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AMUSE:

Acquiring and Modeling Urban Simulation Environments

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Spring 2006

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Motivation

- Visualization of large urban environments is a great challenge for computer graphics
 - An urban environment is a collection of buildings and roads spanning a large area of land and arranged into neighborhoods, blocks, and parcels

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Motivation


- Lafayette, Indiana



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Motivation

- Indianapolis, Indiana



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Motivation


- Beijing, China



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Motivation

- Rome, Italy



Motivation



- Paris, France



Applications



- Urban Planning
 - What would a proposed neighborhood look like?
 - What would an area look like after population growth?
 - What would happen if we put a road here? (road planning)
 - Architectural designs wish to use common building blocks yet have unique and interesting spaces; can we extrapolate a city given a set of building blocks?

Applications



- Emergency Management
 - Can we create a model of a very large urban space to train emergency response personnel?
 - Can we plan evacuation routes and suggest emergency deployments?
 - Can we prioritize policing and resource deployment?
 - Given an urban model struck by a disaster/attack, can we deploy an emergency-relief communication network?
 - Can we deploy approximate structural information, via a PDA, to rescuers using both the urban model and the emergency-relief communication network?

Applications



- Reconnaissance and Rapid Prototyping
 - Can we rapidly build a prototype of an (enemy) location from aerial views?
 - Can we specify the most beneficial places from where to obtain ground views so as to build an urban model for soldier training and other simulations?
 - If an environment is changing, can we indicate from where we need the most updates?

Challenge



- Modeling large urban spaces typically requires significant manual effort, storage, and computation
- Dense urban areas are particularly difficult because they are both very complex and very widespread
 - Size of the environment makes obtaining detailed structural information prohibitive, leaving us with only sparse information

Observations



- Large urban environments exhibit significant repetition
 - Similar structures are repeated at the global level; however, they maintain individuality in local detail
- Widespread digital meta-data is available
 - High-resolution aerial views
 - e.g., 6 inch/pixel
 - City/county/parcel boundaries, road networks, basic building information
 - e.g., input to Google Maps

Approach

- Perform an inverse urban modeling task by inferring the 2D layout of an existing environment
 - Procedural methods have the advantage of exhibiting a high-degree of detail amplification, e.g. using a small number of parameters yields significant plausible details
- The resulting grammar allows us to
 - create modifications to the existing urban environments, in the style of the original
 - determine the most representative areas and layouts

Example: Changing Urban Spaces

Example: Changing Urban Spaces

Example: Changing Urban Spaces

Example: Simplifying Urban Spaces

Thousands of parcels/buildings Dozens of parcels/buildings

Related Work

- Forward-generating grammars (L-systems) for creating plants, cities, and buildings
 - Specify a grammar and few initial parameters, then "grow" the structure
 - × Not based on an actual real-world layout or city
- Photogrammetric Reconstruction and IBMR
 - Build a model from photographs (e.g., Façade, Lightfields, robot-based acquisition)
 - × Requires specialized acquisition systems and does not scale
- Inspiration
 - Epitomes and Vector Quantization
 - Build-by-Number
 - One paper: inferring plant L-system parameters from photographs

Terminology and Assumptions



- Parse
 - From aerial views and meta-data, create a set of production rules and a set of terminals
 - e.g., string to grammar
- Derive
 - Using the production rules, terminals, and a starting configuration, create an urban layout
 - e.g. grammar to string

Terminology and Assumptions



- Parcel
 - consists of a piece of bounded land; might contain building structures
- Block
 - Collection of adjacent parcels; interior boundaries are all imaginary; exterior boundary is a road; all parcels have access to road (egress rule)
- Neighborhood
 - Collection of blocks, separated by roads, and mostly of the same classification (e.g. "residential", "commercial", "industrial", "downtown", etc.)
- Region
 - Collection of neighborhoods, usually separated by major roads

Terminology and Assumptions



- Production
 - Given a region | neighborhood | block, partition by a road | boundary
 - Assumptions (for now): regions are convex polygons, partitions are polylines, production produces two children
- Terminal
 - Is a parcel
 - May or may not contain building contours

Methodology



- 1. Parsing
 - How to parse aerial views and their metadata
- 2. Terminal Simplification
 - Reducing the number of terminals
- 3. Production Simplification
 - Reducing the number of productions
- 4. Novel Derivations
 - Making new layouts

