

# BIOPHYS

## A PHYSICALLY-BASED CONTINUOUS FIELDS ALGORITHM and Climate and Carbon Models

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$s_i = (LAD, LOP, BOP, h/w, LAI_c, LAI_b, CC)$



**CANOPY  
MODELER**

# The Chasm

**NDVI**



**Climate, Carbon  
Modeler**

Satellite  
 $s_i(\lambda) = (LOP, BOP, Cg, h/w, LAI_C, LAI_b, CC)$

Remote Sensing Algorithms  
 $F_{par} = F(s_i | LC)$   
Land Cover (LC) Classification

Parameter Maps  
( $F_{par}, LC$ )  
 $LAI = F(F_{par}, LC)$

LOP&BOP= leaf&background optical properties  
 $Cg$  = Canopy geometry  
 $LAI_C$  = Crown leaf area index  
 $LAI_b$  = Branch leaf area index  
 $CC$  = fractional crown cover

Classification Algorithms (CA)  
 $LC = CA(s_i)$

Climate, Carbon Models

$\alpha, F_{par}, g_c, Z_0, P_v, NP_v, LAI = F(LAI, F_{par}, LC)$   
B Cndtns, Validation, Assimilation

Problems  
**Algorithm Locality:** Spectral change with illumination view geometry, background, season etc.affects robustness.  
**“Chicken & Egg Syndrome”:** Radiance correction for view and illumination angle to infer biophysical parameters depends on the parameters.  
**Parameter Inconsistency:** Different algorithms for different parameters.  
**Error structures:** Error, uncertainty characterization approximate, empirical.  
**Model utility:** Inability to directly assimilate satellite radiances in CWE models.

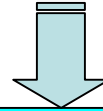
# BIOPHYS

An Analysis Framework linking 30+ Years  
of Canopy Radiative Transfer Model  
Research Directly to Biophysical  
Parameter Retrieval

# Satellites

(spectral, angular, temporal)

$s_i$  (áí)



Biophys Framework

CANOPY CRT MODEL (Tables)

$\{s_i\}_= (LOP, BOP, Cg, LAI_C, LAI_b, CC)$

**Satellite Measurement (L2)**

$\{s_i \pm \theta \text{ } s_i\}_i$  i bands, angles, dates

Table Lookup

$\{LOP, BOP, Cg, LAI_C, LAI_b, CC\}_k \in \{s_i \pm \theta \text{ } s_i\}_i$

**Compute Parameter Statistics**

$\langle \{LOP, BOP, Cg, LAI_C, LAI_b, CC\} \rangle$

**Model Parameters**

$\alpha, F_{par}, g_c, Z_0, P_v, NP_v =$

$F(\langle \{LOP, BOP, Cg, LAI_C, LAI_b, CC\} \rangle)$

Numerical, Non-analytic

CANOPY CRT  
MODELS  
GOMS, GORT, 5-scale

CWE Parameters  
 $\alpha, F_{par}, g_c, Z_o, P_v, NP_v, \dots$   
Canopy RT Parameters  
LAD, LOP, BOP, h/w, LAI<sub>c</sub>, LAI<sub>b</sub>, CC

Climate, Carbon  
Models

BIOPHYS

MODIS SURFACE  
Multi-Date BRDF Values  
 $(\theta_i, \phi_i)$   
Canopy Parameter Values  
LAD<sub>i</sub>, LOP<sub>i</sub>, BOP<sub>i</sub>, h/w<sub>i</sub>, LAI<sub>c</sub>, LAI<sub>b</sub>, CC<sub>i</sub>

MODIS  
MEASURED BRDF  
 $(\theta_i, \phi_i, t_i)$

o  
o  
o  
LookUp  
Table

CANOPY CRT MODELS  
GOMS, 5-scale

CWE Parameters  
 $\alpha, F_{par}, g_c, Z_o, P_v, NP_v, \dots$   
Canopy RT Parameters  
LAD, LOP, BOP, h/w, LAI<sub>c</sub>, LAI<sub>b</sub>, CC

Climate, Carbon Models

BIOPHYS/MFMM  
SIMPLE, ELEGANT

MODIS SURFACE  
Multi-Date BRDF Values  
( $\theta_i, \phi_i$ )  
Canopy Parameter Values  
LAD<sub>i</sub>, LOP<sub>i</sub>, BOP<sub>i</sub>, h/w<sub>i</sub>, LAI<sub>c</sub>, LAI<sub>b</sub>, CC<sub>i</sub>

