

# **GLINT**

**February 1999**

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**Task Report - Naval Research Laboratory  
Contract N00014-97-D-2014/001**

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## **GLINT**

### **Introduction**

GLINT is the acronym for Geo Light Imaging National Testbed. Schafer has supported this AFRL program in three basic areas for the Surveillance Technologies Branch (DEBS). The first area is the collation of information to develop a target database of geosynchronous (GEO) satellites. Each possible target needs to be investigated in order to obtain permission from the owner/operator to illuminate it. This information, when obtained, would be part of the database as well. The second area addresses a major operational concern for GLINT. A necessary prerequisite to illumination of GEO satellites by GLINT is the ability to ascertain that the object acquired is the intended target. The GEO satellites are unresolved to any single pupil associated with aperture sizes we are now capable of building. GLINT will be able to image them using an active Fourier telescopic scheme. However, due to international agreements and the safety of instrumentation on-board these satellites, a non-imaging method to first identify that the acquired object is the target in question. This second area is the Signatures Program. Its goal is to pursue non-imaging techniques to obtain satellite information and reduce uncertain identification (ID) in preparation for active illumination by GLINT. The final area of support was the study on a relay mirror experiment that could be performed with GLINT.

### **GLINT Target Database**

A database was created to enable target selection for the passive experiments and also for passive ID prior to active illumination by GLINT. Information on the satellite systems was included to aide the passive identification process. The database now contains the following information: catalog number, satellite name, bus structure, status, orbital inclination, eccentricity, orbit type, longitude (if applicable), number of revolutions per day, country, and a description field. The database also contains information on whether the satellite can be seen from New Mexico or Maui, and if it has been illuminated. Finally, the database contains a table of estimated magnitudes which provide an indication of the range in brightness that the satellite would have when observed. A sample view of the database is shown in Figure 1.

Schafer also supported the GLINT science team. The preliminary GEO database that was created for the passive effort was presented to some other members of the science team. Search capabilities were added to the database as shown in Figure 2.

Briefings were prepared for the GLINT Technical Interchange Meeting (TIM) with the GLINT contractors and science team to present information on the target database. Schafer presented a summary of the target selection database development at the TIM in Socorro, NM on 4 November 1998 (Attach 1).

### **Signatures Program**

Schafer supported the passive identification aspect of GLINT in the Signatures Program. Simulations were performed to tie radiometry simulation results to observations that were made at Capilla Peak Observatory of GEO satellites. The results with TASAT 6.0 were below expectations as seen in Figure 3. At the time of this final report, TASAT 7.11 had not yet been evaluated by Schafer.

Preparations were made to enable color photometry data reduction and analysis to be performed in the Schafer offices. Background studies were performed in several areas to support the GLINT science team and the passive ID portion of GLINT – IR detectors, pattern recognition statistics and algorithms, and Fourier telescropy.

Implementation of pattern recognition algorithms for the passive identification of satellite bus structures with a probability of correct identification was completed. These algorithms utilize broadband Johnson B, V, R, I photometry. The algorithms that have been tested are k-nearest neighbor, Gaussian classifier, and Mahalanobis distance used with principal components analysis. The latest results are shown in Figures 4 through 6. Schafer also developed additional pattern recognition code to aide in the determination of an optimal data processing algorithm for the color photometry data. This information allowed the principal component classifier to be updated and refined.

Schafer traveled to Maui to attend the SILC (SOI In Living Color) BattleLab project kickoff meeting. This is a spectral photometry project with similar goals to the Signatures Program. Data analysis techniques developed for GLINT will be used in this project. This contract also supported the travel to Maui by Schafer to enable and support a color photometry experiment using the Advanced Electro-Optical System (AEOS) 3.6-m telescope.

Schafer efforts also included extending the photometric study into the Infrared under an initiative entitled Near IR Field Test (NIFTE). Infrared detectors, cameras, and spectrographs were researched and the information compiled was delivered to AFRL. General support was provided for the IR experiment plans. This included radiometric studies of the solar flux in the IR.

Schafer prepared a paper for the Space Control Conference in 1999 entitled "Color Photometry of GEO Satellites" which summarizes the Signature program's results for the contract period (Atch 2).

Activities related to advocacy of the color photometry technique at the Combined Intelligence Center (CIC) and the BattleLab initiative SOI In Living Color (SILC) have been reported under Task 109.

### **Relay Mirror Experiment**

Schafer conducted a preliminary analysis of a proposed AFRL relay mirror experiment. After investigating several options, we recommended orbits to use for relay mirror experiments between Kirtland Air Force Base (AFB), NM and White Sands Missile Range NM, and between Kirtland AFB and Maui. Orbital propagation simulations were performed for experiments involving one or two relay mirrors between each pair of sites. These simulations showed how link availability, link ranges, and required beam turning angles and slew rates varied with relay mirror altitude. Schafer also wrote Matlab scripts to model laser propagation (using Fresnel diffraction calculations) through the relay mirror links. The models included the effects of wavefront error, jitter, and variable focusing by the transmitter and relay mirror optics. Schafer computed throughput efficiency and required laser pulse energy for a laser target designation demonstration. Finally, simulated images of tactical targets were created against a desert background to highlight target identification issues. This work was documented in a briefing presented

at the 23 June 1998 meeting of the AFRL relay mirror experiment working group (Atch 3).



# Target Selection Database

Dr. Mara Payne (Schafer Corp.)  
for

AFRL/DEBS GLINT Program

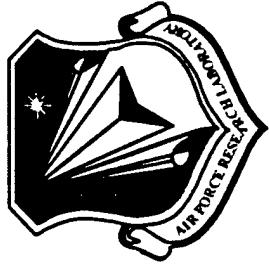
4 November 1998  
GLINT TIM (Socorro)

**Schafer**



# Target Selection Database

- Contains 697 GEOs
- Identifies GEOs visible from NM and Maui
- Includes estimate of visual magnitude for 10 solar phase angles
  - Radar cross-section converted to visual magnitude
  - Relation developed by John Lambert (Boeing) for GEOs
  - Limited to those GEOs with RCSSs
- Work in progress



# Sample Data

**SSC # 22911** Name: Solidaridad 1

Illumination  Seen From: Maui

Socorro

Bus Structure: HS-601

Status: Active

Inclination: 0.009

Stabilization: Three-axis - active

Orbit: Geostationary

Revs/Day: 1.0027

Eccentricity: 0.0002

Degrees from Zenith: 39.4

Inclination: 0.003

Stabilization: Three-axis - active

Orbit: Geostationary

Revs/Day: 1.0027

Eccentricity: 0.0002

Degrees from Zenith: 39.8

Inclination: 0.003

Stabilization: Three-axis - active

Orbit: Geostationary

**SSC # 22930** Name: DBS 1

Illumination  Seen From: Maui

Socorro

Bus Structure: HS-601

Status: Active

Revs/Day: 1.0027

Eccentricity: 0.0002

Degrees from Zenith: 39.8

Inclination: 0.003

Stabilization: Three-axis - active

Orbit: Geostationary

Inclination: 0.003

Stabilization: Three-axis - active

Orbit: Geostationary

Inclination: 0.003

Stabilization: Three-axis - active

Orbit: Geostationary

**SSC # 22930** Name: DBS 1

Illumination  Seen From: Maui

Socorro

Bus Structure: HS-601

Status: Active

Revs/Day: 1.0027

Eccentricity: 0.0002

Degrees from Zenith: 39.8

Inclination: 0.003

Stabilization: Three-axis - active

Orbit: Geostationary

Country: Mexico

Description: COMM-CIVIL

Solar Phase Angle (degrees)	0	10	20	30	40	50	60	70	80	90
Estimated Magnitudes	10.9	11.3	11.7	12.1	12.4	12.7	13.0	13.3	13.5	13.8

Solar Phase Angle (degrees)	0	10	20	30	40	50	60	70	80	90
Estimated Magnitudes	10.6	11.1	11.5	11.9	12.2	12.5	12.8	13.1	13.3	13.5

**Schaefer**



# Orbital Types

## Type 1 GEOSTATIONARY

- Inclination less than 2 degrees
- Eccentricity less than .1
- Revs per day between .9 and 1.1
- These objects stay roughly in the same part of the sky with little drift. Thus they are called GEO Stationary.

## Type 3 GEO SYNCHRONOUS

- These objects have their revs per day between .9 and 1.1 so they have *some* synchronicity with the earth even though their eccentricity and inclination may cause it to be complicated. Thus they are called GEO Synchronous.

## Type 2 GEO LONGISTATIONARY

- Inclination is greater to or equal to 2 degrees
- Eccentricity is less than .1
- Revs per day between .9 and 1.1
- These are objects that, because of large inclination, have potentially a lot of north/south motion but stay over the same longitude. Thus they are called GEO Longstationary.

## Type 4 DEEP SPACE

- These objects' revs per day are simply less than .9 so they are out beyond the GEO belt. Thus they are called Deep Space.

**Schäfer**

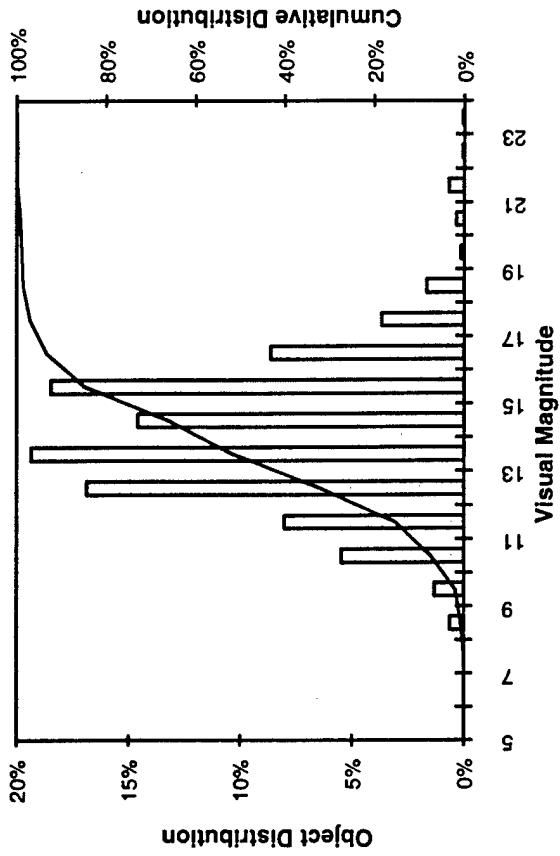


# Estimated Magnitudes

## Deep Space Optical Catalog

Developed by John V. Lambert, Boeing North American - Space Operations Center, using optical and radar cross-section data collected at Maui on deep space objects. (1996)

## Deep Space Brightness Distribution



$$M_v = M_{\text{sun}} - 2.5 \log [(a \text{ RCS}) / (4 \pi R^2)]$$

where

$$M_{\text{sun}} = -26.74 \text{ visual magnitudes}$$

$$a = \text{Albedo} = .20 \text{ (from earlier, unpublished analyses)}$$

$$R = \text{range}$$

The estimated apparent magnitude was calculated for midnight September 22, 1997 using OPAS (Orbital Propagation Analysis Software).

