Evaluation of Projection Errors using Commercial Satellite Imagery

Jim Bethel Purdue University, School of Civil Engineering Remote Sensing Seminar, 15 September, 2004 **Outline of Presentation** 

•Traditional Mapping Polynomial, Image Warping, Rubber Sheeting Approach

•Description of Photogrammetry Approach: Physical Model and Replacement Model

- •Example of Physical Model
- •Example of Replacement Model
- •Evaluation of Projection Errors Using Vendor Supplied Replacement Model
- •Conclusions

## Mapping Polynomials or Rubber Sheeting



 $r = a_0 + a_1 X + a_2 Y + a_3 XY + a_4 X^2 + a_5 Y^2$  $c = b_0 + b_1 X + b_2 Y + b_3 XY + b_4 X^2 + b_5 Y^2$ 



#### Image

For each point we create two equations. We need at least as many equations as unkowns. If more, then we use least squares. It is like a regression problem: linear, easy. But we are confounding the effects of sensor, platform motion, and terrain relief. What should be the order of the polynomial ?

## Graphical View of Rubber Sheet Transformation (2<sup>nd</sup> order, 12-parameter)



## Mapping Polynomials or Rubber Sheeting

If the terrain is flat, the sensor has narrow field of view, the sensor is nadir looking, and the ground sample distance is large, then *you can get reasonable results using the approach of mapping polynomials*.

The accompanying Quickbird image (0.61m pixel) shows the pitfalls of mapping polynomials when the above conditions do not apply. The two marked points have the same XY and they would get mapped into the same (row, col), but clearly that is wrong. You could expand the polynomial by adding some Z-terms. But that would not work. *Modeling the actual physical imaging process is the only way*.



## Photogrammetric Approach to Image Geometry

**Physical Sensor and Platform Models** 

•Sensor (Camera): aperture size, focal length, scanning elements, optical distortion

•Platform : orbit parameters: altitude, velocity, etc., orientation / attitude, rates and accelerations



Use physical model **Directly**, vendor or manufacturer defines the generic model: equations and constants. For a particular image, numerical values come from

•Vendor: support data supplied with image, metadata, ephemeris data, etc.

•User: obtain numerical values using ground control points

Advantage: parameters have physical meaning, flexibility

**Disadvantage:** vendor may not want to share, each one different Use physical model **Indirectly** via Rational Polynomial Coefficients (Replacement). For a particular image, numerical values come from

•Vendor: for certain products, vendor supplies 80 term RPC

•User: can obtain via regression using a dense 3D grid and corresponding image points, based on physical model

Advantage: same parameters for all sensors, easy for software applications

**Disadvantage:** no physical meaning to the coefficients



## Physical Model



Schematic of telescope optics layout for modern remote sensing camera

### Schematic of Spot Optics



From Pease, Satellite Imaging Instruments

#### Cutaway Drawing of Spot Sensor



Dimensional stability is very important to maintain good focus. The structural tubes are made from carbon fiber material with a small negative thermal expansion coefficient. The titanium fittings have a positive thermal expansion coefficient that *just cancels* the tubes. (That is good engineering!)

From Pease, Satellite Imaging Instruments

## Attitude Sensing



Quickbird Assembly

Compare

Spot: 0.2 deg @ 820 km => 2870m

Quickbird: 3 sec (a)  $450 \text{ km} \Rightarrow 7\text{m}$ 





Star Tracking Camera CT-601 from Ball Aerospace

3 arc second accuracy

## Terrestrial Photograph of Orion





#### Sensor parameters:

Focal length, principal point location, lens distortion, line rate, detector (pixel) size

#### **Platform parameters:**

Location X,Y,Z, time, attitude roll, pitch, yaw, kepler orbit elements (*a*,*e*,*i*,W,w,n)

Relate ground point and image point by equations with the above *actual physical* parameters, rather than the generic  $a_0$ ,  $a_1$ ,  $a_2$ , ... parameters.











Development of the Condition Equations for a Space Based Pushbroom Camera (Using SPOT as an Example)

#### Development of SPOT Condition Equation – Good Model for Generic Pushbroom Camera from LEO



Must have approximations for

 $\Omega$ , i,  $\omega$ , a, e

$$e = \frac{\sqrt{a^2 - b^2}}{a}$$

 $t_{f}$ : time at frame center, in header (metadata)

delta-t: delta-time from frame center, equals 0.001504 sec \* line,

$$t = t_f + \Delta t$$
  
orbit period,  $\tau = 2\pi \sqrt{\frac{a^3}{GMe}}$ 

GMe = 398600.5E09 m<sup>3</sup>/s<sup>2</sup>  $a_s = r_e + alt_s$ ,6378137m + 822000m = 7200137m  $\tau = 2\pi \sqrt{\frac{7200137^3}{398600.5E09}}$   $\tau = 6080.259$  min  $\tau = 101.338$  sec

