Infrastructure Inventory Compilation Using Single High-Resolution Satellite Images

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Overview

- Objective
- Building Inventory Data Structure
- Why Building Inventory?
- Available Building Inventory Databases
- Why Remote Sensing?
- Proposed Algorithm
 - Sensor Models
 - Rational Function Model
 - Image Acquisition Geometry
 - 3D Reconstruction Algorithm
 - Measurement Error
- Implementation (MIHEA)
- Results and Calibration



Objective

To develop a semi-automated method for spatial and structural information extraction from single satellite images to be used in building inventories



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Building Inventory Data Structure

- Address or Lon/Lat
- Age or year of construction
- Height or number of stories
- Footprint Area
- Structural type
- Occupancy type
- Roof type
- Cladding

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- Shape irregularity
- Height irregularity

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Why Building Inventory?

Vital information to:

- Decision Makers
- Urban Planners
- Disaster Response
- Loss Estimation
- Portfolio management



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Available Inventory Databases

Sources:

- Tax Assessors files
- Government files (e.g. FEMA, GSA ...)
- Sanborn maps (US only)
- Real estate files
- Insurance portfolios

Problems:

- Not always available in digital format
- Incomplete for many attributes
- Unavailable for many regions (e.g. developing countries)
- Updating only some information



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Why Remote Sensing?

- Remote and hard to reach locations
- Digital formats
- Can augment missing information
- Large Coverage
- Frequent updates
- Eventually will be more cost effective than land survey

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Building Inventory Data Structure

- Address or Lon/Lat *
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- Cladding*
- Shape irregularity*
- Height irregularity*
- * information that can be obtained from imagery



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Sensor Models

Physical Sensor Model

- Sensor dependent
- Parameters have physical significant
- Parameters are statistically uncorrelated
- Rigorous and not always available to users

Generalized Sensor Model

Generic

- Sensor independent
- Real-time computation



Rational Function Models (RFMs)

 Describes the image-toground relationships.

 Generalization of polynomial models (ratio of two polynomial functions)

 $r_n = \frac{f_1(\phi_n, \lambda_n, h_n)}{f_2(\phi_n, \lambda_n, h_n)}$ $c_n = \frac{f_3(\phi_n, \lambda_n, h_n)}{f_4(\phi_n, \lambda_n, h_n)}$



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RFMs cont'd

Rational Polynomial Coefficients (RPCs)

$$f = \sum_{i=1}^{3} \sum_{j=1}^{3} \sum_{k=1}^{3} a_{ijk} \phi^{i} \lambda^{j} h^{k}$$

$$\begin{split} f &= a_1 + a_2 \lambda + a_3 \phi + a_4 h + a_5 \lambda \phi + \\ a_6 \lambda h + a_7 \phi h + a_8 \lambda^2 + a_9 \phi^2 + a_{10} h^2 + \\ a_{11} \phi \lambda h + a_{12} \lambda^3 + a_{13} \lambda \phi^2 + a_{14} \lambda h^2 + \\ a_{15} \lambda^2 \phi + a_{16} \phi^3 + a_{17} \phi h^2 + a_{18} \lambda^2 h + \\ a_{19} \phi^2 h + a_{20} h^3 \end{split}$$





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Image Acquisition Geometry

Approximate image acquisition geometry and satellite orientation can be described by sensor's <u>elevation</u> and <u>azimuth</u> angles



Height Metrology

- Image coordinates for the corner of a building at ground level (*r_{ground}*, *c_{ground}*), and its corresponding roof-point coordinates (*r_{roof}*, *c_{roof}*)
- Sensor's collection azimuth angle







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3D Reconstruction Algorithm

$r_{roof} = \frac{f_1(\phi, \lambda, h_2)}{f_2(\phi, \lambda, h_2)}$ $c_{roof} = \frac{f_3(\phi, \lambda, h_2)}{f_4(\phi, \lambda, h_2)}$ $r_{ground} = \frac{f_1(\phi, \lambda, h_1)}{f_2(\phi, \lambda, h_1)}$ H $c_{ground} = \frac{f_1(\phi, \lambda, h_1)}{f_4(\phi, \lambda, h_1)}$ H $c_{ground} = \frac{f_3(\phi, \lambda, h_1)}{f_4(\phi, \lambda, h_1)}$ $r_{ground} = \frac{f_3(\phi, \lambda, h_1)}{f_4(\phi, \lambda, h_1)}$ $r_{ground} = \frac{f_3(\phi, \lambda, h_1)}{f_4(\phi, \lambda, h_1)}$	
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3D Reconstruction Algorithm - cont'd

- Homogeneous
- Nonlinear

Im

Over-determined

Solution: Trust-Region Dogleg Method

 $\begin{aligned} \frac{f_1(\phi,\lambda,h_1)}{f_2(\phi,\lambda,h_1)} &- r_g = 0\\ \frac{f_3(\phi,\lambda,h_1)}{f_4(\phi,\lambda,h_1)} &- c_g = 0\\ \frac{f_1(\phi,\lambda,h_2)}{f_2(\phi,\lambda,h_2)} &- r_r = 0\\ \frac{f_3(\phi,\lambda,h_2)}{f_4(\phi,\lambda,h_2)} &- c_r = 0\\ (h_2 - h_1) - H = 0 \end{aligned}$

Starting point for the iterative solution: by linearizing RFM $a_1+a_2\lambda+a_3\theta+a_4\theta$

$r_{g} = \frac{a_{1} + a_{2}\lambda + a_{3}\varphi + a_{4}h_{1}}{b_{1} + b_{2}\lambda + b_{3}\phi + b_{4}h_{1}} + \varepsilon_{1}, c_{g} =$	$=\frac{c_1+c_2\lambda+c_3\varphi}{d_1+d_2\lambda+d_3\varphi}$	$\frac{1}{d_4h_1} + \varepsilon_2$				
$r_r = \frac{a_1 + a_2\lambda + a_3\phi + a_4h_2}{b_1 + b_2\lambda + b_3\phi + b_4h_2} + \varepsilon_3, c_r =$	$=\frac{c_1+c_2\lambda+c_3\phi}{d_1+d_2\lambda+d_3\phi}$	$\frac{+c_4h_2}{+d_4h_2} + \varepsilon_4$				
$A \cdot \mathbf{x_0} = b$	$A = \begin{bmatrix} r_g b_2 - a_2 \\ c_g d_2 - d_2 \\ r_r b_2 - a_2 \\ c_r d_2 - d_2 \end{bmatrix}$	$r_g b_3 - a_3$ $c_g d_3 - d_3$ $r_r b_3 - a_3$ $c_r d_3 - d_3$	$r_g b_3 - a_3$ $c_g d_3 - d_3$ 0 0	$\begin{bmatrix} 0\\0\\r_rb_3-a_3\\c_rd_3-d_3 \end{bmatrix}, \mathbf{x_0}$	$= egin{bmatrix} \phi^* \ \lambda^* \ h_1^* \ h_2^* \end{bmatrix}, \ b$	$= \begin{bmatrix} a_1 - r_g b_1 \\ c_1 - c_g d_1 \\ a_1 - r_r b_1 \\ c_1 - c_r d_1 \end{bmatrix}$



Measurement Error

- Image dependent (shadows and obstacles along the line of sight)
- User (operator) dependent



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MIHEA (Mono-Image Height Extraction Algorithm)

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Selection Tool

Object S

Top Point

Ground Poi

Roof Point: End Selectio

Ru

3D Di

Image Processing Package

 Compatible with IKONOS and QuickBird camera models

Export results to database

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Questions?