**Schaefer**

## Color Photometry of GEO Satellites

T. E. Payne (Schafer), D. J. Sanchez (UPR), S. A. Gregory (UNM), L. G. Finkner (Boeing), E. Caudill (AFRL), D. M. Payne (Schafer), L. Kann (AFRL), C. K. Davis (Boeing)

### Abstract

The Air Force Research Laboratory Directed Energy Directorate (AFRL/DE) Space Surveillance Technologies Branch is pursuing non-imaging techniques to obtain information on the identity and status of geosynchronous (GEO) satellites. On-going experiments have provided visible color photometry data on several GEO satellites for space object identification (SOI) purposes. These data have been analyzed using statistical probabilities to produce quantified identifications of satellite bus types and the satellites themselves. Additional analysis has been performed to ascertain the relative usefulness of each photometric bandpass in order to optimize this color photometry technique. The results and conclusions of this work to-date are presented.

### Introduction

GEO satellites pose unique challenges to the Space Surveillance Network to track, identify, and to determine anomalous behavior because the majority of them are three-axis stabilized and have no relative motion with respect to the observer. This stability makes radar tracking and imaging not feasible unless the satellite has become unstable, parts of the satellite move, or the orbit is not truly geosynchronous. Therefore, optical tracking and photometric signatures are the tools that are used most often to track, identify, and determine status or anomalies. Space object identification (SOI) of GEO satellites is a difficult task given the faint, unresolved character of the optical signal. Surveillance deficiencies have been identified related to detection of changes of status of GEO, anomaly resolution, general GEO intelligence, and the lack of global coverage. SOI needs include identification of class and type, status determination (operational or not), and anomaly resolution. Potential solutions have been identified with color photometry as one possibility to aide in resolving cross-tags and status determination. [1]

Color photometry, also known as multi-spectral photometry, and in some circles as hyperspectral which connotes very narrow bands has its origins in astronomy. Astronomical photometry is defined as the measurement of the apparent brightnesses of an object in various wavelength bands in the optical or infrared regions of the electromagnetic spectrum. The brightnesses are usually referred to in magnitudes

where  $m_1 - m_2 = -k \log_{10} \left( \frac{f_1}{f_2} \right)$ . The magnitudes assigned to the objects are  $m_1$  and  $m_2$  with energy

fluxes  $f_1$  and  $f_2$  and  $k$  is a constant. The minus sign is chosen so that brighter objects have smaller numeric values of  $m$ . Since the definition of magnitude involves two objects, magnitude is a relative concept. If a photometric system with several filter bands at different wavelengths is used, by taking the difference in magnitudes measured in two different bands, color or a color index can be formed. For instance, if the standard Johnson broadband filter system is used, the filter bands are denoted U, B and V. The wavelengths corresponding to those bands are shown in Figure 1. Then a color index can be defined as

$$B - V = K - 2.5 \log_{10} \frac{\int F_B(\lambda) d\lambda}{\int F_V(\lambda) d\lambda}; \text{ where } K \text{ is a constant. } F_x \text{ is the observed energy flux through filter}$$

x. In reality, the infinite integral is cutoff by the finite extent of the filter bandpass. So a color index essentially measures the ratio of flux between characteristic wavelengths. [2] A larger numerical value for the color index indicates a redder color, while a smaller numerical value indicates bluer color.

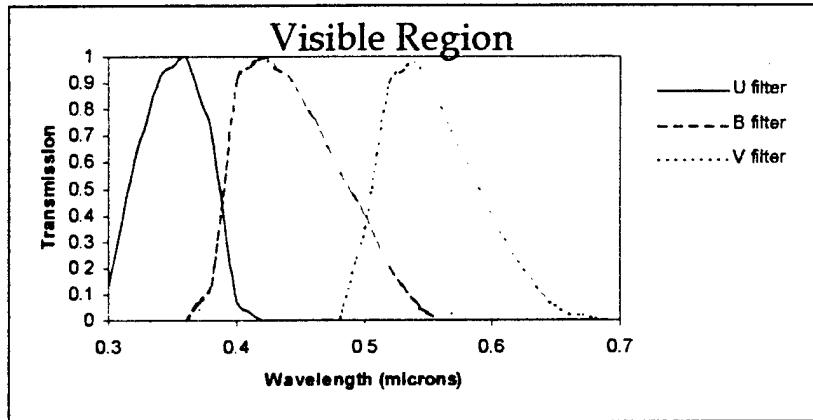


Figure 1. Standard Johnson filter bands U, B, V

Previous work supported by the AFRL Directed Energy Directorate (DE) on simulations of spectral signatures (reflected electromagnetic spectrum off GEO satellites) concluded that color photometry was a promising technique for addressing problems in GEO SOI. [3] [4] Additionally, color photometry observations were sponsored by AFRL/DE showing that different satellites had different color characteristics. [5] Prior photometric works by John V. Lambert (Boeing) and W. I. Beavers (MIT/LL) have also shown that information about the satellite was available in the brightnesses and the colors. The difficulty in any optical non-imaging technique is how to exploit that information and yield quantified variables about the satellite's identity and status. This has been the goal of the current on-going effort supported by AFRL/DE.

The current work presented here has endeavored to classify the satellite type by using its photometric colors. The extended standard Johnson system has been used to-date. These filters are designated B, V, R, and I. The properties of these filters are summarized in Table 1. [6] The current observations are being made at the Capilla Peak Observatory (CPO) outside of Albuquerque, NM. The U filter has not been used to-date due to lack of transmission at these wavelengths at this observatory. The current set of observations span from June 1998 to the present. As of this writing, 175 observations have been taken on 10 satellites, some of which are in a cluster. Table 2 contains the list of satellites.

Table 1. Johnson Filter Characteristics

Filter	Mean Wavelength ( $\lambda$ )	FWHM ( $\lambda$ )
B	4417	960
V	5505	827
R	6690	1744
I	8000	1425

Table 2. Satellites.

Satellite Name	Type
Solidaridad 1	Hughes 601
Solidaridad 2	Hughes 601
Anik E1	GE Satcom 5K
Anik E2	GE Satcom 5K
DBS 1	Hughes 601
DBS 2	Hughes 601
DBS 3	Hughes 601
AMSC 1	Hughes 601
Gstar 4	LMAS 3000
Spacenet 4	LMAS 3000

### Data Reduction and Analysis

The data is obtained using a CCD camera with the B, V, R, and I filters. Observations of stars are made for radiometry and color calibrations. Sky flat fields, bias, and dark frames are taken and used in the data reduction to minimize noise and errors in the photometric data. The instrumental magnitude is calculated using the Image Reduction and Analysis Facility (IRAF) software package. IRAF contains a large selection of computer programs for general image processing, reduction and analysis of optical and IR CCD data. [7] The instrumental magnitude ( $m_{instrument}$ ) is obtained by measuring the intensity on ( $I_{on}$ ) and off ( $I_{off}$ ) the object of interest using the IRAF software and using the relation

$$m_{instrument} = -2.5 \log \left( \frac{(I_{on} - I_{off}) \times N_{pixels}}{t_{exposure}} \right); \text{ where } N_{pixels} \text{ is the number of pixels subtended by the object and } t_{exposure} \text{ is the exposure time of the observation.}$$

object and  $t_{exposure}$  is the exposure time of the observation. The instrumental magnitude is then converted to a standard magnitude using a relation between the standard magnitude and the instrumental magnitude as shown in Figure 2. All other filters are similarly treated. This process removes the effects of the Earth's atmosphere and the telescope. The final standard magnitudes have an uncertainty of .03 magnitudes and are exoatmospheric