t<sub>p</sub>: time from ascending node to perigee

#### Condition Equation cont'd.

$$t_p = \frac{\tau}{\pi} \tan^{-1} \left[ \frac{\sqrt{1-e}}{\sqrt{1+e}} \tan(\omega/2) \right] - \frac{\tau}{2\pi} \frac{e\sqrt{1-e^2} \sin\omega}{1+e\cos\omega}$$
$$\Delta t_p = t - t_p$$

$$M_n = \frac{2\pi\Delta t_p}{\tau}$$
, mean anomoly

 $E = e \sin E + M_n$ , (kepler equation, E : eccentric anomaly) solve iteratively for E

 $R_s = a(1 - e \cos E)$ ; vector from earth center to satellite



Construct  $M_b$  from 3 sequential rotations applied to XYZ (ECEF) to bring them parallel to xyz (instantaneous satellite system)



f: true anomaly

$$\sin f = \frac{\sqrt{1 - e^2} \sin E}{1 - e \cos E}$$
$$\cos f = \frac{\cos E - e}{1 - e \cos E}$$
$$f = \tan^{-1} \left(\frac{\sin f}{\cos f}\right)$$

#### Condition Equation, cont'd.

The XYZ obtained in this way will be only approximately correct and we must allow for refinements, modeled as second order polynomials of time:

$$\Delta X = \delta X_0 + \delta X_1 \Delta t + \delta X_2 \Delta t^2$$
$$\Delta Y = \delta Y_0 + \delta Y_1 \Delta t + \delta Y_2 \Delta t^2$$
$$\Delta Z = \delta Z_0 + \delta Z_1 \Delta t + d Z_2 \Delta t^2$$

Likewise the attitude (orientation) produced by the prior rotation matrix will be only approximately correct and we must allow for refinements to the attitude, again modeled as second order polynomials of time:

$$\Delta \omega = \delta \omega_0 + \delta \omega_1 \Delta t + \delta \omega_2 \Delta t^2$$
$$\Delta \varphi = \delta \varphi_0 + \delta \varphi_1 \Delta t + \delta \varphi_2 \Delta t^2$$
$$\Delta \kappa = \delta \kappa_0 + \delta \kappa_1 \Delta t + \delta \kappa_2 \Delta t^2$$

Condition Equation, cont'd.

We put these small refinement rotations into matrix as follows:

 $\mathbf{M}_{\mathbf{a}} = \mathbf{M}_{\Delta \kappa} \mathbf{M}_{\Delta \varphi} \mathbf{M}_{\Delta \omega}$ 

We must also account for a tilt or inclination of the camera. In the case of SPOT this is a cross track tilt (+/- 27 degrees) about the x (motion) axis, implemented by a stationary (but moveable) mirror:

$$\mathbf{M}_{t} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos\alpha & \sin\alpha \\ 0 & -\sin\alpha & \cos\alpha \end{bmatrix}$$

In the case of an *agile* spacecraft such as IKONOS or Quickbird, this pointing can be any arbitrary cross-track, in-track, or spin attitude, and thus requires 3 rotations:

$$\mathbf{M}_{t} = \mathbf{M}_{z}(\gamma)\mathbf{M}_{y}(\beta)\mathbf{M}_{x}(\alpha)$$

Note that we are over parameterized with rotations here. You cannot carry all as unknowns. But it may be convenient to separate in this way to make if clear which physical effect the parameter refers to.



Combine terms, eliminate scale

$$\begin{bmatrix} 0\\ y\\ -f \end{bmatrix} = \lambda \begin{bmatrix} U\\ V\\ W \end{bmatrix} \qquad 0 = -f\frac{U}{W}$$
$$y = -f\frac{V}{W}$$

 $F_x = f \frac{U}{W}$ 

small correction)

$$F_y = y + f \frac{V}{W}$$

We can also add some other inner orientation parameters such as lens distortion, principal point offset, etc.

So how many parameters do we have? There are 5 groups,

- •Orbit parameters  $\Omega, i, \omega, a, e, t_{f}$  (6)
- •Position corrections  $\delta X_0, \delta X_1, \delta X_2, \delta Y_0, \delta Y_1, \delta Y_2, \delta Z_0, \delta Z_1, \delta Z_2$  (9)
- •Attitude corrections  $\delta\omega_0, \delta\omega_1, \delta\omega_2, \delta\varphi_0, \delta\varphi_1, \delta\varphi_2, \delta\kappa_0, \delta\kappa_1, \delta\kappa_2$  (9)
- •Pointing

Inner orientation

$$\alpha_{t}(1)$$
  
x<sub>0</sub>, y<sub>0</sub>, f, k<sub>1</sub>(4)

Total here is 29, some will be held constant (maybe at zero), we may add some. Stochastic treatment is guided by redundancy, geometric strength of figure (parameters known to be highly correlated will probably not both be carried as unknowns), and by uncertainties For SPOT we get an approximation of the off-nadir attitude from the angle readout of the mirror position. For Quickbird, we have the attitude described by quaternion elements, throughout the scene.

