The Information is in the Maps: Representations & Algorithms for Mapping among Geometric Data

Leonidas Guibas
LELAND STANFORD JUNIOR UNIV CA

09/30/2015
Final Report

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The Information is in the Maps: Representations and Algorithms for Mappings among Geometric Data
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12. **SUPPLEMENTARY NOTES.** Enter information not included elsewhere such as: prepared in cooperation with; translation of; report supersedes; old edition number, etc.

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Final Progress Report, 09/01/2014 to 08/31/2015 for AFOSR FA9550-12-1-0372, “The Information Is In the Maps: Representations & Algorithms for Mappings among Geometric Data”

Final Executive Summary

The goal of this effort has been to develop a set of mathematical and computational tools for describing, analyzing, computing and exploiting relationships and mappings between geometric data sets, both pairwise and in higher-order combinations, or in loose networks of interrelated sets. The objective is to analyze geometric data sets jointly, organizing data collections into (possibly overlapping) groups of related sets or parts thereof, separating what is common from what is variable within each group and across groups, and understanding the main axes of variability.

The basic thesis of the work has been that geometric data sets are best understood not in isolation but within a “social network” of related data sets and their associated maps and correspondences. These relational networks can interconnect data sets into societies where the “wisdom of the collection” can be exploited in performing operations on individual data sets better, or in further assessing relationships between them. By creating such societies of data sets and their associations in a globally consistent way, we enable a certain joint understanding that provides the powers of abstraction, analogy, compression, error correction, and summarization over the data.

For example, given a collection of images with shared content across a number of object categories (e.g., airplanes), our network analysis techniques are able to learn the shared categories and discover the object(s) in each category contained in each image. Furthermore, this is accomplished in a fully unsupervised manner and the results surpass some state of the art methods that use supervision. Of course supervision can be added to further improve the results.

Of particular interest this past year has been algorithms for relating and interconnecting diverse modalities that provide information about objects in the world, including images, sketches, 3D scans, 3D models, and language. Different modalities often capture distinct types of information about the nature and state of objects in the world, so that the information fusion made possible by this integration creates new integrated knowledge not available separately from any of the modalities.

Unlike traditional data fusion, in our setting fusion is possible not only at the level of object instances but also across object categories, through the abstraction mechanisms we have developed in our networks. In particular, we have aimed to provide additional information or knowledge about captured signals (e.g., images), in a real time setting -- information that is
NOT present in the raw signal but is inferred from contextual knowledge present in the network. For example, when we see a chair partially occluded by a table, we can usually make a pretty good guess about what the occluded portions looks like, because we may see other identical chairs in the same environment, or because we have memory of having seen other similar chairs in analogous settings in the past. We have aimed to endow computers with exactly this ability to “imagine the unseen” and have made substantial progress on this front.

Specific 3rd Year Accomplishments/Findings

I. Common Embedding Spaces for Multimodal Data, Combining Visual Representations and Language

We have developed a new method for structuring multi-modal representations of shapes according to semantic relations. We learn a metric that links semantically similar objects represented in different modalities. First, 3D-shapes are associated with textual labels by learning how textual attributes are related to the observed geometry. Correlations between similar labels are captured by simultaneously embedding labels and shape descriptors into a common latent space in which an inner product corresponds to similarity. The mapping is learned robustly, by optimizing a rank-based loss function under a sparseness prior for the spectrum of the matrix of all classifiers. Second, we extend this framework to relating multi-modal representations of the geometric objects.

The key idea is that weak cues from shared human labels are sufficient to obtain a consistent embedding of related objects even though their representations are not directly comparable. Technically, we accomplish the assignment of labels to 3D geometry by learning a low-rank classifier matrix that recognizes similarities of labels through correlations in shape. This permits information sharing across geometrically similar objects as well as semantically related labels. In experiments, we can clearly see an advantage in performance over baseline methods that ignore this side information. Moreover, we have generalized the idea of multi-label classification through a low-dimensional latent space to obtain a novel cross-modal embedding of objects. This can be used for object retrieval across different modalities and for interactive explorations of complex data spaces.

We have evaluated our method against common base-line approaches, investigated the influence of different geometric descriptors, and demonstrated a prototypical multi-modal browser that relates 3D-objects with text, photographs, and 2D line sketches.

This work has appeared at the 2015 Eurographics Symposium on Geometry Processing.