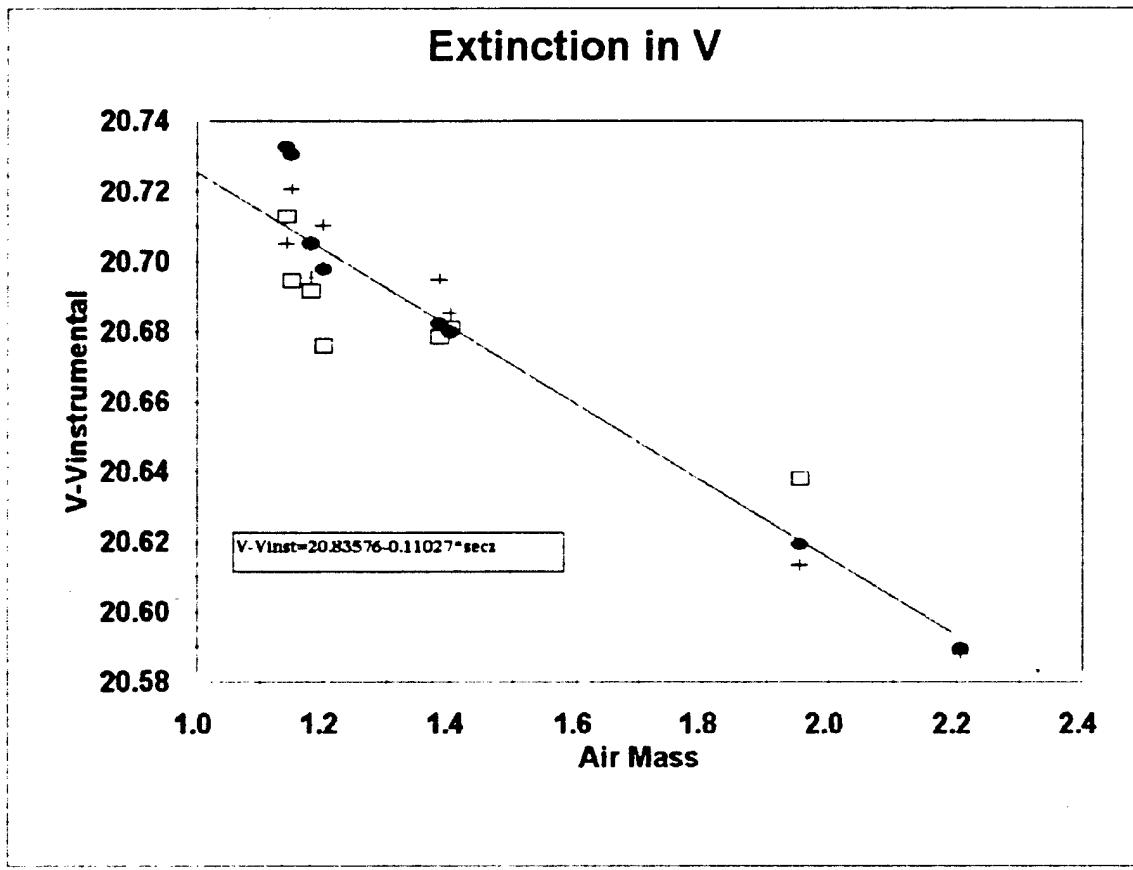


Figure 2. Extinction in the V filter.

### Photometric Analyses and Results

As can be seen in Table 2, all the satellites under consideration are three-axis stabilized. Their features are very similar as is apparent in Figure 3. The color photometry data taken on these satellites was processed using two different types of groupings. One group separates the 10 satellites into 3 classes by type of

payload. The second group separates the satellites into 5 classes by type of satellite. Table 3 shows these groupings.

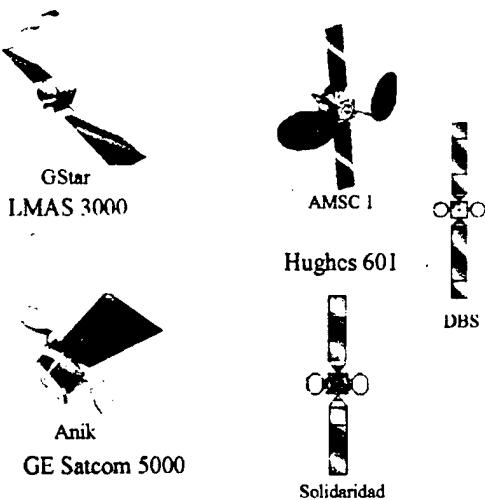


Figure 3. Depictions of some of the satellites.

Table 3. Satellite groupings.

Satellite Name	Type	Group 1	Group 2
Solidaridad 1	Hughes 601	1	1
Solidaridad 2	Hughes 601	1	1
DBS 1	Hughes 601	1	4
DBS 2	Hughes 601	1	4
DBS 3	Hughes 601	1	4
AMSC 1	Hughes 601	1	5
Anik E1	GE Satcom 5K	2	2
Anik E2	GE Satcom 5K	2	2
Gstar 4	LMAS 3000	3	3
Spacenet 4	LMAS 3000	3	3

For illustrative purposes, Figures 4 - 5 plot the data when magnitudes and colors are plotted against one another. The data is plotted using Group 1 designations. By themselves, these data do not separate out the satellite classes completely.

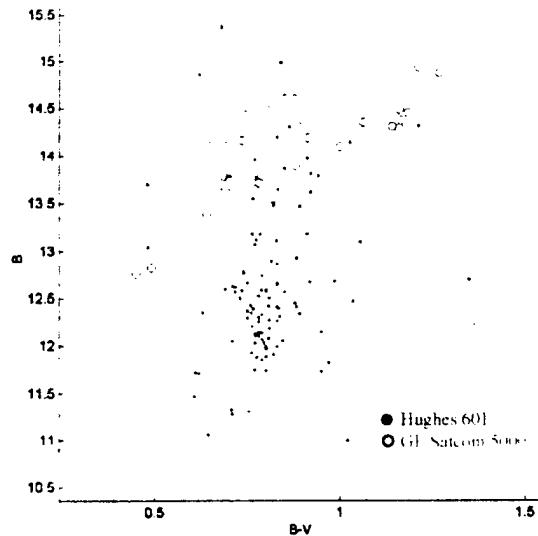


Figure 4. Color-Magnitude Plot

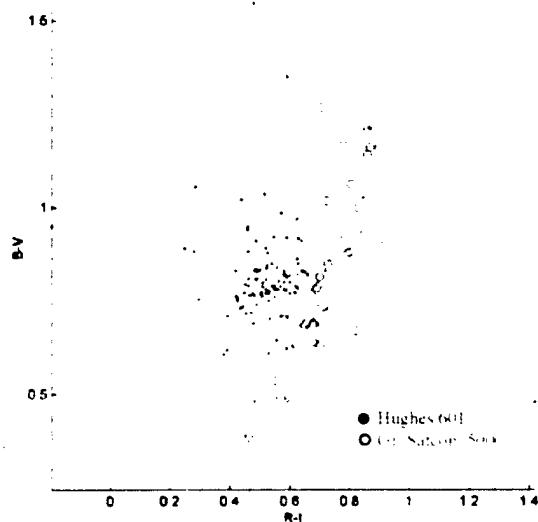


Figure 5. Color-Color Plot

Therefore, it was determined that pattern recognition algorithms may be applicable to this problem and may separate the classes better. In addition, the application of these algorithms would enable quantified confidence levels to be calculated on the ability to correctly identify the satellite class. Three different discriminants or classifiers are currently being tested: K-nearest neighbor, Gaussian classifier, and Mahalanobis distance. These are each used in combination with principal component analysis to obtain a percentage of correct identification of each group. A brief description of each algorithm follows.

In pattern recognition terminology, the magnitude and color data creates a feature space. A feature space is defined by the different features of the data. These features describe the data in terms that are of interest for a specific problem. For instance, medical data on people can define a feature space whose dimensions are height, weight, age, and sex. In this case, the dimensions of our feature space are the magnitudes B, V, R, I, and the colors, B-V, V-R, and R-I. Feature space, in general, is a n-dimensional space that is defined by the properties that describe the population of interest.

Principal component analysis is a common type of preprocessing that aims at reducing the number of input variables or the dimensions of the feature space while maintaining the most significant relationships reflected by the data. This technique attempts to identify an m-dimensional subspace of the n-dimensional feature space that seems most significant, and then projects the data onto this subspace. [8]

K-nearest neighbor is a classifier in which the position in feature space of the k-nearest neighbors whose identities are known is compared with the position of the data point whose identity is to be determined. This relative position is used to determine the identity of the unknown data point, where k is an integer variable ranging from 1 to the number of data points minus one. With the current data, the best results are obtained when  $k = 3$ . For example, if the three closest data points in the feature space to the unknown data point are identified as Hughes 601 (using Group 1), then the unknown data point is identified as Hughes 601 with 100% probability.

The Gaussian classifier fits a Gaussian with a mean and standard deviation in feature space to each class. An unknown data point's location in feature space is then compared with the position, shape, and extent of the Gaussians in the feature space. If the unknown data point is at a location in feature space that is described by only one Gaussian, then its identity is known to 100% probability. If that location is shared by two Gaussians, then the Euclidean distance from the mean of each Gaussian is computed and relative probabilities are generated that the unknown data point belongs to each population represented by each Gaussian [9].

The Mahalanobis distance is used in the algorithm above instead of the Euclidean distance. When the Gaussian classifier is used, there is a possibility of error unless all the Gaussians have the same shape and extent and the probabilities are equal that the unknown data point is any of the classes. So, classification based on the nearest Gaussian may be near optimum only if appropriate additional weights are introduced by using a Mahalanobis distance. [10]

These color photometry data were processed using the above techniques. Their results were then studied to determine which filter combinations, groups, and classifiers yielded the best results. When the data was combined using Group 2, the results were much poorer than when the data was combined using Group 1. One major factor is the limited amount of observations. The results for Group 1 using the three classifiers are shown in Figures 6 - 8. They show correct identification using k-nearest neighbor, 81.14% using the Gaussian and 86.29% using the Mahalanobis distance. The results for Group 2 using k-nearest neighbor are 52.00% correct identification, using a Gaussian they are 45.71%, and the results are 55.43% using the Mahalanobis distance.

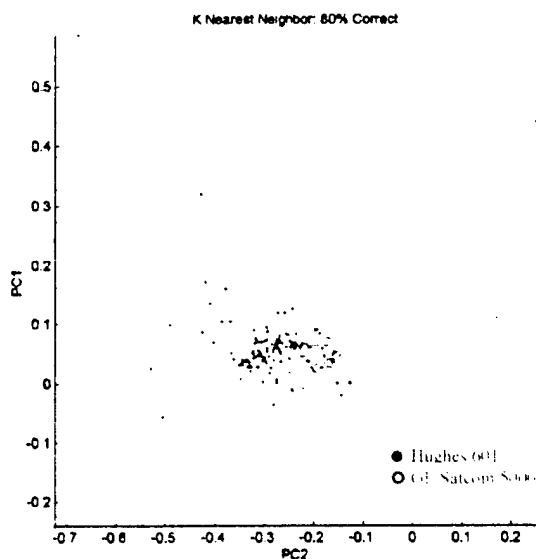


Figure 6. K-nearest neighbor ( $k=3$ ).

