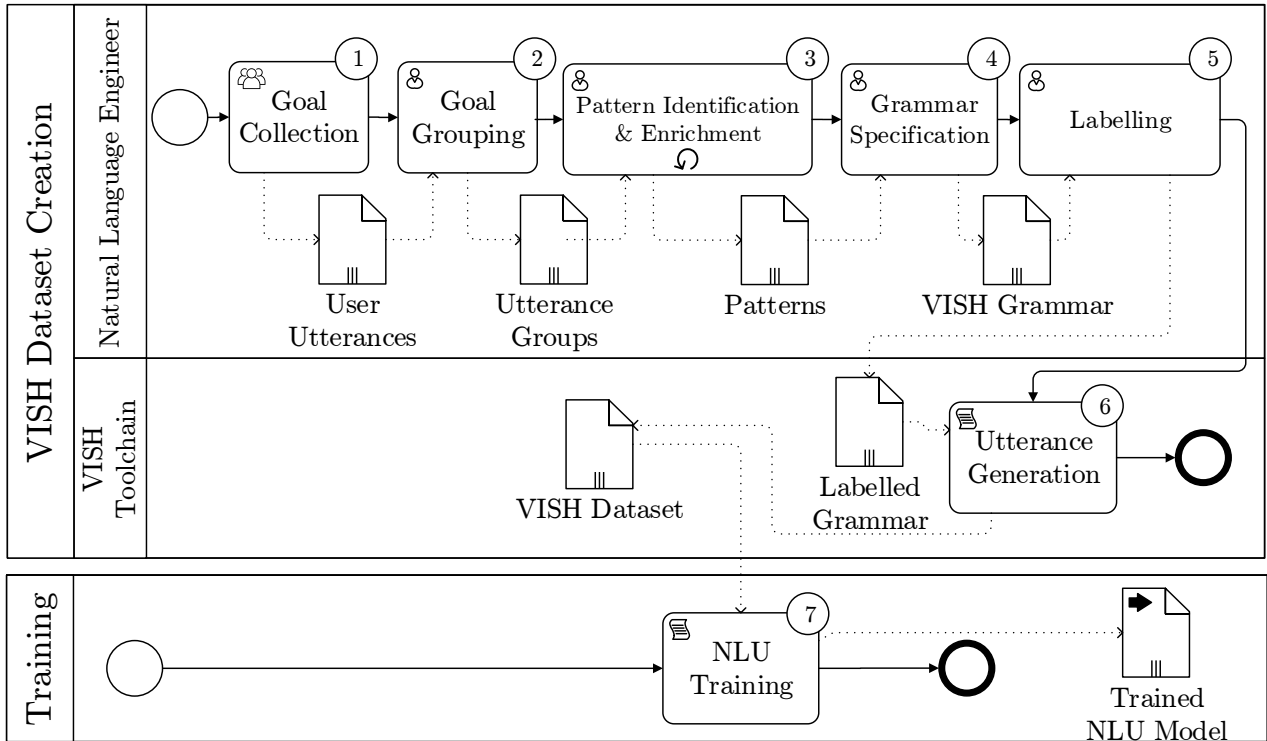
Physical Parameter : temperature, humidity, brightness