MEASUREMENT = [(RED = 0.0487 & NIR = 0.2172 at SZA = 30°),  
(RED = 0.0372 & NIR = 0.1730 at SZA = 45°),  
(RED = 0.0458 & NIR = 0.1474 at SZA = 50°)]

Canopy Parameter Possibilities  
LAD<sub>i</sub>, LOP<sub>i</sub>, BOP<sub>i</sub>, h/w<sub>i</sub>, LAI<sub>c</sub>, LAI<sub>b</sub>, CC<sub>i</sub>

TABLE LOOKUP  
RETURN ALL ROWS FROM LOOK UP TABLE : SELECT \* FROM TABLE WHERE  
(SCENE\_REFL\_RED\_SZA30 BETWEEN 0.0487-0.01 AND 0.0487+0.01 AND SCENE\_REFL\_NIR\_SZA30 BETWEEN 0.2172-0.01 AND 0.2172+0.01)  
AND (SCENE\_REFL\_RED\_SZA45 BETWEEN 0.0372-0.01 AND 0.0372+0.01 AND SCENE\_REFL\_NIR\_SZA45 BETWEEN 0.1730-0.01 AND 0.1730+0.01)  
AND (SCENE\_REFL\_RED\_SZA50 BETWEEN 0.0458-0.01 AND 0.0458+0.01 AND SCENE\_REFL\_NIR\_SZA50 BETWEEN 0.1474-0.01 AND 0.1474+0.01);

LookUp Table

Climate, Carbon Models  
 $\alpha, F_{par}, g_c, Z_o, P_v, NP_v, \dots$   
 $F_i$  (LAD, LOP, BOP, h/w, LAI<sub>c</sub>, LAI<sub>b</sub>, CC)

Can Use Non-Spectral Data (e.g. slope, aspect, biome, time, GDD) to Further Constrain Solutions

Compute Parameter Statistics

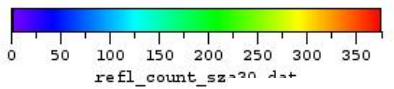
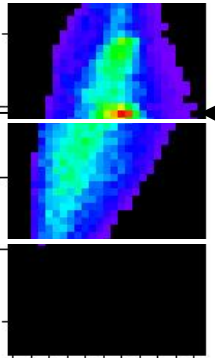
$LAD_i, LOP_i, BOP_i, h_i/w_i, LAI_{Ci}, LAI_{bi}, CC_i$

SIMPLE EXAMPLE  
USING GEOSAIL

Canopy vis' nir  
Canopy Parameters

70,000  
values

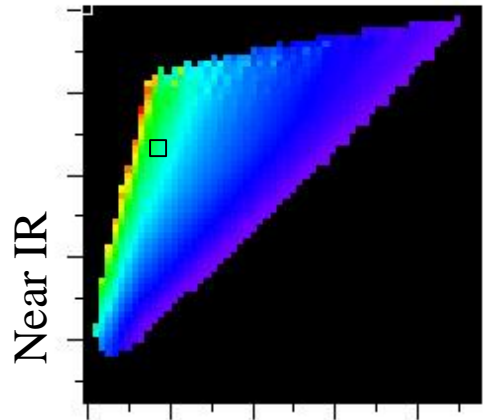
# Obs within a  
(±0.5%) vis' nir cell  
SZA = 30



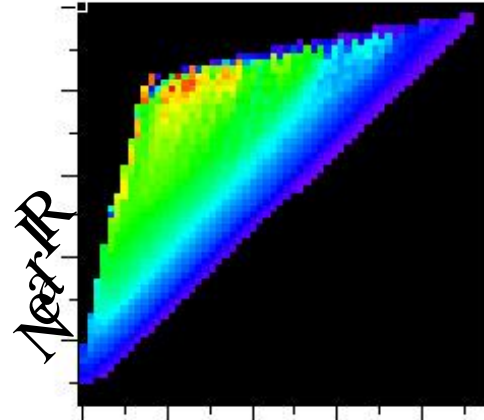
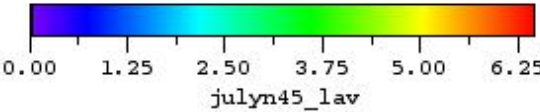
$LAI_{avg}$  within