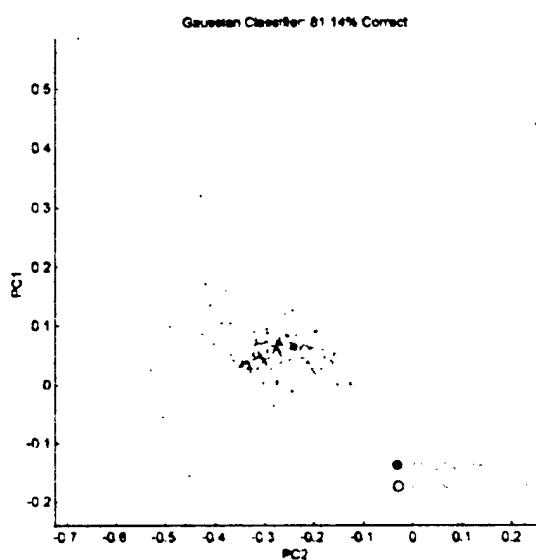


Figure 7 Gaussian classifier

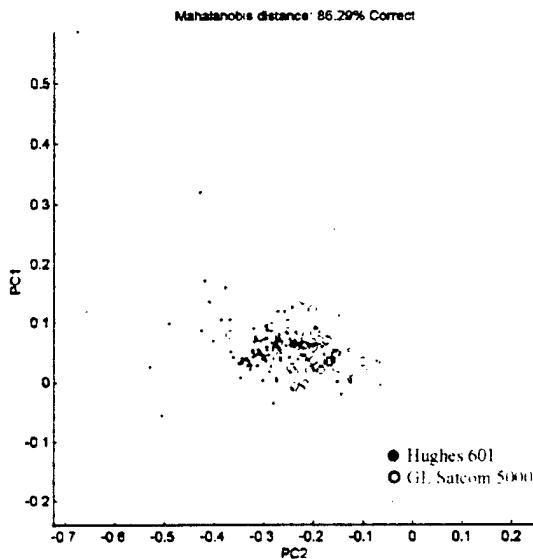


Figure 8. Mahalanobis distance.

### Conclusions

Our results show that there is utility in color photometry data of GEO satellites and that the information can be obtained to achieve identification with a good probability. This technique shows promise in minimizing and correcting cross-tags. Because of the limited amount of data, plans have been made to continue to take color photometry of these satellites in the filters discussed here. No one filter combination stood out as yielding more information than the others. Plans have been made to obtain color photometry data using different filters in addition to the current ones. These new filter characteristics were designed from the information obtained from the spectral simulations previously cited and from the work of Poelman. [3] [4] [11] The detailed shape of an energy distribution (i.e. spectrum) is determined by a few basic physical parameters. By an astute choice of filters, features in the energy distribution that are sensitive to these parameters can be isolated. [2] Further efforts are in the areas of pattern recognition algorithm development, incorporation of solar and sensor angles (phase angles) into the feature space, the continued development of a color photometric database, and the possibility of photometry in other regions of the spectrum. Some of this work will be in collaboration with the Space Battlelab initiative SOI In Living Color (SILC). The goal of these efforts is threefold. first, to obtain the ability to identify active payloads; second, to be able to provide a tip-off of anomalous behavior; and finally, to be able to provide information on dead satellites, i.e. to provide some information on what is wrong.

### References

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9. Rogers, S. K., and Kabrisky, M., An Introduction to Biological and Artificial Neural Networks for Pattern Recognition, SPIE, Bellingham, WA, 1991.
10. Pao, Yoh-Han, Adaptive Pattern Recognition and Neural Networks, Addison-Wesley, Reading, MA, 1989.
11. Poelman, C. J., and Meltzer, S. R., Spacecraft Identification By Multispectral Signature Analysis Using Neural Networks, PL-TR-97-1053, March 1997.

# Bifocal Relay Experiment

Dustin Johnston

Tom Brown

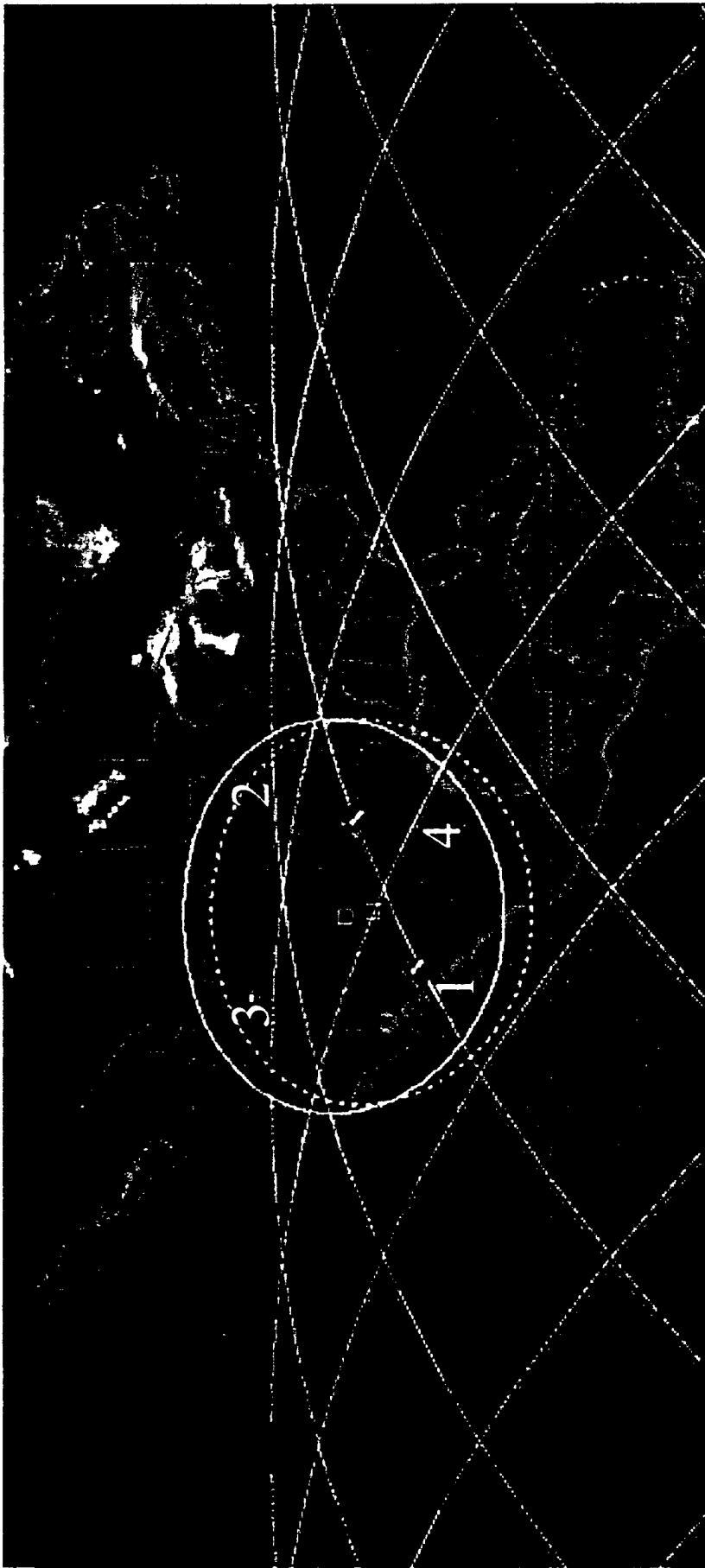
Bob O'Leary

Don Payne

23 June 98

# Outline

- SOR to WSMR Relay Mirror Experiment
  - 1-Bounce Relay
    - » Beam turning angle issue
    - » Link availability
    - » Throughput
  - 2-Bounce Relay
    - » Link availability
- SOR to Maui Relay Mirror Experiment
  - 1-Bounce Relay
    - » Link availability
    - » Throughput
  - 2-Bounce Relay
    - » Link availability
- Imaging / target identification issues

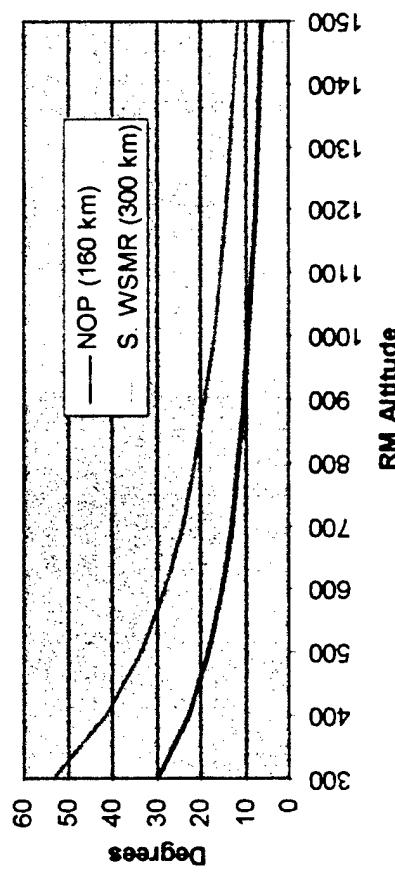


Minimum elevation angle at both sites:  $30^\circ$   
Altitude = 1000 km

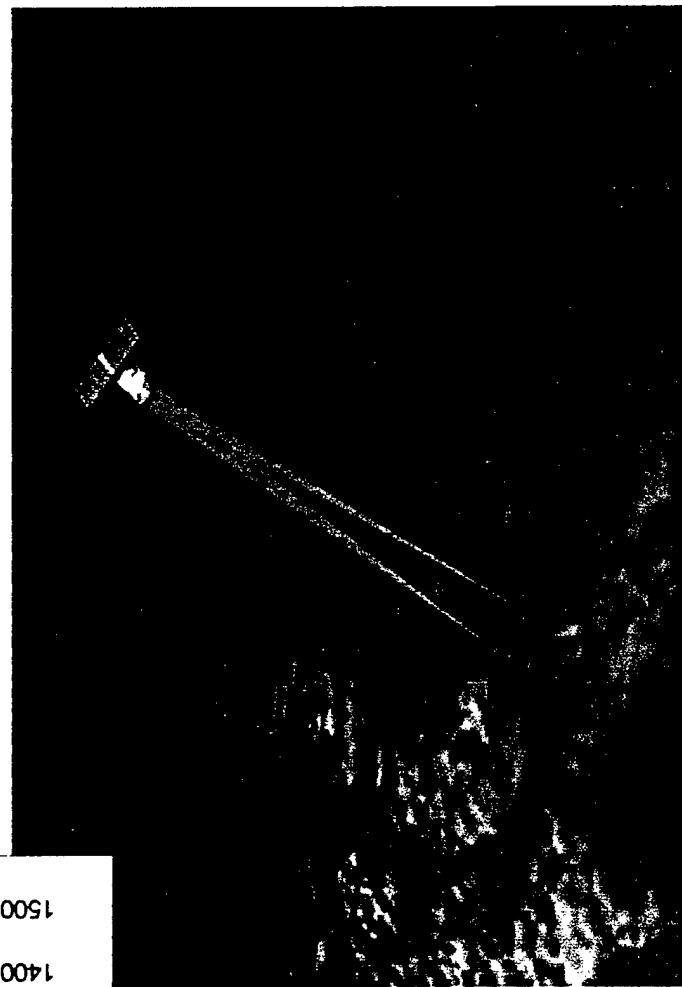
# SOR to WSMR 1-Bounce Relay

- Sharp beam-turning angle

Beam Turning Angle - RM Halfway Between Sites



- Altitude: 500 km
- Inclination: 40°



- Results of 1-year simulation (1-minute time steps)

