

# A CORPUS OF GERMAN CITIZEN CONTRIBUTIONS IN MOBILITY PLANNING: SUPPORTING EVALUATION THROUGH MULTIDIMENSIONAL CLASSIFICATION

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## Motivation

- Political authorities in democratic countries regularly consult the public in order to allow citizens to voice their ideas and concerns on specific issues.
- When trying to evaluate the (often large number of) contributions in order to inform decision-making, authorities often face challenges due to restricted resources [8, 1].
- Pre-structuring with supervised machine learning can make the evaluation process more efficient.
- We identified four common tasks in dialogue with practitioners whose support through automation would benefit the evaluation of participation processes:
  - i) the detection of arguments
  - ii) assessing argument concreteness
  - iii) locating the contributions
  - iv) structuring according to topics
- We introduce the **CIMT PartEval Corpus** including annotations for several thousand citizen contributions from six mobility-related planning processes in five municipalities in German language.

## Overview of the CIMT PartEval Corpus

task	unit level	total units	datasets					
			CD_B	CD_C	CD_M	CQ_B	MC_K	MC_O
i) argument components [4]	sentences	17,852	10,442	1,704	2,193	1,505	2,008	
ii) argument concreteness [5]	sentence spans	1,127	679	92	110	55	191	
iii) geographic location [6]	token spans	4,830	4,087	743				
iv) thematic categorization [7]	documents	697						697

Information on the datasets, deriving from public participation processes in German municipalities:  
**CD\_B, CD\_E, CD\_M:** Map-based online dialogues for improving cycling in three cities, i.e. Bonn, Cologne (district Ehrenfeld), and Moers.  
**CQ\_B:** CD\_B was supplemented with a representative survey of the population (questionnaires).  
**MC\_K:** The city of Krefeld held a map-based online dialogue for developing a mobility concept.  
**MC\_O:** The district office of Altona (city of Hamburg) implemented a map-based online dialogue as part of the transformation of the quarter Ottensen into a traffic-calmed neighborhood.

## Subcorpus i) Argument Components

- Citizens communicate their ideas through arguments. In spatial planning processes, this usually involves describing a problem or condition, from which a proposition is derived.
- We focus on the identification of argumentative sentences and on the classification of two clausal properties using the following annotation scheme:
  - Major positions (mpos):** options for actions that are being proposed
  - Premises:** reasons that attack or support a major position or another premise
  - Non-argumentative (non-arg):** sentences without premise or major position
- Annotation process:
  - Ten percent of the dataset was coded by three coders, the remainder split equally.
  - Overall Fleiss' kappa agreement between 0.72 (CQ\_B) and 0.80 (CD\_C)
  - Coding inconsistencies were solved in a curation step by two process supervisors.
- Overview of the resulting annotation of argument component categories for the datasets:

	CD_B	CD_C	CD_M	MC_K	CQ_B	all
total	10,442	1,704	2,193	2,008	1,505	17,852
non-arg	1,153 (11.0%)	197 (11.6%)	382 (17.4%)	431 (21.5%)	172 (11.4%)	2,335
mpos	2,851 (27.3%)	603 (35.4%)	404 (18.4%)	961 (47.9%)	1,083 (72.0%)	5,902
premise	6,700 (64.2%)	951 (55.8%)	1,452 (66.2%)	685 (34.1%)	373 (24.8%)	10,161
overlap	262 (2.5%)	47 (2.8%)	45 (2.1%)	69 (3.4%)	123 (8.2%)	546

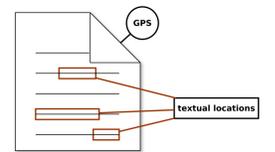
- With multiple processes differing in format and process purpose, this subcorpus facilitates a comprehensive evaluation of machine learning methods including their potential for generalization [3].

## Subcorpus iii) Geographic Location

- The geographic location of citizens' contributions can be of great importance to the evaluation as it allows geo-referencing of contributions and clustering of ideas by location.
- Depending on the data source, geo-references may not be available (e.g. paper questionnaires).
- To determine the geographic coordinates of a document's associated location by its textual content (*text-based geo-location*), we provide annotations of *textual locations* and *GPS coordinates*.

**Textual location:** Sequence of words that refers to the spatial placement of a contribution (named entities or constructions with more fine-grained location information that can be unambiguously marked on a map).