Standard Deviation of  
 $LAI_{avg}$  within  
(±0.5%) vis' nir cells

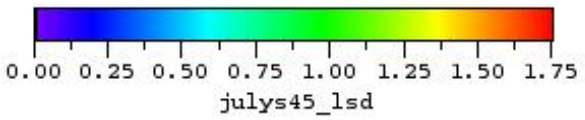
$LAI_i = LAI_{Ci} \times CC_i$



Visible Reflectance



Visible Reflectance





# Initial Retrieval Test

Landsat, vis, nir ( $\pm 0.01$ )

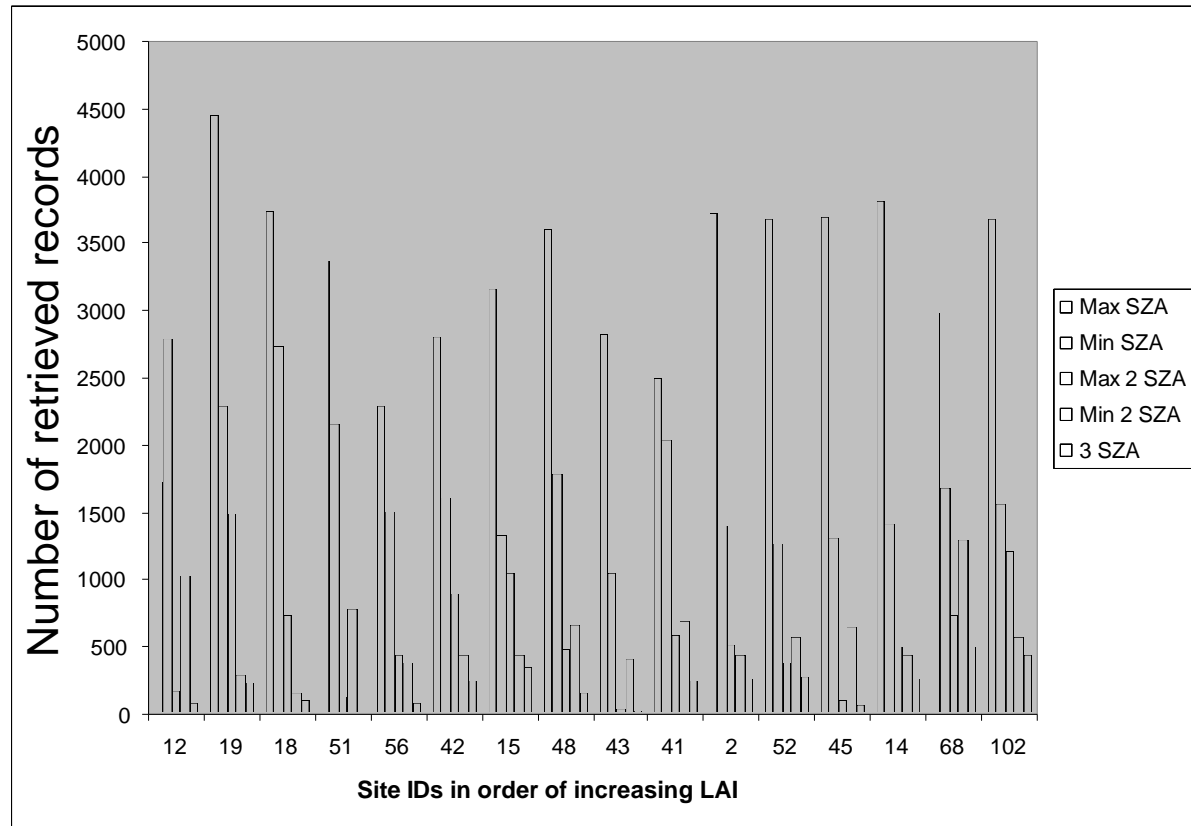
16 sites, 3 solar angles (3 dates)