- Average link open 3.9 minutes
- Average uplink: 677 km
- Average downlink: 709 km
- Average beam turning angle: 25°
  - » Minimum: 15 °
- Maximum slew rate: 188 µrad/s

# SOR to WSMR 1-Bounce Relay

## Satellite Altitude vs. Link Availability

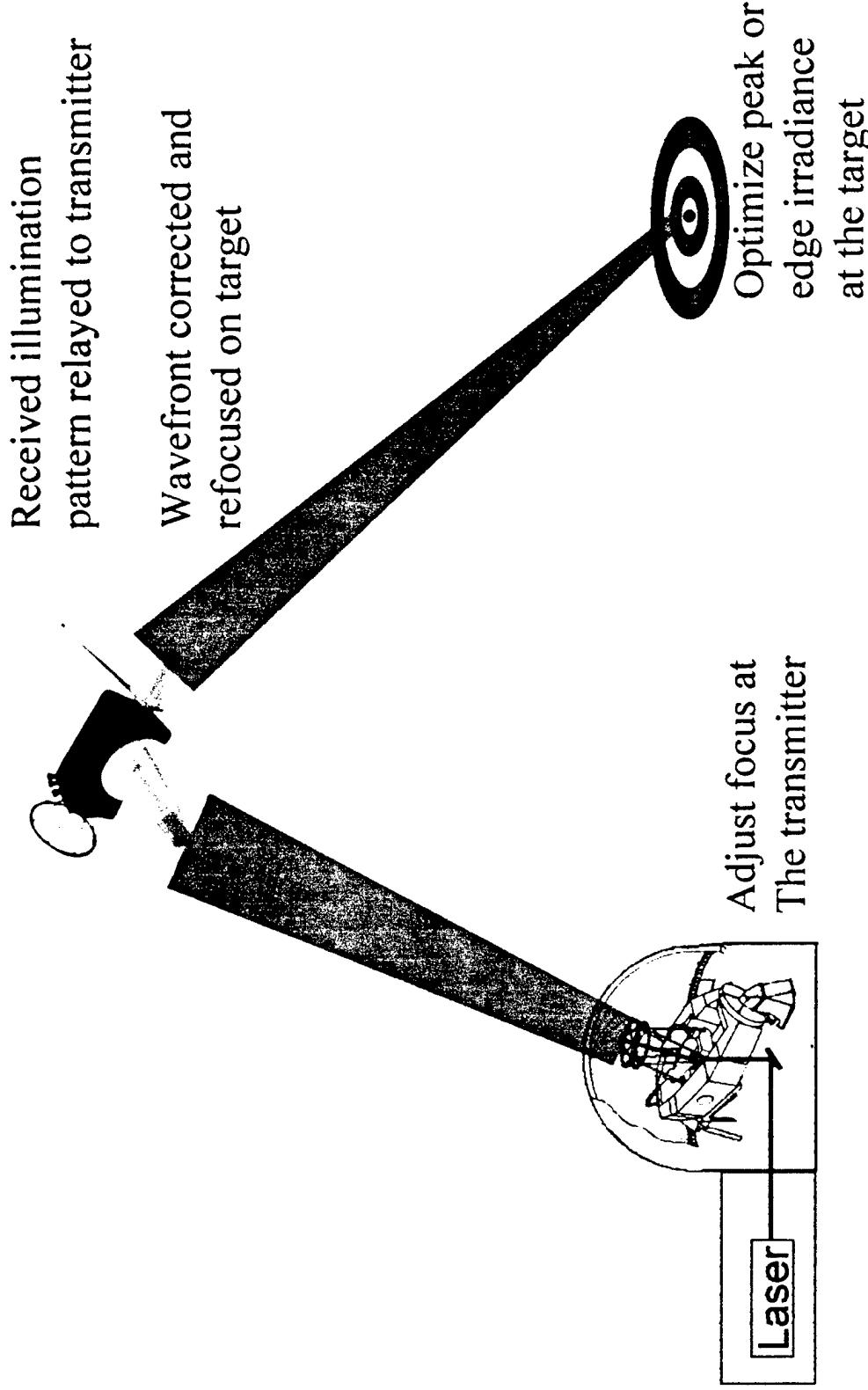
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- Inclination:  $40^\circ$
- Minimum elevation angle:  $30^\circ$

### Satellite Altitude

Links / year	1108	1641	1668	1500 km
Average link time	3.9 min	7.6 min	11.1 min	

# Minimizing Diffraction Loss



Source: John Erkkila, Logicon RDA

# Target Irradiance / Fluence at Edge of 1-m Spot SOR 1-bounce relay to WSMR

Transmitter diameter = 3.5 m	Transmitter optics transmission = 0.6
$\lambda = 1 \mu\text{m}$	Bifocal optical transmission = 0.8
Uniform illumination of xmtr aperture	Atmospheric transmission each way: 0.8
No jitter or wavefront error	

Xmtr focus adjusted to maximize edge fluence at tgt (Fresnel diffraction calculation)  
Relay mirror focused at target

## Uplink range = downlink range

500 km	1000 km	1500 km	
<u>Relay mirror diameter</u>	1 m	$1.1 \cdot 10^{-5}$	$8.6 \cdot 10^{-6}$
	1.5 m	$6.5 \cdot 10^{-6}$	$1.1 \cdot 10^{-5}$

W/cm<sup>2</sup> per watt transmitted  
J/cm<sup>2</sup> per J/pulse transmitted

## Link Efficiency SOR 1-bounce relay to WSMR

Transmitter diameter = 3.5 m	Transmitter optics transmission = 0.6
$\lambda = 1 \mu\text{m}$	Bifocal optical transmission = 0.8
Uniform illumination of xmtr aperture	Atmospheric transmission each way: 0.8
No jitter or wavefront error	

Xmtr focus adjusted to maximize edge fluence at tgt (Fresnel diffraction calculation)  
Relay mirror focused at target

### Uplink range = downlink range

500 km	1000 km	1500 km
<u>Relay mirror diameter</u>	1 m	0.24
	1.5 m	0.38

Fraction of transmitted pulse energy inside 1-meter spot at target

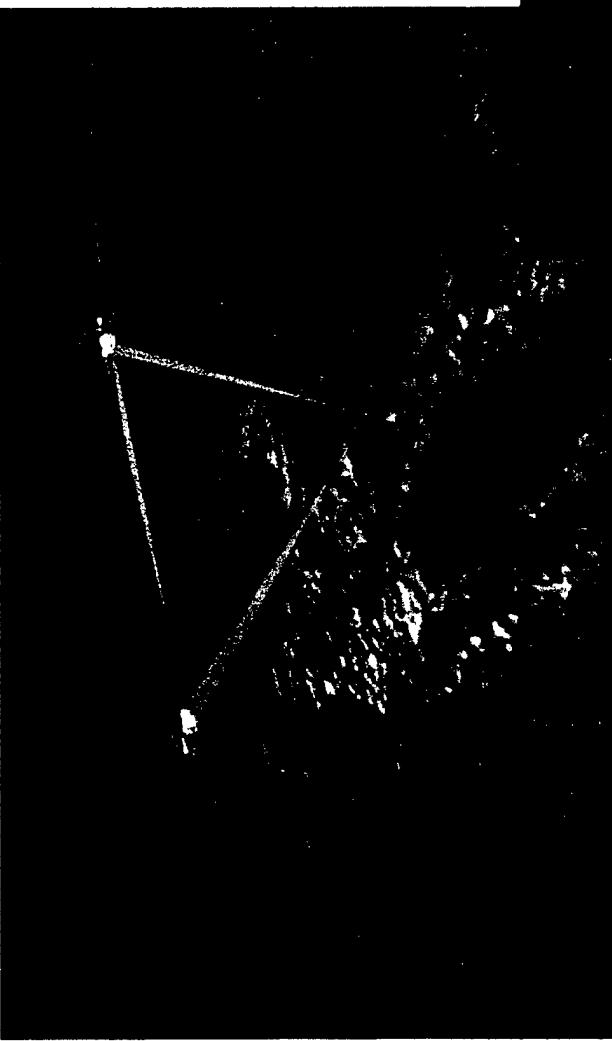
# Target Designation Application

- At LASSOS Kickoff Meeting, a typical laser target designator pulse energy was given as 100 mJ / pulse
- Conservatively, assume we need this much energy per pulse in the laser spot for the relay mirror experiment
  - Future work should include modeling target reflectivity, sensor, and engagement geometry
- The table below gives required transmitter pulse energies based on link efficiencies given in previous chart

