Numerical Methods for Beautification of Reverse Engineered Geometric Models

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# **Reverse Engineering**

- Engineering converts a concept into an artifact
- Reverse engineering converts an artifact into a concept
- Desired result is a representation of the design intent, not a simple copy
- Problem: Reverse engineered models suffer from inaccuracies caused by
  - ★ sensing errors during data acquisition
  - \* approximation and numerical errors from reconstruction algorithms
  - \* possible wear of the artifact
  - manufacturing method used to make the artifact

# **Beautification**

 Goal: Reconstruct an *ideal* model of a physical object with intended geometric regularities
 Only for engineering objects with planar, spherical, cylindrical, conical and toroidal surfaces (and blends)
 Previous approaches:

- Augment the surface fitting step by constraint solving methods [Fisher,Benkő]
- Manually identify features like slots and pockets and use them to drive the segmentation and surface fitting [Thompson]
- Our approach: Beautification, improve the model in a post-processing step

# **Beautification Strategy**

#### Analyser

Detect potential regularities which are approximately present in the initial model

#### Reconstruction

Reconstruct an improved model, fix topological prob- <br/>
lems, align model with coordinate axes, etc.

### Hypothesizer

Solve a constraint system derived from the regularities which describes a complete, improved model with likely regularities (only a subset of the constraints will be mutually consistent)

# **Beautification Strategy**

### Analyser

Detect potential regularities which are approximately present in the initial model

### Reconstruction

Reconstruct an improved model, fix topological problems, align model with coordinate axes, etc.

### Hypothesizer

#### **Constraint Selection**

Based on priorities and inconsistencies select a set of likely constraints

# $\downarrow$ $\uparrow$

#### **Constraint Solver**

Try to solve constraint system and indicate inconsistencies (solvability test)

# **Geometric Regularities**

### Directions

- Parallel directions
- Directions with same angle relative to a special direction
- Symmetrical arrangements of directions



#### Axes

- Axis intersections
- Aligned axes
   Parallel axes arranged equally spaced along lines, grids or on cylinders



### Positions

- Equal positions
- Positions equal under projection
- Positions arranged regularly on lines and grids

#### Parameter

- Equal lengths
- Equal angles
- Integer relations
- Special values:
  - \* integers
  - ★ simple fractions

#### Numerical Methods for Beautification

# **Constraint Selection & Solving**

Core of beautification problem:

constraint selection and solving

★ Regularities become sets of geometric constraints

- Model topology is described by geometric constraints
- Algorithm has to select desirable constraints such that the constraint system has a solution
- Finding a solution is a secondary task

## **Geometric Constraints and Objects**

Geometric constraints used to express regularities:

- $d_1^t d_2 = \cos(\alpha)$  Const./var. angle between directions  $d^{t}d = 1$ Normalize direction
  - $\mathbf{p} = \frac{1}{n} \sum_{k=1}^{n} \mathbf{p}_{k}$  Average position  $\sum_{\mathbf{k}} \alpha_{\mathbf{k}} \mathbf{s}_{\mathbf{k}} = \sigma$  $\mathbf{p} \in \mathbf{O}$
  - $\|\mathbf{p_1} \mathbf{p_2}\| = \lambda$  Const./var. distance between positions
    - Linear relation between scalars
    - Vertex on geometric object

Geometric object types:

★ plane, sphere, cone, cylinder, torus, circular ellipsoid

★ straight line, circle, ellipse

**\*** vertex

 $\star$  direction, angle, length

# **Selection & Solving Strategy**

I. Prioritize the regularities based on

- how well they are satisfied in the initial model
- how common and desirable the related regularity is
- II. Select initial subset  ${\bf S}$  from all detected regularities
  - Resolve simple inconsistencies using selection rules
  - Favour regularities with high priorities
- III. In order of highest to lowest priority add regularities c from S to a constraint system C
  - Try to solve constraint system C with new regularity c
  - Add c to C if extended system is solvable
  - Otherwise adjust S using the selection rules from II as c is not used

# **Prioritizing Regularities**

 Priorities determine which regularity to choose in case of an inconsistency

 Computed as weighted average with values in [0,1]: w<sub>c</sub>(r)(c<sub>e</sub>w<sub>e</sub>(r) + c<sub>q</sub>w<sub>q</sub>(r) + c<sub>b</sub>w<sub>b</sub>(r))
 \* with merit functions in [0,1]