Methodology



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1. Parsing

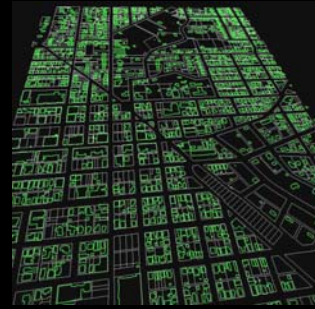


- Parse aerial views in a top-down fashion to produce a set of production rules for creating the urban layout
- Existence Question
 - Does such a grammar exist?
- Answer:
 - Yes! It is exactly one production rule for each partition and exactly one terminal for each parcel
- *The interesting work is in simplifying and compacting the grammar so that it can be used in a flexible fashion*

Example Aerial View I



Example Metadata I



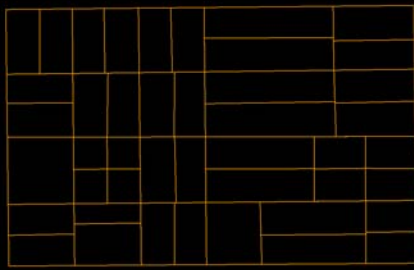
Example Aerial View II



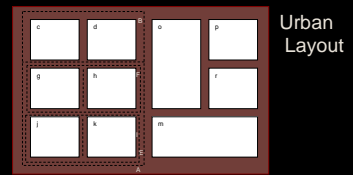
Example Metadata II



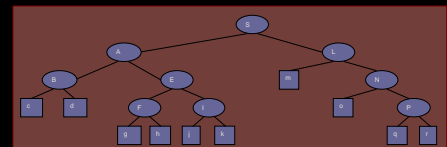
Example Metadata II



Parse Tree



Urban Layout



Parse Tree

Production Rules

$S \rightarrow AL$
 $A \rightarrow BE$
 $L \rightarrow mN$
 $B \rightarrow cd$
 $E \rightarrow FI$
 $F \rightarrow gh$
 $I \rightarrow jk$
 $N \rightarrow oP$
 $P \rightarrow qr$

} ~regions
 } ~neighborhoods
 } ~blocks

(lower case = terminals)
 (upper case = region|neighborhood|block)

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2. Terminal Simplification

- Reduce the number of terminals
 - Find a compact "dictionary" of urban structures
- Serves to
 - make grammar more compact
 - compress the data
 - e.g. image compression
 - prioritize the terminals
 - e.g. guides which parcels should be captured in more detail

