Generating Furniture for Augmented Reality Applications using Natural Language

Sebastian Buntin
Gothaer Bildungsgesellschaft mbH
Buergeraue 2
Gotha, Germany
sebastian@buntin.de

ABSTRACT
In this paper, I present a method for modeling furniture using natural language as a high level description for different furniture parts. Furthermore I show how to use these phrases for the interactive and procedural modeling of 3D furniture objects. Thereafter I will give an introduction outlining how to apply this system to real time augmented reality applications, e.g. in the furniture industry.

CCS Concepts
• Computing methodologies → Mesh geometry models; Mixed / augmented reality;

Keywords
procedural modeling, furniture modeling, natural language processing, augmented reality

1. INTRODUCTION
With the current development of computer technology, especially augmented reality and 3D graphics, the problem of content generation rises. With a demand of quicker product cycles, better and more interactive ways of product customization, the content generation becomes more and more complex. To aid with augmented reality, real-time content creation needs to be possible and easily usable. This can be achieved in using natural language as an input for the content generation system.

The novelty of this paper lies in the specific generation of 3D furniture objects using natural language, especially for augmented reality applications. The parametrization of the model generator is done using a natural language based approach where a parser parses the language into a top-down structure which is then converted into a tree-like scenegraph. Furthermore I will explain how to use this system in augmented reality applications using cameras with and without depth sensors. The presented framework will be called FURNZ.

Permission to make digital or hard copies of all or part of this work for personal or classroom use is granted without fee provided that copies are not made or distributed for profit or commercial advantage and that copies bear this notice and the full citation on the first page. Copyrights for components of this work owned by others than the author(s) must be honored. Abstracting with credit is permitted. To copy otherwise, or republish, to post on servers or to redistribute to lists, requires prior specific permission and/or a fee. Request permissions from permissions@acm.org.

© Copyright held by the owner/author(s). Publication rights licensed to ACM. ISBN.

DOI:
3.1 Introduction

Furniture can be distinguished into a set of four categories (see e.g., [Böö50]), furniture for seating (chairs etc.), furniture for storage (cupboards, wardrobes, etc.), furniture for lying (sofas, beds, etc.) and tables. Within those categories, the furniture have a more or less similar design that follows their function (e.g. one would not find a vertical bed very comfortable) and the furniture parts themselves can be described in a tree-like structure (see [Bei15]).

This concept can be used to describe a basic set of rules to generate a certain furniture object in these four categories - optional features would be doors or a back wall - and can be described procedurally. The word ”procedurally” has to be taken quite literally here, as the base parts are modeled using classes as known from object oriented programming. These classes represent methods (”procedures”) which generate certain furniture objects based on given attributes. Using standard OOP-concepts, complex objects can be derived from simpler classes. For example, a base class can describe a common rack. Attributes will be width, height, depth and number of shelves. The individual shelf height is then calculated based on the rack’s total height and the shelves distributed equally. One derived class might allow to parametrize the individual shelf height, another derived class could add an even number of doors.

The base parts are highly attributed and linked together in a tree-like structure. For example, a rack consists of at least one shelf (attribute: width, depth, thickness), up to two side walls (height, thickness, depth depends on shelf), zero or one back wall, zero or one stand. A chair on the other hand can be described by zero to several stands, a seating area and zero or two armrests and so on.

3.2 Natural Language

3.2.1 User Studies

To find a proper system how people use natural language to describe furniture, two user studies were conducted. In the first study, the test candidates were shown eleven different pictures of randomly chosen furniture (Bookshelf, Chair, Bed, Armchair, Table, Rack, Bed, Rack, Table, Office chair, Couch). The task was to then describe the shown furniture objects using short notes or sentences.

The average age of the 17 participants was 28.2 years. Three male and 14 female participants with no background in computer science took part in the study. The participants used an average of 3.4 notes per furniture object to describe it (with \( \sigma = 1.46 \)).

One thesis statement which was to investigate here is, that people tend to use a top-down approach, when describing furniture objects, e.g. first name the furniture category, then describe main features, then details. 82% of the cases (\( \sigma = 0.38 \)).

After an in-depth analysis, the general structure which was chosen by the participants to describe a furniture object was [furniture category] followed by [material], [main feature] followed by [details]. In 46% of the cases, the material was used after the main feature was described, so these two points should be considered interchangeable. This analysis helps to build a general structure for the grammar parser and information on how the user is supposed to enter the natural language description into the system. The “knowledge” here is actually based on how a majority of people would describe a furniture object, so the user does not have to adjust (much) when using the system.