 $w_e(r)$  numerical accuracy of regularity

 $\mathbf{w}_{\mathbf{c}}(\mathbf{r})$  constant indicating how common the regularity type is

 $\mathbf{w}_q(\mathbf{r}) \quad \mbox{specific quality/desirability depending on type} \\ \mbox{and involved arrangements and constants}$ 

 $w_b(r)$  constant minimum desirability

 $\star$  and weighting constants  $\mathbf{c}_{e}, \mathbf{c}_{q}, \mathbf{c}_{b}$ 

# **Quality Factors**

- Specific quality  $\mathbf{w}_q(\mathbf{r})$  of a regularity is determined by factors like
  - Special values involved (fractions with small integers preferred)
  - Number of involved B-rep elements with common boundary
  - \* Number of occupied positions for symmetrical arrangements on grids, circles, cylinders, ...
  - \* Number of B-rep elements of same geometric type
- $\mathbf{w}_{q}(\mathbf{r})$  is weighted average of merit functions for these factors

## **Selection Rules**

- Selection rules to resolve simple inconsistencies between constraints:
  - Constraints between the same objects with different constants
  - Different constraints between objects which are identified by coincidence constraints

### Generic selection rule:

A rule  $(n_1,R_1,n_2,R_2)$  is violated if at least  $n_k+1$  constraints in the set  $R_k$  are selected for each  $R_k \neq \emptyset$ 

- One interpretation:
  - If at least  $n_1 + 1$  elements of  $R_1$  are selected, then at most  $n_2$  elements of  $R_2$  should be selected

# **Initial Selection**

Initially all constraints are selected

- Consecutively enforce each rule such that a maximal set of constraints with largest priorities are selected
- Basic concept for rule enforcement algorithm:
  - If a rule is violated deselect sufficient constraints with lowest priority
  - Check if these deselections allow the selection of previously deselected constraints by checking previously enforced rules

Regularities A, B, C, D with w(A) < w(B) < w(C) < w(D)

A B C D 3 rules enforced in sequence:
• • • • Initially all regularities selected

●: regularity selected, □: regulairty deselected

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Regularities  $\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D}$  with  $\mathbf{w}(\mathbf{A}) {<} \mathbf{w}(\mathbf{B}) {<} \mathbf{w}(\mathbf{C}) {<} \mathbf{w}(\mathbf{D})$ 

**ABCD** 3 rules enforced in sequence:

- 1:  $\mathbf{R_1} = \emptyset$ ,  $\mathbf{n_1} = \mathbf{0}$ ,  $\mathbf{R_2} = {\mathbf{A, B}}$ ,  $\mathbf{n_2} = \mathbf{1}$ At most either A or B, not both, can be selected
  - $\rightarrow$  Deselect A due to lower priority

•: regularity selected,  $\Box$ : regulairty deselected

Regularities  $\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D}$  with  $\mathbf{w}(\mathbf{A}) {<} \mathbf{w}(\mathbf{B}) {<} \mathbf{w}(\mathbf{C}) {<} \mathbf{w}(\mathbf{D})$ 

**A B C D** 3 rules enforced in sequence:

1:  $\mathbf{R_1} = \emptyset, \mathbf{n_1} = \mathbf{0}, \mathbf{R_2} = \{\mathbf{A}, \mathbf{B}\}, \mathbf{n_2} = \mathbf{1}$ 2:  $\mathbf{R_1} = \{\mathbf{B}\}, \mathbf{n_1} = \mathbf{0}, \mathbf{R_2} = \{\mathbf{C}, \mathbf{D}\}, \mathbf{n_2} = \mathbf{1}$ If B is selected, then at most either C or D, not both, can be selected  $\rightarrow$  Deselect B due to lower priority

●: regularity selected, □: regulairty deselected

Regularities  $\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D}$  with  $\mathbf{w}(\mathbf{A}) {<} \mathbf{w}(\mathbf{B}) {<} \mathbf{w}(\mathbf{C}) {<} \mathbf{w}(\mathbf{D})$ 

**A B C D** 3 rules enforced in sequence:

●: regularity selected, □: regulairty deselected

Regularities  $\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D}$  with  $\mathbf{w}(\mathbf{A}) {<} \mathbf{w}(\mathbf{B}) {<} \mathbf{w}(\mathbf{C}) {<} \mathbf{w}(\mathbf{D})$ 