Urban Dictionaries

Aerial data Dictionary Reconstruction

words

Simplification Pipeline

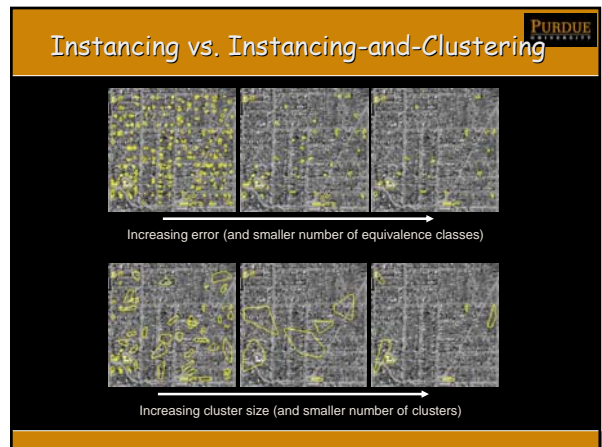
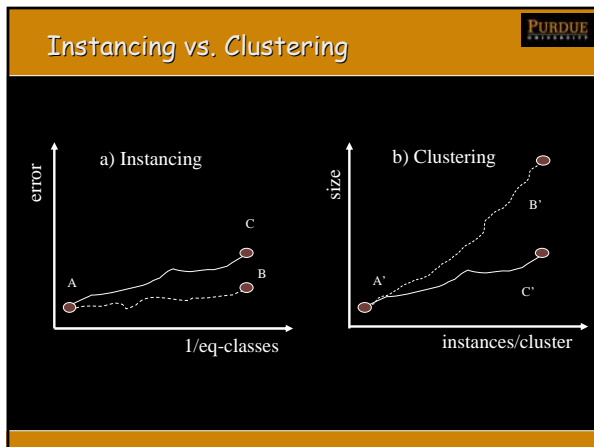
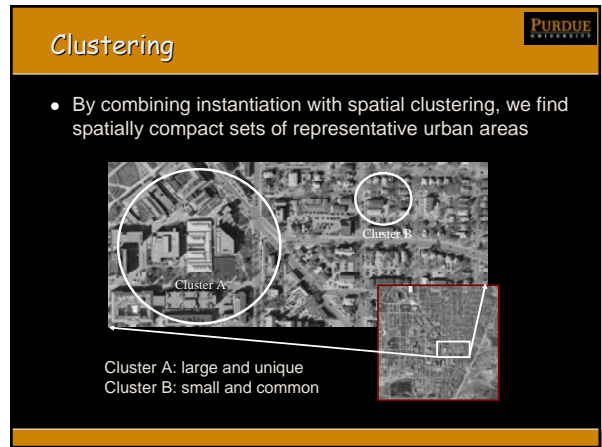
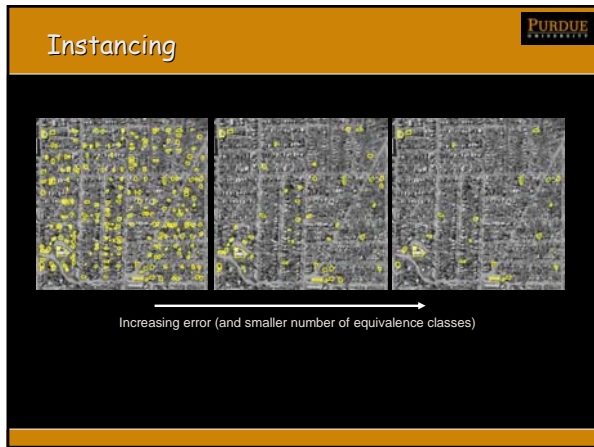
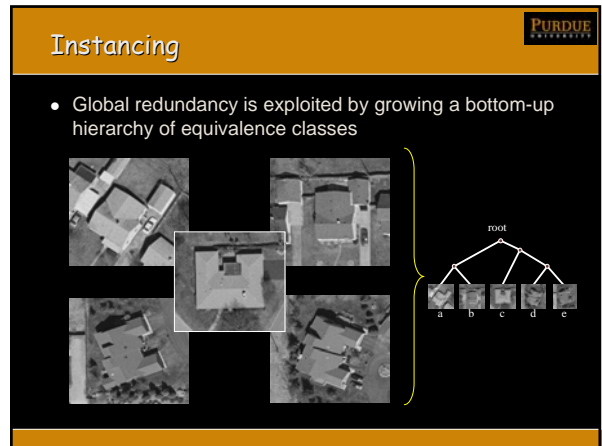
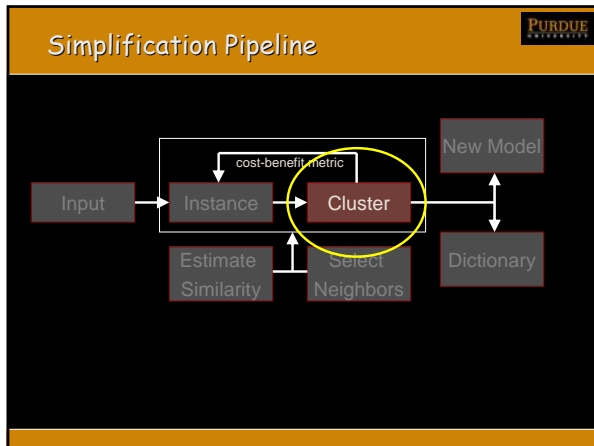
```

    graph LR
      Input --> Instance
      Instance --> Cluster
      Cluster --> NewModel[New Model]
      Cluster --> Dictionary
      NewModel -- cost-benefit metric --> Instance
      EstimateSimilarity1[Estimate Similarity] --> Instance
      SelectNeighbors1[Select Neighbors] --> Instance
      EstimateSimilarity2[Estimate Similarity] --> Cluster
      SelectNeighbors2[Select Neighbors] --> Cluster
  
```

Simplification Pipeline

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```



Instancing vs. Instancing-and-Clustering

Increasing error (and smaller number of equivalence classes)

Increasing cluster size (and smaller number of clusters)