Depending on the source of information about ground control points, we may need to do some prior transformations such as,





## Replacement Model



For the third order model, only terms with  $i+j+k \le 3$  are allowed. Those terms are shown below.

 $1, x, y, z, x^{2}, y^{2}, z^{2}, xy, xz, yz, x^{2}y, xy^{2}, x^{2}z, xz^{2}, y^{2}z, yz^{2}, x^{3}, y^{3}, z^{3}, xyz$ 

#### Rigorous Sensor Model Parameter Estimation &

#### **RPC** Parameter Estimation



Estimate RPC parameters using the *many* fictitious ground and image points



Project fictitious ground points into image by rigorous parameters



Fictitious ground points within volume

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LINE_NUM_COEFF_5: +7.042740238830377E-04	
LINE_NUM_COEFF_7: -2.118277231082864E-04	
LINE_NUM_COEFF_8: +2.806916823545727E-04	
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LINE_NUM_COEFF_12: +1.981967500179333E-05	
LINE_NUM_COEFF_13: -2.260502539903590E-05	
LINE_NUM_COEFF_15: -1.119638233066729E-05	
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LINE_DEN_COEFF_3: +2.170877012059137E-03	
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Erdas Imagine / Orthobase support for IKONOS RPC data – note the line\_numerator coefficients go up to #20, this implies a 3<sup>rd</sup> order polynomial

Evaluation of Projection Errors Using Vendor Supplied RPC Rational Polynomial Coefficients



Error Pattern for This Image – Seems to be Common Bias plus Smaller Random Part



Errors from control checkpoints



Error vectors decomposed into bias and random parts

Conjecture: The bias is itself a random vector that is consistent within an image but different between images (corollary: the biases between "same orbit" images might be correlated)



Graphical Depiction of Errors from Three Images (Conjecture – verify with actual images & GCPs)



For a single image with GCPs, can estimate bias and random part this way

$$\begin{bmatrix} e_x \\ e_y \end{bmatrix} = \lambda_{bias} \begin{bmatrix} \cos \theta \\ \sin \theta \end{bmatrix} + \begin{bmatrix} r_x \\ r_y \end{bmatrix}$$

$$a = \lambda_{bias} \cos \theta$$
$$b = \lambda_{bias} \sin \theta$$

Want to estimate one lambda and theta for each image, making the a,b substitution allows one to use a linear estimation model





# The Study (Supported by NGA)

## **Imagery Sites**





Purdue, IKONOS, stereo pair (in 2 segments)



Indianapolis, Quickbird, single image

# GPS Survey, 4 receivers, 8 sessions, NGS control and photo-ID points for imagery evaluation











Photo ID A13

NGS CP D31



NGS CP BUCK



NGS CP BUCK

Selected occupations of NGS control points and photo ID points.



#### Error Vectors of Thailand Imagery 1 (Left)



#### Error Vectors of Thailand Imagery 1 (Left)



#### Error Vectors of Fallon Imagery (Left)



#### Error Vectors of Fallon Imagery (Left)



## Error Vectors of Purdue Imagery (Left)

(Left and Right refer to a stereo pair)

Error Vectors Scattering Plot (Scale of Magnitudes is 50:1)



## Error Vectors of Purdue Imagery (Left)



Error Vectors, Decomposed into Bias (----) and Random Parts

Summary of Results for 70 Images, for 2 sensors, each with 2 classes, bias and random errors corresponding to 40% confidence region

Satellite <sup>1</sup>	Tilt Angle <sup>2</sup>	e <sub>b</sub> <sup>11</sup> (Pixel)	e <sub>r</sub> <sup>12</sup> (Pixel)
IKONOS	Small	4.15	0.95
IKONOS	Large	6.49	0.96
QuickBird	Small	11.21	1.27
QuickBird	Large	17.57	0.96

## Conclusions

•The results are surprisingly good considering that we are doing "direct geopositioning", i.e. RPC's come from the satellite navigation data (i.e. no control points used in projection – only for checking)

•Furthermore, by far, most of the error is in the common bias term, which means if you introduce one high quality control point, and augment the RPC's with shift terms, you are down in the 1-2 pixel error range

•We have recommended a method of error propagation using these eb/er terms in image space, current Eb/Er terms from the NITF standard have ambiguous definition and are not applied uniformly.