**A B C D** 3 rules enforced in sequence:

 $\bullet \quad \bullet \quad \bullet \quad \bullet$ 1:  $\mathbf{R_1} = \emptyset, \mathbf{n_1} = \mathbf{0}, \mathbf{R_2} = \{\mathbf{A}, \mathbf{B}\}, \mathbf{n_2} = \mathbf{1}$ 2:  $\mathbf{R}_1 = {\mathbf{B}}, \mathbf{n}_1 = \mathbf{0}, \mathbf{R}_2 = {\mathbf{C}, \mathbf{D}}, \mathbf{n}_2 = \mathbf{1}$  $\leftarrow$  B modified, so recheck rule 1 • 3:  $R_1 = \{B, D\}, n_1 = 0, R_2 = \{A, C\}, n_2 = 1$ If B or D are selected, then at most either A or C, not both, can be selected  $\rightarrow$  Deselect A due to lower priority

●: regularity selected, □: regulairty deselected

Regularities  $\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D}$  with  $\mathbf{w}(\mathbf{A}) {<} \mathbf{w}(\mathbf{B}) {<} \mathbf{w}(\mathbf{C}) {<} \mathbf{w}(\mathbf{D})$ 

**A B C D** 3 rules enforced in sequence:

3:  $\mathbf{R_1} = \{\mathbf{B}, \mathbf{D}\}, \mathbf{n_1} = \mathbf{0}, \mathbf{R_2} = \{\mathbf{A}, \mathbf{C}\}, \mathbf{n_2} = \mathbf{1}$ 

 $\leftarrow \mathbf{A} \text{ modified, so recheck rule 1}$ 

- $\rightarrow$  Cannot reselect  ${\bf B}$  due to rule 2
- $\rightarrow$  Final selection

●: regularity selected, □: regulairty deselected

Numerical Methods for Beautification

# **Numerical Solvability Test**

- After initial selection each regularity is consecutively added to a constraint system in order of priority
- Each time the constraint system has to be checked for solvability:
  - ★ System is solved as numerical minimization problem (numerically stable BFGS, BFGS/Gauss-Newton hybrid, ...)
  - $\star$  System is solvable if least-squares error is  $\sim 0$
  - ★ Guarantees solvability up to numerical tolerance
- If system is solvable, the regularity is accepted
- Otherwise (conditional) selection rules involving the deselected regularity are used to check if alternative regularities can be selected

# Examples



Detected Regularities	89	382	216
After initial selection	34	263	156
Finally selected	23	117	93
Time	$\sim$ 2h	$\sim$ 25h	$\sim$ 23h

#### Numerical Methods for Beautification

# Results

Major regularities can be enforced on improved model
 Symmetrically arranged face groups
 Orthogonal/parallel directions, ...

- Some problems due to ambiguities remain
  - Choice of special values for lengths and angles not always consistent
  - There is a choice between regularities of high quality and regularities with small errors
- Slow computation
  - Most of the time spent on solving constraint systems (one constraint system solved for each regularity)
     Only proof of concept (see payt clide)
  - Only proof of concept (see next slide)

# **Solvability Test Improvements**

- To speed up algorithm: faster solvability test
- Detect structural inconsistencies without solving the constraint system
- Use degree-of-freedom analysis to build up a consistent constraint graph
- Solve the constraint system numerically only once
- Currently under investigation:
  - \* Speeds up solvability test from hours to minutes
  - Further work required to ensure selection of consistent system

# **Selection Improvements**

• To improve quality of models: intelligent selection

- Ambiguities between approximate regularities cause inconsistent selection of regularities with respect to design intent
  - ★ E.g. do we have a cube with edge lengths 2 or a block with edge lengths 2 and 2.1, or ... ?
  - \* Local decisions can favour the block arrangements
- Make decisions in the context of the whole model, not locally with respect to inconsistencies
- Reduce/replace user-defined constants for priorities by simpler methods based on multiple-choice questions and learning

# Conclusion

- We have given a general approach to improve inaccurately reverse engineered geometric models
- The current numerical method improves models, but
  - Running time is too long
  - Regularity selection not consistent with respect to design intent
- Current work is addressing these problems