Instancing and Clustering

Quality (left) and cost (right) for increasing amounts of instancing/clustering

Increasing error

Example Terminal Simplifications

Aerial images Meta-data Dictionary

Reconstruction Synthetic City

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3. Production Simplification

- Goal
 - Find a set of representative "rules" of the urban environment that can instantiate the same space, new spaces, and similar spaces
- Method
 - Find a dictionary of "rules" to build-up the urban space, for example:
 - Discover the rules by analyzing the layouts
 - Provide core rules that can represent all possible layouts

3. Production Simplification

- Rule Clustering
- Rule Canonization

3. Production Simplification

- Rule Clustering
- Rule Canonization

Recall the Original Production Rules...

- $S \rightarrow AL$
- $A \rightarrow BE$
- $L \rightarrow mN$
- $B \rightarrow cd$
- $E \rightarrow FI$
- $F \rightarrow gh$
- $I \rightarrow jk$
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- $P \rightarrow qr$

Full specification of the production rules

Alternate Rule Notation

- $S \rightarrow A_0 A_1^{180}$
- $A_0 \rightarrow B_0 E_0$
- $B_0 \rightarrow cc$
- $E_0 \rightarrow B_0 B_0$
- $A_1^{180} \rightarrow E_1 m$
- $E_1 \rightarrow m B_1^{90}$
- $B_1^{90} \rightarrow cc$

Shape = letter; position = subscript; rotation = superscript

Rule Shape Similarity

- $S \rightarrow AA$
- $A \rightarrow BE \mid mE$
- $B \rightarrow cc$
- $E \rightarrow BB \mid mB$

Rules grouped based on similar shape but *ignoring* differences in location and rotation

Rule Shape Similarity

- $S \rightarrow AA$
- $A \rightarrow BE \mid mE$
- $B \rightarrow cc$
- $E \rightarrow BB \mid mB$

= "Cluster by similar shape"

Rules grouped based on similar shape but *ignoring* differences in location and rotation

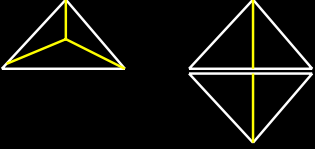
Rule Clustering

- Similarity(A,B) =
 - $W_1 \text{ShapeSim}(A,B) + W_2 \text{LocationSim}(a,B) + W_3 \text{PositionSim}(A,B) + W_4 \text{PartitionSim}(A,B) + W_5 \text{TypeSim}(A,B)$
- Production rule clustering:
 - Define each rule by a n-dimensional vector
 - Perform k-means clustering to obtain k clusters of rules
 - Choose a representative rule from each cluster
- Effectively "infer" the most popular/representative production rule styles

Note: ???

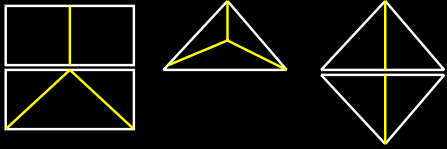
Rule Canonization

- Reduce the full grammar to a more compact one using only a small number of flexible yet powerful canonical production rules
- Consider triangles:




Rule Canonization

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- Consider rectangles:




Rule Canonization

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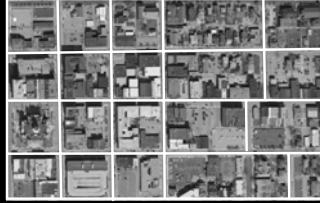
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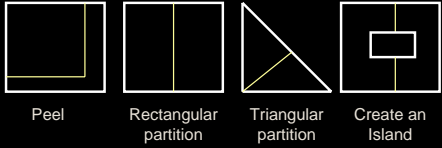
Rule Canonization

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Rule Canonization

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Peel Rectangular partition Triangular partition Create an Island Etc...

Note: ???

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Novel Derivations

- Given production rules and an initial structure, derive an urban layout
- Examples:
 - Change original urban space
 - "Move a road" in the original urban space
 - Fill a new region with an urban space similar to the original
 - Grow an urban area

Novel Derivations

- Need to support subset of affine transformations:
 - Translation (e.g., "move a region")
 - Rotation (e.g., "re-orient a neighborhood")
 - Scale (e.g., "stretch/squish a block")

Transformations

- Translation
 - Easy to handle
- Rotation
 - Easy to handle
- Scale
 - ???