Effect Type : increase, decrease, toggle, assign

Scope : global, device

Goal Type : device command, direct, indirect

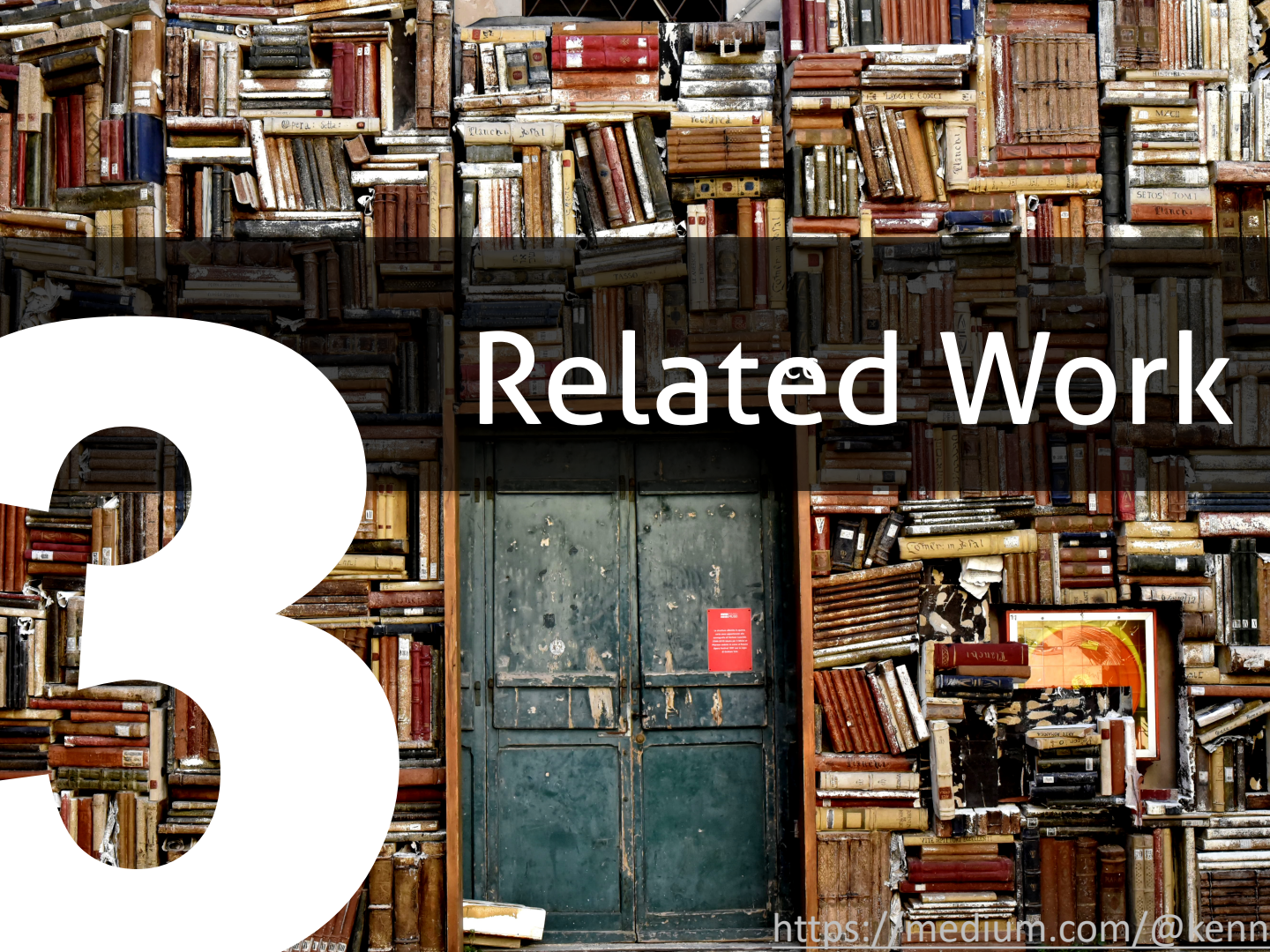
VISH: Voice Interface for Smart Home



VISH Dataset

<https://vsr.informatik.tu-chemnitz.de/projects/2019/growth/>

Total Words: 33,618,437	Entities: 6,819,062	Effect Types: 4
Utterances: 4,743,745	Indirect Goals: 4,00,854	Device Types: 305
Unique Words: 822	Direct Goals: 742,891	Action Types: 130
Entity Types: 7	Total Actions: 674,711	Parameter Types: 8