Another interesting fact is, that users tend to describe the furniture in a more or less constant amount of notes. Although the difference between the participants is quite big (between one and eight notes for the same object) with \( \sigma \) for each individual object ranging between 1.8 and 2.0, the individual changes, how many notes one participant used to describe different furniture is rather small with \( \sigma \) between 0.3 and 1.1.

The second user study used a completely different method. The group of 18 participants (14 female and 4 male), all different from the participants taking part in the first study, with an average age of 22.1 years, were given the task to pick two items from the following list:

- furniture for sitting
- furniture for lying
- furniture to put things onto
- wardrobes, cupboards, closets, locker

After having chosen their two items, they were given the task to describe a piece of furniture in a way that a designer could draw it. It should be noted, that the study was conducted in the German language and that all the words in list item 4 (wardrobes, cupboards, closets, locker) are described with one German word - “Schräne”.

Category 1 was chosen 8 times, category 2 was chosen 12 times, category 3 and category 4 were chosen 8 times. In general, the participants used more notes (7.47 with \( \sigma = 2.59 \), \( min = 2 \) notes, \( max = 12 \) notes) to describe the individual furniture objects than in the first study.

In this study, 83% used a top-down approach, meaning the participants first named the subtype of furniture (e.g. a chair, a bed, a wardrobe) and then went further into detail. 77% (14 out of 18) used to name the material and color (for
the parsing of the material part. I refer to [Bun16]) immediately after they named the subcategory. The others named the main feature or features first. After this, the structure of the furniture was described (56%) from a coarse to a more detailed view usually (80% of the 56%) starting with the total measurements of the object. Interestingly, none of the participants in the first study provided any kind of measurement, not even adjectives like "big", "small" etc. After this, the details were described (ornaments, decorations, special designs like painted pictures on the object, etc.)

3.2.2 A linguistic model

The main goal of the user studies was to find a linguistic model which people generally use to describe furniture in day-to-day life and which will found a base for the language parsing. The studies showed that most people (over 80%, 82% in the first study, 83%) use a top-down approach to describe the structure of the object. Top-down in this case describes the way to describe the object from the general structure to finer details and naming the type of furniture (the subcategory, according to [Bö65]) first. To ease the software development, especially the language parser, in the current software version the model presented here has been used.

As mentioned in the last chapter, the structure found for describing furniture is:

- [type/subcategory]
- [main feature]
- [material]
- [measurements]
- [sub-features]
- [visual details]

where everything but the type/subcategory is optional. This allows for an efficient parsing of the linguistic input into a top-down structure. An example would be: "A rack with five shelves, made from oak wood, 2 meters tall, 80 centimeters wide, with a back wall" which then will be parsed into an object (in the sense of object-oriented programming) representing the given sentence and its attributes.

- Rack
- 5 shelves
- back wall present
- Size: 2.1m × 0.8m
- Material: oak wood

Automatically set attributes
- board thickness: 2cm
- shelf height 33cm
- no feet
- ...

3.3 Language Processing

The basic vocabulary was found by analyzing corresponding literature (e.g. citeboeth05) and business databases gratefully provided by some well known German furniture companies. This gives a good starting point as the vocabulary is based on commonly used wording in the furniture industry. One of the main issues is that the vocabulary used in the furniture industry can be highly complex and different terms with the same meaning in every day language could be used. [Mil14] showed that the number of adjectives to describe (in his case) sound rises quickly and becomes quite diverse when taken from user experience or user interviews. This shows that the unrestricted usage of “every day language” is not feasible and that an approach using controlled natural language ([Kuh14]) which reduces the the vocabulary to a predefined subset should be used. After choosing a basic set of nouns, adjectives and adverbs, several services can be used to extend the vocabulary such as the Wortschatz (wortschatz.uni-leipzig.de). The Wortschatz provides a web interface for finding synonyms (related meaning) and antonyms (opposite meaning) in the German language.

The language processing part of the application is based on a self-developed part-of-speech tagger which uses a lexicon of the German language to annotate the provided words. While annotating the words, the word is transformed into its base form using the Wortschatz-Project (see Chapter 3.2) to allow easier processing.

In the second step, the inflection (especially if working with the German language) of the words needs to be analyzed and the base form has to be found. This is done using a simple baseform lexicon.