II. The ShapeNet Repository

In order to develop and test our shape mapping algorithms at scale, we have initiated an effort to collect and annotate a large corpus of 3D CAD models that we call ShapeNet (http://shapenet.cs.stanford.edu). This is a joint effort with Professors Pat Hanrahan and Silvio
Savarese at Stanford, and Tom Funkhouser and Jianxiong Xiao at Princeton. The repository contains 3D models from a multitude of semantic categories and organizes them under the WordNet taxonomy. In addition to categories, ShapeNet currently provides consistent rigid alignments and bilateral symmetry planes for each 3D model. These annotations, as well as other planned semantic annotations, are made available through a public web-based interface to enable data visualization of object attributes, promote data-driven geometric analysis, and provide a large-scale quantitative benchmark for research in computer graphics and vision. Planned annotations include object parts and part names, local as well as global symmetries, physical properties such as size and weight, and affordances / functionality (how the shape is used). Maps and correspondences between shapes will be included, as well as between shapes and images.

ShapeNet aims to fill a large gap in the 3D repositories that currently exist, which are either large (e.g., the Trimble 3D Warehouse, 2.5M shapes) but poorly annotated, or annotated but small (e.g., the Princeton Shape Benchmark, 1.8K shapes). Recently large data sets of images, such as Imagenet (Deng et al. 2009, 14M images organized into 20K categories associated with “synsets” of WordNet) have played a major role in key data-driven advances in computer vision, in part by providing rich training data for machine learning algorithms. The same has occurred with natural language processing (NLP – e.g., in machine translation) and our belief is that similar breakthroughs can happen with 3D data.

This is a seed effort intend to lead to a separately supported project.

III. Shape Completion for Incomplete 3D Scans

Acquiring 3D geometry of an object is a tedious and time-consuming task, typically requiring scanning the surface from multiple viewpoints. In work this past year we focused on reconstructing complete geometry from a single scan acquired with a low-quality consumer-level scanning device, even in the presence of significant occlusions (and of course self-occlusions). Our method is class-based and uses a network of example 3D shapes to build structural part-based priors that are necessary to complete shapes in that class. In our representation, we associate a local coordinate system to each part and then learn the distribution of positions and orientations of all the other parts from the network, which implicitly also defines the positions of symmetry planes and symmetry axes. At the inference stage, this knowledge can be transported to the new scan and used to analyze incomplete point clouds with substantial occlusions, because observing only a few regions is still sufficient to infer the global structure. Once the parts and the symmetries are estimated, both data sources, symmetry and database, are fused to complete the point cloud, providing much better results than either of them alone could.

Our main technical contribution is a data-driven technique for estimating shape structure from incomplete point clouds. The key difference from previous approaches is that our method does not rely on a global coordinate system — instead every part defines local coordinates, and then all parts are jointly optimized to find their most plausible arrangement. This enables the
prediction of parts in occluded regions, and the estimation of symmetries even if the input partial scan is asymmetric due to occlusions.

We have evaluated our technique on a synthetic dataset containing 481 shapes, and on real scans acquired with a Kinect scanner. Our method demonstrates high accuracy for the estimated part structure and detected symmetries, enabling higher quality shape completions in comparison to all alternative techniques. Furthermore, we have publically released our benchmark data set so that others working in this area can evaluate their methods and compare them to ours.

This work, an instance of “imagining the unseen,” will appear at Siggraph Asia 2015.

IV. 3D-Assisted Feature Synthesis for Novel Views

We are especially interested in being able to link images that represent views of the same object from very different viewpoints. Comparing different views has been a long-standing challenging problem in computer vision, as visual features are not stable under large viewpoint changes. In work this past year, given a single input image of an object, we have developed tools able to synthesize its features for other views, leveraging an existing, modestly-sized 3D model collection of related but not identical objects. 3D models can provide strong prior information to help an algorithm “imagine” what the underlying 3D object should look like from novel views. To accomplish this feature transport to new views, we study the relationship of image patches between different views of the same 3D object, seeking what we call “surrogate patches” — patches in one view whose feature content predicts well the features of a patch in another view. These surrogate relationships are learned from the analysis of a co-aligned set of 3D models in a given class. Note that, indirectly, these surrogate patch relationships encode geometric global or local symmetries of the underlying 3D models, without having to first estimate the latter.