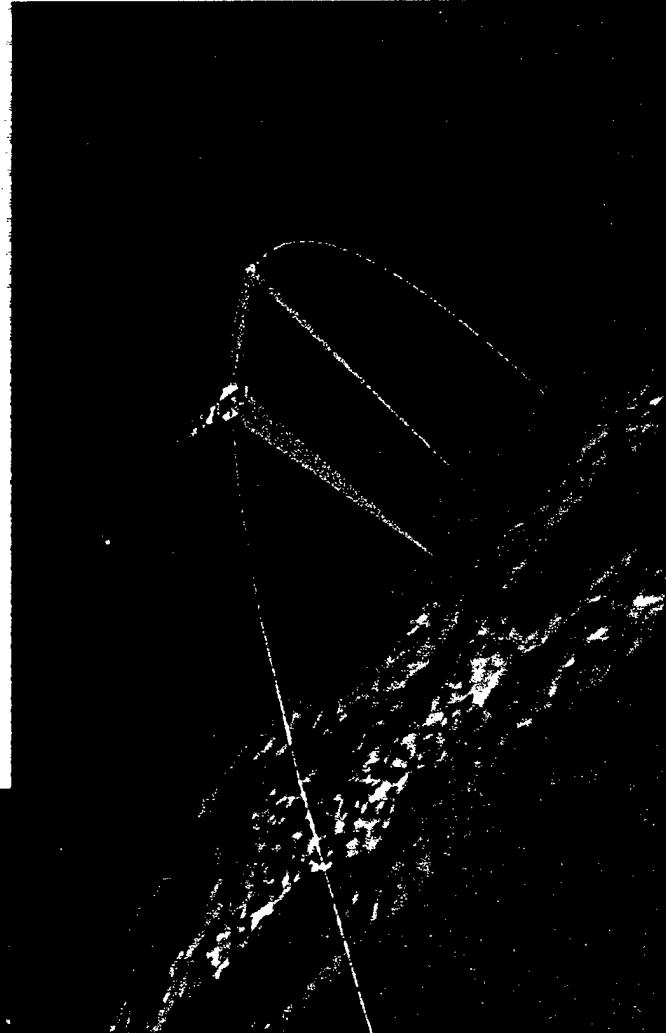
Uplink range = downlink range

<u>Relay mirror diameter</u>	1 m	410 mJ	1.1 J	2.5 J
	1.5 m	270 mJ	450 mJ	1.2 J

# SOR to WSMR 2-Bounce Relay



- Altitude: 500 km
- Inclination:  $40^\circ$
- Satellite separation:  $6^\circ$ 
  - $\sim 720$  km



- Results of 1-year simulation  
(1-minute time steps)
  - Average link open 1.9 minutes
  - Average uplink: 712 km
  - Average downlink: 722 km
  - Average beam turning angle:  $61^\circ$
  - Maximum slew rate:  $715 \mu\text{rad/sec}$

# SOR to WSMR 2-Bounce Relay Satellite Altitude vs. Link Availability

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- Inclination:  $40^\circ$
- Satellite separation:  $6^\circ$  ( $\sim 720$  km)
- Minimum elevation angle:  $30^\circ$

## Satellite Altitude

Links / year	987	1624	1664	1500 km
Average link time	1.9 min	4.9 min	8.0 min	

# Link Efficiency

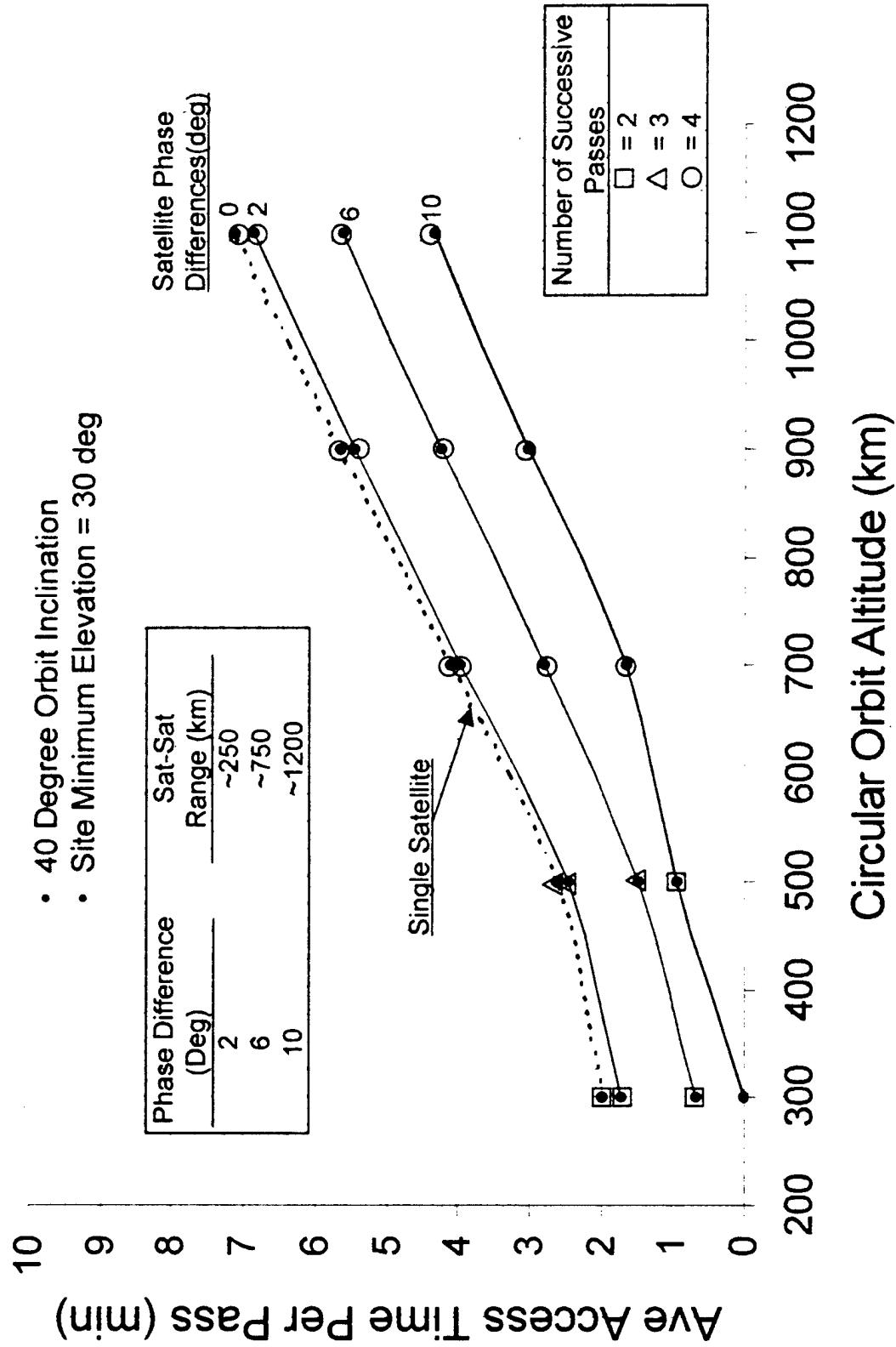
## SOR 2-bounce relay to WSMR

Transmitter diameter = 3.5 m	Transmitter optics transmission = 0.6
$\lambda = 1 \mu\text{m}$	Bifocal optical transmission = 0.8
Uniform illumination of xmtr aperture	Atmospheric transmission each way: 0.8
No jitter or wavefront error	
Transmitter focused at first relay mirror, which is focused at second relay mirror	
Second relay mirror focused at target	

- Results for 1000-km up and downlink; 750-km crosslink
  - Fraction of transmitted pulse energy inside 1-meter spot at target
    - » 1.0-meter relay mirror: 0.043
    - » 1.5-meter relay mirror: 0.12
  - Energy per pulse required at transmitter to put 100 mJ on target
    - » 1.0-meter relay mirror: 2.3 J
    - » 1.5-meter relay mirror: 830 mJ

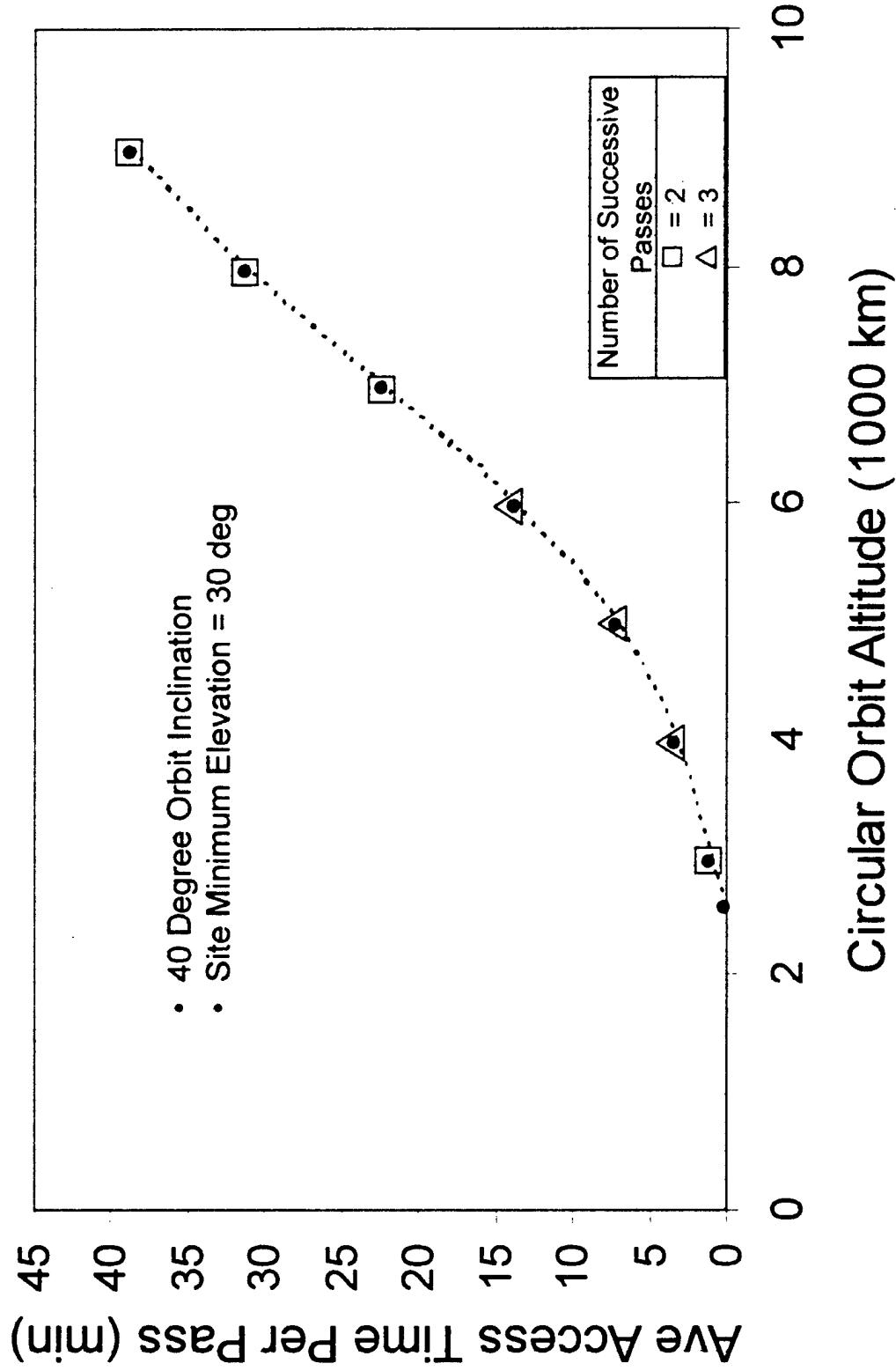
# Access Time Per Pass for One & Two Satellite Relay From SOR to WSMR