**GPS coordinates:** Spatial reference (point)



- Annotation process:
  - Textual locations:
    - \* Ten percent of the dataset was coded by three coders, the remainder split equally.
    - \* Krippendorff's alpha of 0.75
    - \* Coding inconsistencies were solved in a curation step by two process supervisors.
  - GPS coordinates: Data stems from map-based online dialogues where citizens had to assign their contributions on a map to a spatial point of reference.
- The 2,529 contributions that were annotated contain on average two textual locations and exactly one assigned GPS coordinate. Textual locations span on average 4.9 tokens ( $\sigma = 3.48$ ).
- About twelve percent of the tokens included in the contributions are part of a location phrase, a proportion that highlights the task's importance for spatial planning processes.

## Subcorpus ii) Argument Concreteness

- The less specific citizens' ideas are, the more difficult and time-consuming it will usually be for evaluators to derive measures for implementation.
- Automated evaluation of the concreteness of argument components can help practitioners filter out arguments that can be evaluated immediately.
- Our annotation scheme allows for a subdivision into three levels of concreteness:
  - High:** The *what, how and where* is specified (e.g. colour, surface, measurements).
  - Intermediate:** Some specifications are given (location, measures), but there is still room for interpretation.
  - Low:** No specific information (e.g. on location or description of measures) is contained.
- Annotation process:

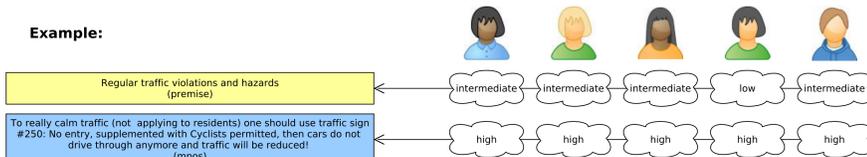
- Coders first interrelated sentences of the same type (i.e. premises or major positions) to form units with coherent sense. Annotation supervisors resolved inconsistencies.

	CD_B	CD_C	CD_M	MC_K	CQ_B	all
mpos	265	40	40	126	42	513
premise	414	52	70	65	13	614
total	679	92	110	191	55	1,127

Tab. 3: Units of interrelated argument components.

- Coders then rated the resulting units' concreteness (five ratings per unit).

### Example:



- Due to the subjective nature, we dispense with a manual curation step in which an unambiguous assignment of concreteness to units is made.
- Perception of concreteness is subjective but not arbitrary (Krippendorff's alpha<sub>w</sub> of 0.46).
- Analysis reveals that citizens clearly tend to write highly concrete arguments. Only twenty percent of the argument units have intermediate or low concreteness.

## Subcorpus iv) Thematic Categorization

- Thematic categorization of citizen contributions enables the joint analysis of contributions with similar topics and the delegation of contributions to the responsible administrative units.
- Most previous work fitted their categorization scheme to the participation processes under consideration. However, from-scratch development for new processes is impractical [2]:
  - The time required for schema development and training data annotation may quickly exceed the effort of a purely manual analysis, especially for processes with fewer contributions.
- One solution is the development of universal categorization schemes.
  - We provide such a solution for transport-related processes.
- Annotation process:
  - All contributions were annotated by three coders.
  - Coding inconsistencies were solved in a curation step by two process supervisors, one an urban planner.
- Overview of the thematic categorization scheme for mobility (including category-wise Fleiss' kappa agreement and class support):

<b>non-motorized transport</b> (kappa=0.79, support=372)	<b>motorized transport</b> (kappa=0.75, support=394)	<b>moving traffic</b> (kappa=0.82, support=466)
<b>cycling</b> (kappa=0.80, support=293)	<b>public transport (local)</b> (kappa=0.71, support=58)	<b>stationary traffic (parking)</b> (kappa=0.75, support=214)
<b>walking</b> (kappa=0.69, support=180)	<b>public transport (long-distance)</b> (kappa=0.60, support=1)	<b>new services</b> (kappa=0.60, support=26)
<b>scooters</b> (kappa=0.86, support=17)	<b>commercial transport</b> (kappa=0.76, support=55)	<b>inter- and multimodality</b> (kappa=0.75, support=1)

- Annotation efforts are ongoing to
  - provide a sufficient data basis for minority classes.
  - add further requirements for public space (e.g. noise, accessibility, quality of stay) to the scheme.

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