1983 Superior National Forest

Site	Avg DBH (cm)	SD DBH (cm)	Stems per m <sup>2</sup>	Basal Fract	Avg BMI (kg/ m <sup>2</sup> )	SD BMI (kg/ m <sup>2</sup> )	Avg LAI	SD LAI	Tree Ht (ft)	Description
12	4.54	2.11	0.16413	0.00032	0.678	0.127	0.484	0.181	20	sparse, low black spruce
19	4.05	1.98	0.25564	0.00041	1.032	0.200	0.692	0.242	25	low black spruce
18	4.24	2.06	0.26062	0.00046	1.093	0.192	0.739	0.254	25	low black spruce
51	5.92	3.04	0.37799	0.00131	3.620	0.501	1.685	0.454	25	low density black spruce
56	8.09	4.29	0.25569	0.00168	5.280	0.345	1.834	0.329		medium density black spruce
102	5.92	2.34	0.91716	0.00292	7.246	0.736	3.670	1.228	50	high density black spruce
42	8.60	5.80	0.25863	0.00218	7.314	0.455	2.279	0.283	60	medium density black spruce
45	7.28	3.65	0.53387	0.00278	8.446	1.253	3.085	0.761	40	medium density black spruce
43	8.44	4.79	0.36305	0.00268	8.696	0.716	2.791	0.476	60	medium density black spruce
68	7.40	2.83	0.66002	0.00325	8.719	0.515	3.475	1.042	45	high density black spruce
48	9.83	3.19	0.33953	0.00285	9.149	1.665	2.700	0.795	50	medium density black spruce
52	9.92	3.91	0.35588	0.00318	10.036	0.574	3.034	0.588	60	low density black spruce
15	12.21	3.83	0.22381	0.00288	10.680	0.675	2.692	0.383	60	dense, mature black spruce
41	13.49	5.73	0.18303	0.00308	11.135	0.517	2.842	0.313	60	high density black spruce
2	14.52	4.43	0.17507	0.00317	12.378	0.830	2.884	0.340	50	dense, mature black spruce
14	13.22	4.13	0.24669	0.00372	13.643	0.587	3.266	0.427	60	dense, mature black spruce