- 40 Degree Orbit Inclination
- Site Minimum Elevation = 30 deg



Data from 24-hour STK run

# Access Time Per Pass For Single Satellite Relay From SOR to MAUI



Note: Radiation shielding is a design issue

Data from 24-hour STK run

# SOR to Maui 1-Bounce Relay

## Link Availability

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- Orbit / link parameters
  - Inclination: 40°
  - Altitude: 5000 km
  - Minimum elevation angle: 30°
- Results of 1-year simulation (1-minute time steps)
  - Link availability (not including PCFLOS)
    - » Number of valid links per year: 1964
    - » Average link is open 9.6 minutes
  - Relay mirror design drivers
    - » Maximum uplink/downlink: 6758 km
    - » Maximum total range: 13489 km
    - » Minimum beam turning angle: 43.5 °
      - Average: 47.4 °
  - » Maximum slew rate: 31.6 µrad / sec

# Target Irradiance / Fluence at Edge of 1-m Spot SOR 1-bounce relay to Maui

Transmitter diameter = 3.5 m  
 $\lambda = 1 \mu\text{m}$   
Uniform illumination of xmtr aperture  
No jitter or wavefront error

Transmitter optics transmission = 0.6  
Bifocal optical transmission = 0.8  
Atmospheric transmission each way: 0.8

Xmtr focus adjusted to maximize peak fluence at tgt (Fresnel diffraction calculation)  
Relay mirror focused at target

## Uplink range = downlink range

<u>Relay mirror diameter</u>	1 m	$2.2 \cdot 10^{-7}$	$1.1 \cdot 10^{-7}$	$6.8 \cdot 10^{-8}$	
	1.5 m	$8.6 \cdot 10^{-7}$	$5.4 \cdot 10^{-7}$	$3.0 \cdot 10^{-7}$	

W/cm<sup>2</sup> per watt transmitted  
J/cm<sup>2</sup> per J/pulse transmitted

Based on calculations by John Erkkila, Logicon RDA

# **Schaefer**

## Link Efficiency

SOR 1-bounce relay to Maui

Transmitter diameter = 3.5 m

$\lambda = 1 \mu\text{m}$

Uniform illumination of xmtr aperture

No jitter or wavefront error

Xmtr focus adjusted to maximize peak fluence at tgt (Fresnel diffraction calculation)

Relay mirror focused at target

Transmitter optics transmission = 0.6

Bifocal optical transmission = 0.8

Atmospheric transmission each way: 0.8

## Uplink range = downlink range

5000 km

6000 km

7000 km

Relay mirror diameter

1 m	0.0017	0.00086	0.00054
(58 J)	(120 J)	(190 J)	

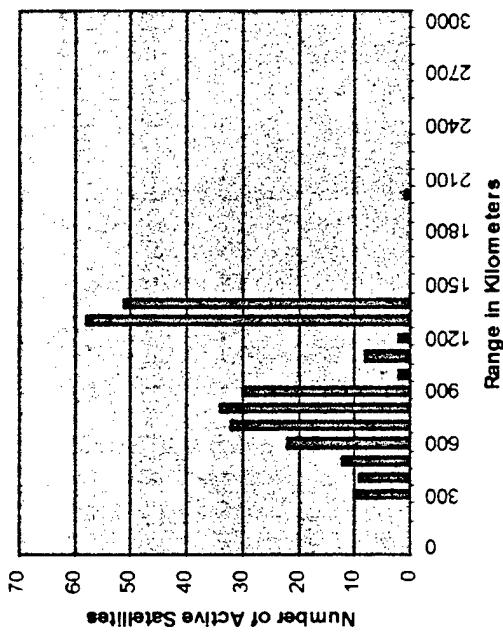
1.5 m	0.0070	0.0043	0.0024
	(14 J)	(23 J)	(42 J)

Fraction of transmitted pulse energy inside 1-meter spot at target  
(Energy per pulse required at transmitter to put 100 mJ on target)

# Schaefer

# Radiation Environment Considerations

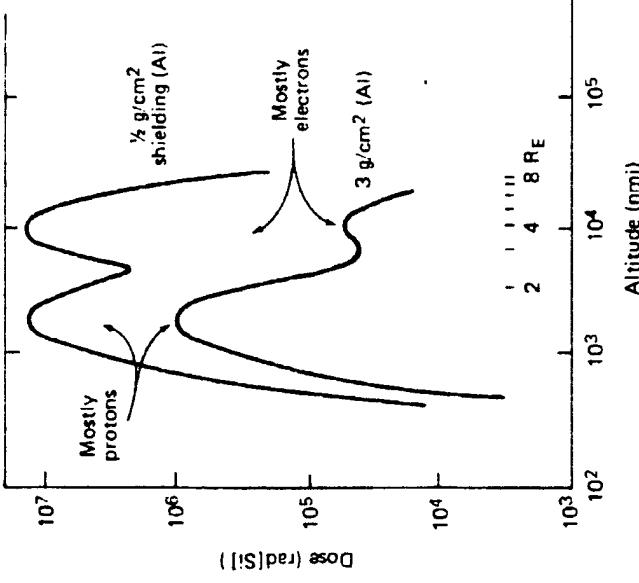
Number of Active Satellites By Altitude Range



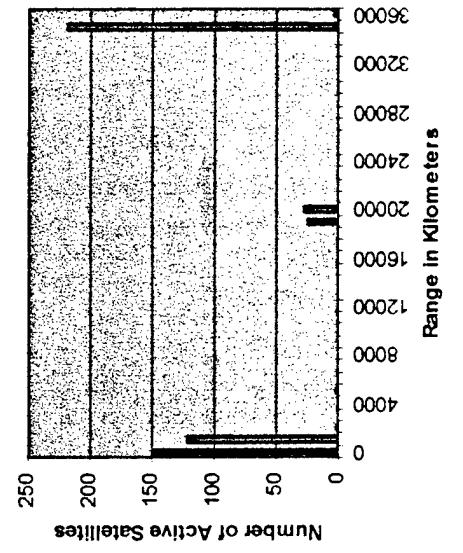
## SPACE VEHICLE DESIGN

Griffin and French, 1991

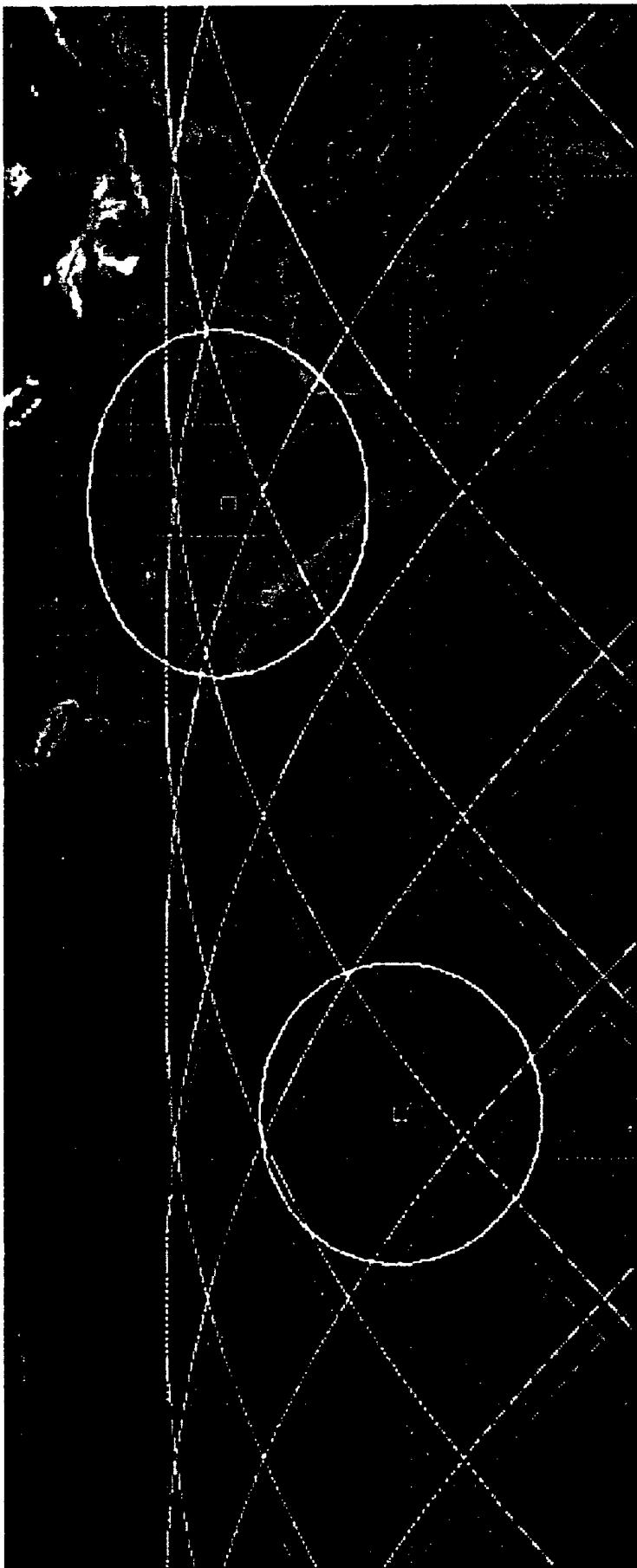
- Electron and proton/spherical shielding
- 10 year mission
- Circular orbits/ $0^\circ$  inclination