When an image of an object is provide, we first estimate its pose and then develop local linear models for predicting its features using features from the same views of our 3D models. We finally use our surrogate relationships for transferring the same linear combination to estimate features for the new view. Based upon these surrogate relationships, we can in fact create feature sets for all views of the latent object on a per patch basis, providing us an augmented view-independent representation of the object. We note that our method can work with many common image feature sets, including HoG, CNN, etc.

The method allows us to compare images of objects from very different views. In addition to demonstrating that we do much better in such cross view comparisons than traditional image-based methods, we have explored a number of other applications of our techniques. These include (a) part-based image retrieval, where we query for similar images within a specified region, aiming again at view independent results, (b) fine-grained image retrieval and object categorization, and (c) instance retrieval (looking for exactly the same object in other images).
This work, a second instance of “imaging the unseen,” will appear at ICCV 2015 (oral).

V. Analyzing and Using the Shape of White Matter Brain Structures

In recent years, a focus of the neuroscience community has been to understand the role of white matter in human cognition and function. We have developed a set of tools to study the 3D shape and shape-variability in major human white matter tracts. First, a mapping tool that extracts the skeleton of a tract and performs fine grained spatial clustering to identify correspondence between areas of tracts from different brains. This tool allows comparison of tissue properties on a fine-scale between individuals, mapping the 3D morphology of these structures across large human populations, thus exploring the available neuroimaging datasets in a principled manner. Second, a synthesis and simulation tool that applies a set of simple shape-deformations, such as bending and shearing, on a given tract. This enables parametrization of the underlying shape space and analysis of inter-individual normal and pathological shape variability. This work is currently ongoing. We have tested the fine-mapping tool on a cohort of subjects from the ADNI dataset and showed that we’re able to accurately map corresponding areas on a variety of tracts across individuals.

Refereed Publications (past year only)


Personnel Supported (past year only)

1. Leonidas Guibas, Faculty PI, 1.00 month
2. Peter Huang, Postdoctoral Fellow, 0.30 month [Shapenet repository]
3. Vladimir Kim, Postdoctoral Fellow, 7.25 months [shape completion]
4. Tany Glozman, Graduate Student, 1.50 months [white matter brain structures]

A Stanford undergraduate student (Ivan Robles) also worked with us his summer on topics related to this grant.
Education

We are planning a new course, CS233: The Shape of Data – Geometric and Topological Data Analysis, based to a significant part on material developed under this grant. This course will be taught at Stanford in the spring of 2016.

Interactions/Transitions

The work described this year’s and in previous reports has been or will be presented in the very top venues in computer vision, computer graphics and machine learning. Earlier work was also presented at the AFOSR Computational Cognition annual meetings.

Several companies have expressed strong interest in our work. We are currently collaborating on topics related to this grant with Adobe, Apple, Autodesk, and Google.

PI Honors/Awards

- 2013 Eurographics Symposium on Geometry Processing, Best Paper Award
- 2013 International Conference on Computer Vision Helmholtz Award (recognizes ICCV papers from ten years ago with significant impact on computer vision research)
1. Report Type

Final Report

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Organization / Institution name

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Grant/Contract Title
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The Information Is In the Maps: Representations & Algorithms for Mappings among Geometric Data

Grant/Contract Number
AFOSR assigned control number. It must begin with "FA9550" or "F49620" or "FA2386".

FA9550-12-1-0372

Principal Investigator Name
The full name of the principal investigator on the grant or contract.

Leonidas J. Guibas

Program Manager
The AFOSR Program Manager currently assigned to the award

James H. Lawton

Reporting Period Start Date
09/01/2014

Reporting Period End Date
08/31/2015

Abstract

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Specifically, we have shown how features for new views of an object can be computed from a single image, and how missing portions of a single 3D scan of an object can be completed -- in both cases exploiting network knowledge.

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We have also demonstrated how to do compute joint cross-modal embeddings, including images, sketches, 3D models, and words -- and use them for retrieval across different modalities.

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**Changes in research objectives (if any):**

**Change in AFOSR Program Manager, if any:**

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**AFOSR LRIR Number**

**LRIR Title**

**Reporting Period**

**Laboratory Task Manager**

**Program Officer**

**Research Objectives**

**Technical Summary**

**Funding Summary by Cost Category (by FY, $K)**

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