Related Work

<https://medium.com/@kenn>

Dataset	Size	Lan	Direct Goal	Indirect Goal	Availability	Pr	E	De	Ac	Approach
Google Home	NA	en	✓	✗	✗	✗	✗	✓	✓	User
Alexa	20K	en	✓	✗	✗	✗	✗	✓	✓	User
IFTTT	NA	en	🕒	✗	✓	✗	✗	✓	✓	User, ECA
DIRHA	183	en	✗	✗	🕒	✗	✗	✗	✗	Experiment
VocADo m@4H	7K	fr	✗	✗	✗	✗	✗	✗	✗	Synthetic & Experiment
Voice-Home2	1,5K	fr	✓	✗	✓	✗	✗	✗	✗	Experiment
Sweet-Home	1,5K	fr	🕒	✗	🕒	✗	✗	✗	✗	Experiment
ATHENA	370	el	🕒	✗	✗	✗	✗	✗	✗	Experiment
Genie	3M	en	✓	✗	✓	✗	✗	✓	✓	Synthetic & Crowdsourcing
VISH	5M	en	✓	✓	✓	✓	✓	✓	✓	User-derived generative grammar



Evaluation

Evaluation Procedure



Dataset preparation : Rasa markdown syntax

Dataset division: 80% training, 20% holdout



Dataset training on NLU models: supervised embedding, pretrained embeddings convert, pretrained embedding spacy

Testing the NLU models: accuracy, precision, F1-score



Evaluation Results

	Supervised 100	Supervised 300	Supervised 500	ConveRT	Mean
<i>Precision</i>	.9834	.9849	.9886	.9756	.9831
<i>Recall</i>	.9834	.9849	.9886	.9756	.9831
F_1	.9834	.9849	.9886	.9756	.9831

*Suitability of VISH dataset for training
state-of-the-art NLU intent classification for
smart home utterances*

Evaluation Results

		State	Quantity	Value	Location	Device	Unit	Action	Mean
Custom	P	1.0	1.0	.9980	1.0	.9999	1.0	.9998	.9997
	R	1.0	1.0	.9999	1.0	.9998	1.0	.9995	.9999
	F_1	1.0	1.0	.9989	1.0	.9998	1.0	.9997	.9998
SpaCy	P	1.0	1.0	.9992	1.0	.9999	1.0	.9998	.9998
	R	1.0	1.0	.9983	1.0	.9998	1.0	.9995	.9997
	F_1	1.0	1.0	.9988	1.0	.9998	1.0	.9996	.9998

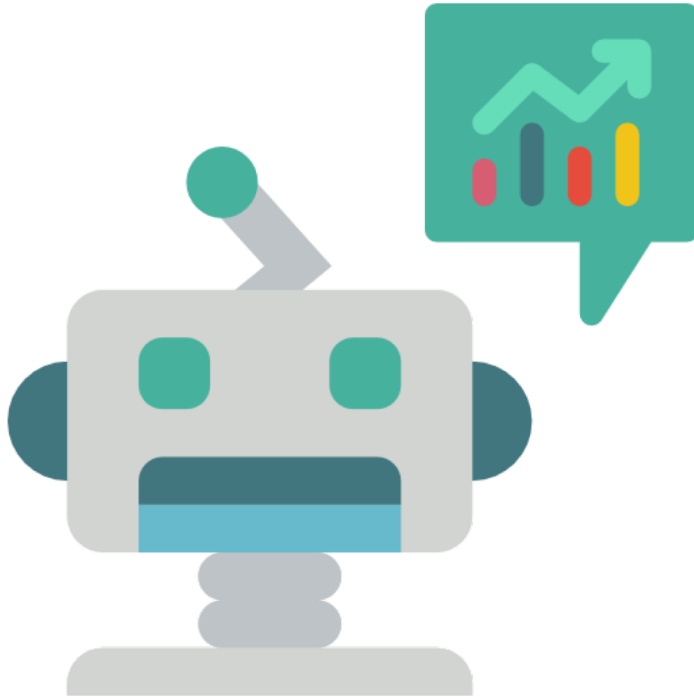
Suitability of VISH dataset for training
state-of-the-art NLU entity extraction for
smart home utterances



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Conclusion

Conclusion & Future Work



VISH dataset can be used to train goal-oriented dialogue systems for smart homes with **high accuracy**

Integrate custom NLU pipeline trained on the VISH dataset

Evaluate the dataset with end-user goals in live IoT settings



VSR

Thank You!

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