Number of Active Satellites By Altitude Range



# SOR to Maui 2-Bounce Relay



- Altitude: 1000 km
  - Inclination:  $40^\circ$
  - Satellite separation:  $45^\circ$ 
    - $\sim 5600$  km
- If satellite altitude is at least 600 km,  
only a phase change is required  
to enable SOR-Maui link

# SOR to Maui 2-Bounce Relay

## Link Availability

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- Results of 1-year simulation (1-minute time steps)
  - Link availability (not including PCFLOS)
    - » Number of valid links per year: 570
    - » Average link is open 5.4 minutes
- Relay mirror design drivers
  - Maximum uplink: 1702 km
  - Maximum crosslink: 5647 km
  - Maximum downlink: 1702 km
  - Maximum total range: 9043 km
  - Minimum beam turning angle:  $22.6^\circ$ 
    - » Average:  $72^\circ$
  - Maximum slew rate:  $365 \mu\text{rad} / \text{sec}$

# Link Efficiency

## SOR 2-bounce relay to Maui

Transmitter diameter = 3.5 m	Transmitter optics transmission = 0.6
$\lambda = 1 \mu\text{m}$	Bifocal optical transmission = 0.8
Uniform illumination of xmtr aperture	Atmospheric transmission each way: 0.8
No jitter or waveform error	
Transmitter focused at first relay mirror, which is focused at second relay mirror	
Second relay mirror focused at target	

- Results for 1300-km up and downlink; 5600-km crosslink
  - Fraction of transmitted pulse energy inside 1-meter spot at target
    - » 1.0-meter relay mirror: 0.0013
    - » 1.5-meter relay mirror: 0.0045
  - Energy per pulse required at transmitter to put 100 mJ on target
    - » 1.0-meter relay mirror: 77 J
    - » 1.5-meter relay mirror: 22 J

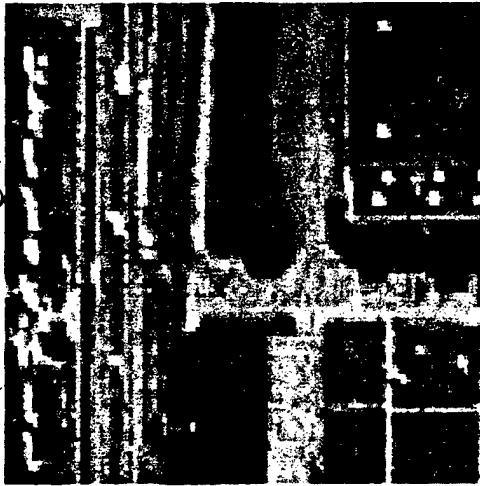
# Imaging Considerations

## Example imagery from Earth Watch web site

(actual image)



(simulated image)



Imaging  
aperture  
diameter

.5-m resolution

3-m resolution

1 m

1600 km

4900 km

1.5 m

1200 km

2500 km

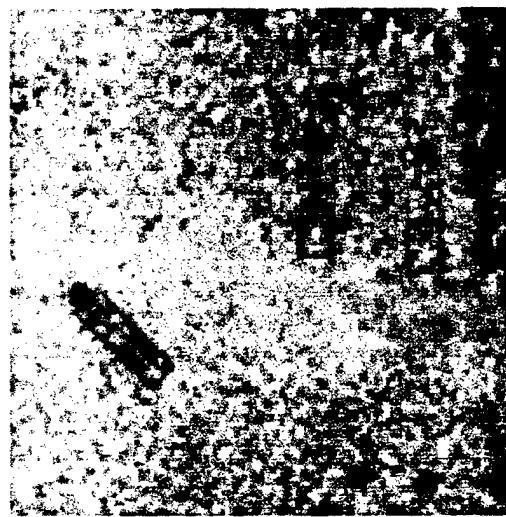
7400 km

(Range at which Rayleigh criterion is met at  $0.5 \mu\text{m}$ )

# Synthetic Color Images with 1.5m Aperture Camouflaged Scud Launcher Against Desert Background



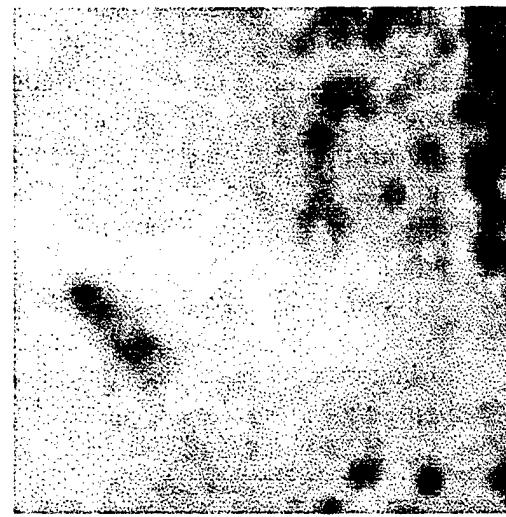
(truth) .25 m/pix



1500 km .55m/pix



2000 km .73m/pix



500 km .18m/pix



1000 km .37m/pix

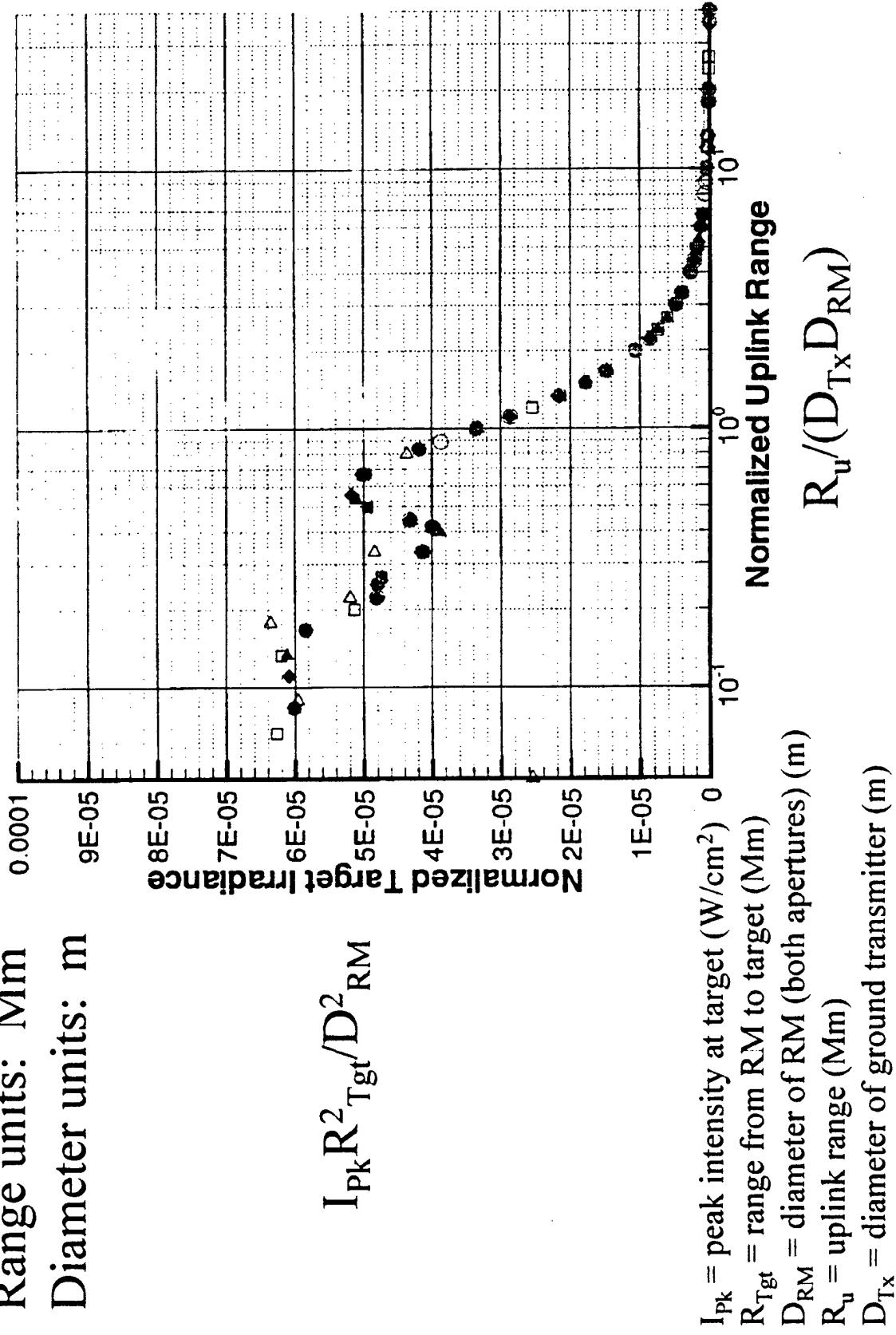
20,000 km 7.33 m/pix

- Modeling of jitter and wavefront error
  - Jitter: convolve Gaussian function with diffraction patterns
  - Wavefront: add random residual phase error to aperture function
- Modeling of sensor / receiver systems
  - Target identification system
  - Laser target designator
  - Doppler wind measurement
- Modeling of pointing, acquisition, and tracking
  - Analysis of multi-wavelength operations
    - Thermal issues at mid- to long-IR

Backup slides follow

# Throughput (Diffraction Only)

Range units: Mm  
Diameter units: m



Source: John Erkkila, Logicon RDA