Observation: Scaling/Stretching

H region V split H region H split V region V split V region V split

Conclusion:

- V stretches handled by H partition
- H stretches handled by V partition

Production Rules Revisited...

- $S \rightarrow^V AA$
- $A \rightarrow^H BE \mid mE$
- $B \rightarrow^V cc$
- $E \rightarrow^H BB$
- $E \rightarrow^V mB$

Example Derivations

- Original urban space
 - $S \rightarrow AA$
 - $S \rightarrow BEmE$
 - $S \rightarrow ccBBmmB$
 - $S \rightarrow cccccmmcc$
- Stretched urban space
 - $S \rightarrow AA$
 - $S \rightarrow BEmE$
 - $S \rightarrow BEBEmEmE$
 - $S \rightarrow ccBBccBBmmBmmB$
 - $S \rightarrow cccccccccccmmccmmcc$

Note: for simplicity, these derivations are showing 1D strings – data is really 2D

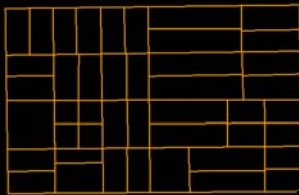
Original Urban Space I



Original Urban Space I



Original Urban Space I



Stretched Urban Space I

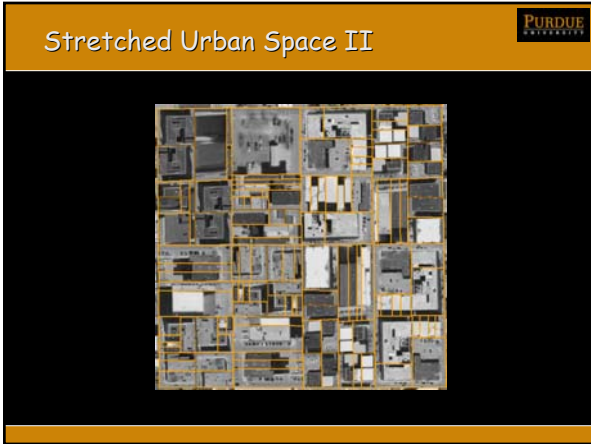
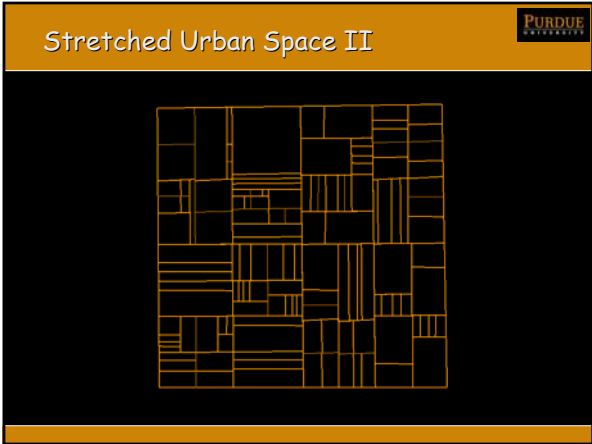
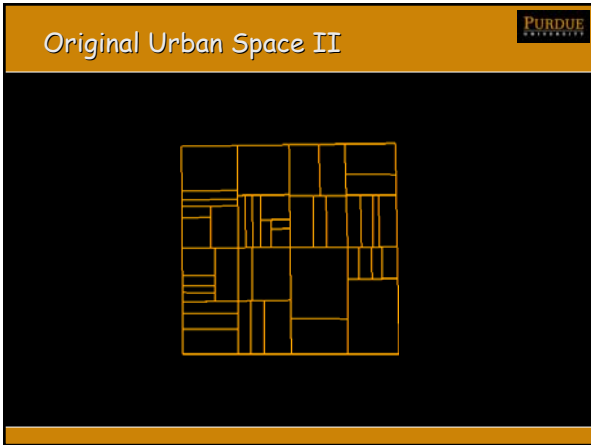
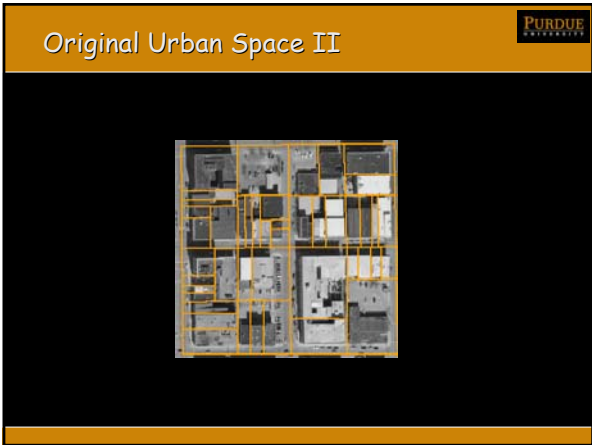
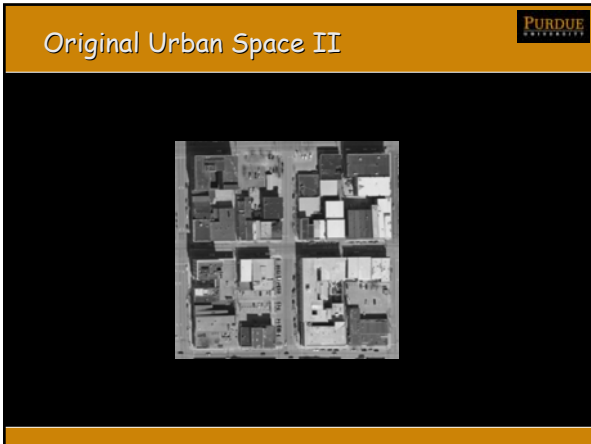
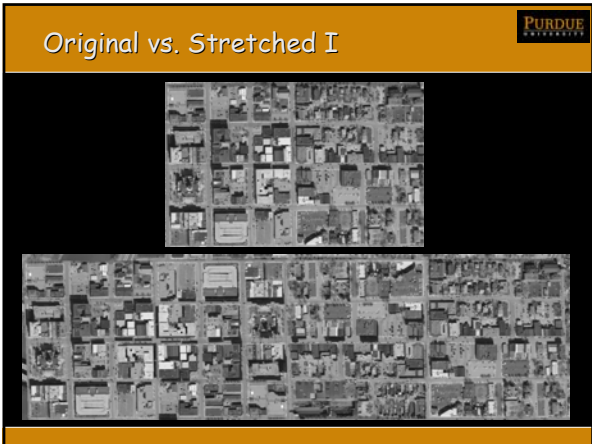