The biggest problem with the specification of attributes using natural language is "intent vs. interpretation". This means, that the user might intend something different with his input than the framework "interprets" by parsing and analyzing it. For example, when asked about the color "blue" people might think of colors ranging from "light blue" to "navy blue", all being valid shades of "blue". Since the earliest speech processing systems in computer graphics ([Bol80]), this remains a more or less unresolved issue, as different people have a different personal definitions of the meaning behind descriptive words (e.g. adjectives like "blue", "tall", "slim"). To resolve this issue, the user can provide further modifiers (more, less, wider, taller, thinner; e.g. "more shelves", "thicker boards") in several refining steps. This frees the user from specifying discrete values for parameters, as this might contradict the intention of this system. Using the further refinements, the system still has enough potential to find the user’s intended attribute although specifying numerical values is also possible.

When the user now provides a natural language input, the input is parsed into an object representing the furniture. The attributes of the furniture the user does not provide are set to default values. As mentioned, the user then can refine the object by providing further input. For example, the simplest input could be "a rack". This would result in a small rack with five shelves using a standard material. The variable attributes for width, height, number of shelves, presence of a back wall, presence of foot stands, size of the stands etc can then be provided in a further refining step. Also, set attributes can be modified using adjectives such as "wider", "taller", "thicker" "thinner" etc. This leads to two different input methods. The first one would be to specify the whole object parameters in one sentence ("A wide rack made of oak wood with seven shelves on small foot stands without a back wall") or in several steps which refine the last given steps ("a rack" "wide" "oak wood" "seven shelves" "small foot stands" "no back wall"). Between each step, the user is presented with an image of the furniture after each step.

The parser will generate a tree-like structure of the furniture parts, setting position attributes where necessary (e.g. the two racks on the bench obviously need to have their base on top of the bench to avoid geometry overlaps in the ren-
dering). These position attributes will directly influence the transformation matrix of the resulting geometry node in the furniture’s scene graph.

4. PUTTING IT ALL TOGETHER

After extracting the attributes and parameters from the natural language input, the geometry generator needs to generate the specified 3D objects. The language parser parses the natural language input into an oop-style class. This class contains attributes for the specific furniture type. As an example, the class definition for racks is given.

For each furniture type, a base class is defined. This means, there is a class for chairs, beds, tables, armchairs, benches, sofas, etc. Each class provides attributes for the basic features of the desired furniture object, e.g., the Rack-class provides `num_shelves`, whereby a Wardrobe-class would provide attributes for drawers and hangers (int `num_drawers` and bool `hanger_present`). To create a new type of furniture, one just need to derive from either the abstract class `BasicFurniture` or, if the new type is a subtype of an already existent class, derive from this class. At the moment, this is implemented in C++ and needs to be compiled. The extension of the framework using cling (see [VCNR12]), a C++ interpreter built on top of Clang and the LLVM compiler infrastructure is currently in the planning stage.

The class also provides the methods to generate the furniture geometry data based on the set attributes. As mentioned, if an attribute is not specified by the user, a default value has to be used. To apply this on the basic rack example (see figure 2), the generator would look like:

To allow for more complex furniture (see chapter 3.3), the object will be represented in a scenegraph-like structure, where a node represents a basic furniture object (e.g., a rack) and contains the geometry (vertices, normals, texture coordinates, indices, etc) and the 4x4 transformation matrix of the object as well as material parameters.

5. AUGMENTED REALITY

The augmented reality part in the application is implemented using C++, OpenGL and ARToolkit. For the camera-only version, two models exist. One uses AR-Markers like the HIRO marker which will be placed on the floor. In the AR Application, this marker will be used to place the furniture.

6. RESULTS

The results show the furniture objects for “A small Rack with 4 shelves made of dark wood” and “A small table made of wood” rendered in an augmented reality application using ARToolKit5 and a HD Webcam (Microsoft LifeCam HD-3000). The rendering is in real time with 50 frames per second (the framerate is limited by the camera). The tracking is done using the Hiro-Pattern. The language processing takes less than a second on standard PC hardware (e.g., Quadcore Intel i7, 3GHz, nVidia Geforce GTX960).

The whole framework allows for the easy generation of (basic) furniture for augmented reality application, quickly helping users to get a first impression of how certain furniture pieces will look in a specific environment. The user does not need to interact with the system in any “complicated” way, the furniture is created simply by naming the type and some parameters. Combined with the language processing unit, the whole system is fast enough to be used in a real-time augmented reality environment.

Figure 1: “A small wooden table”

Figure 2: “A small Rack with 4 shelves made from wood” in an AR Application

A longer, more detailed and in-depth version of the paper with supplement material is available at https://buntin.de/research/

7. REFERENCES