Stretched Urban Space I



Stretched Urban Space I

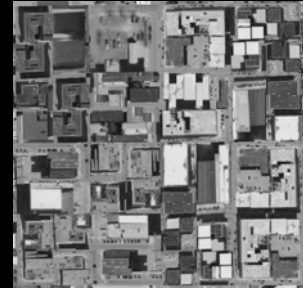




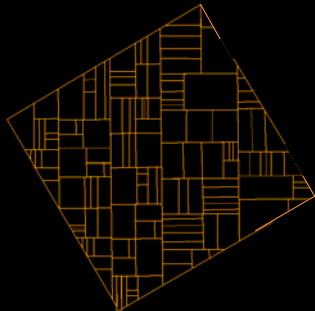
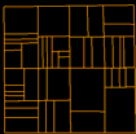
Stretched Urban Space II



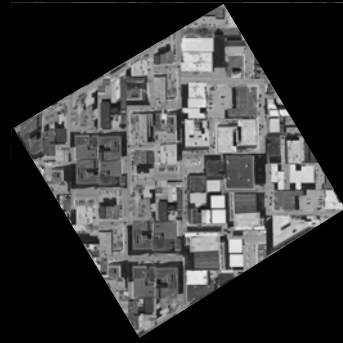
Original vs. Stretched II



Original vs. Stretched III



Original vs. Stretched III



Demo



Conclusions



- Urban Modeling is fun!
- Ability to be able to take views of an entire urban space and make an editable model out of it is very enticing
- Similar to fractal-based compression, the key is to find a good set of generators
 - Fortunately, urban environments offer significant amount of structure (and repetition) which we can exploit

Future Work



- Procedural Simplification
 - Specify canonical procedural rules?
 - Infer canonical procedural rules?
- Obtain data for a larger/more-interesting set of cities
 - Chicago, Rome, Paris, Cusco, etc.
- Full Inverse Modeling
 - Combine with Build-by-Numbers
- Applications
 - Road planning
 - Growth algorithms
 - Rapid prototyping

Thank you!