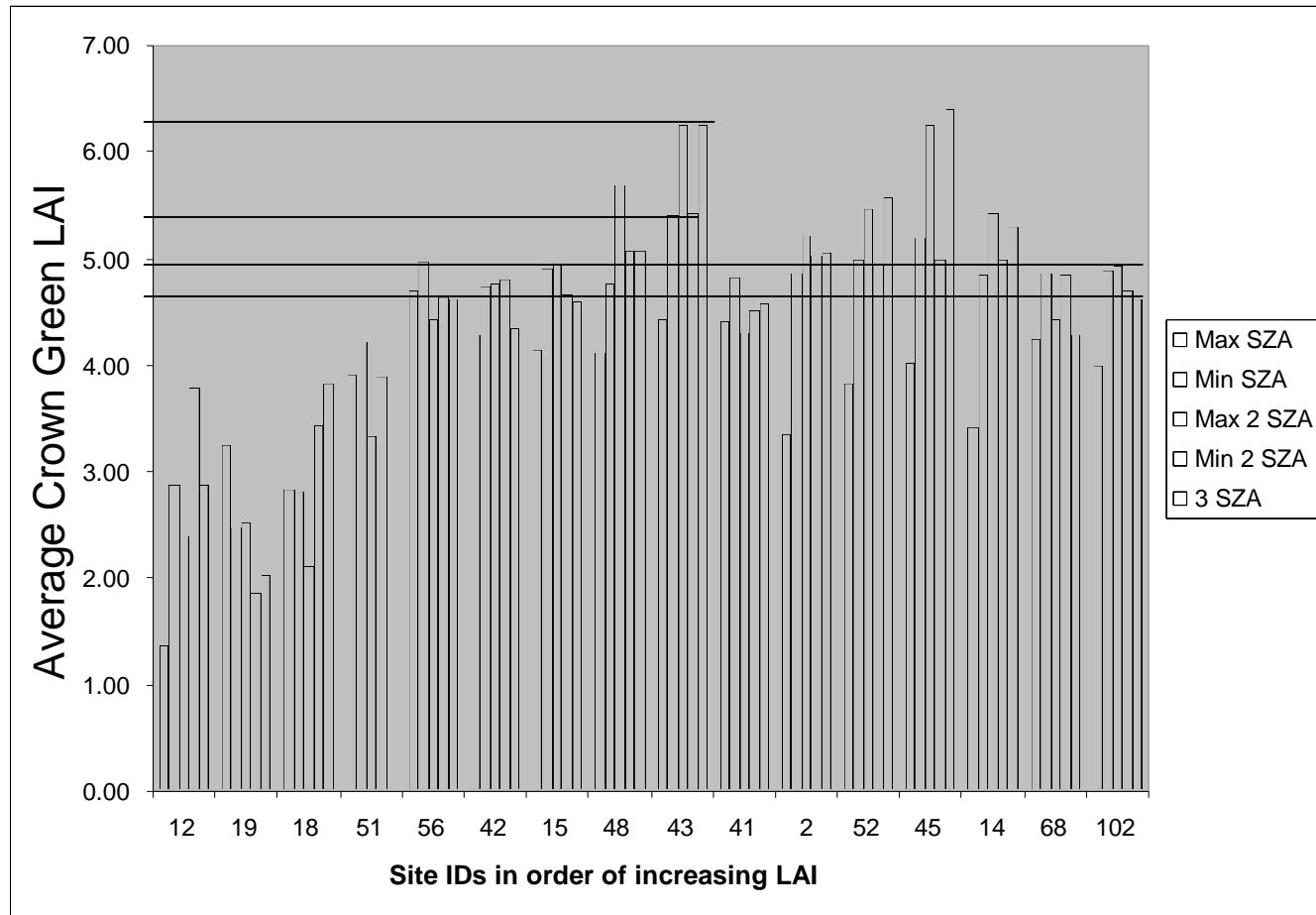
# Retrieval Test

## Landsat, vis, nir, 3SZA's Crown Green LAI



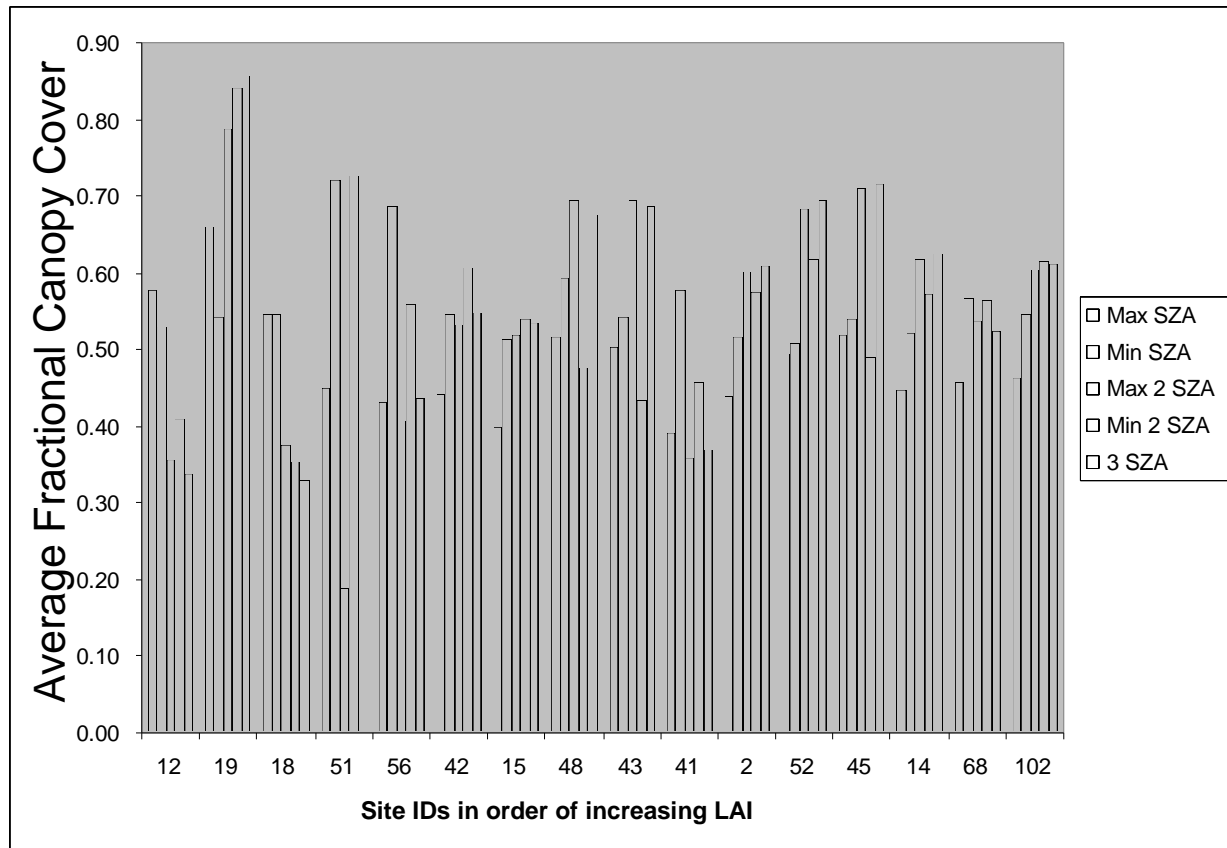
- Three SZA retrievals drastically reduced the number of records returned.
- For only one SZA, generally the maximum SZA returned more records.
- With two SZA either the low pair or the high pair returned the most records.

# Average Retrieved Crown Green LAI



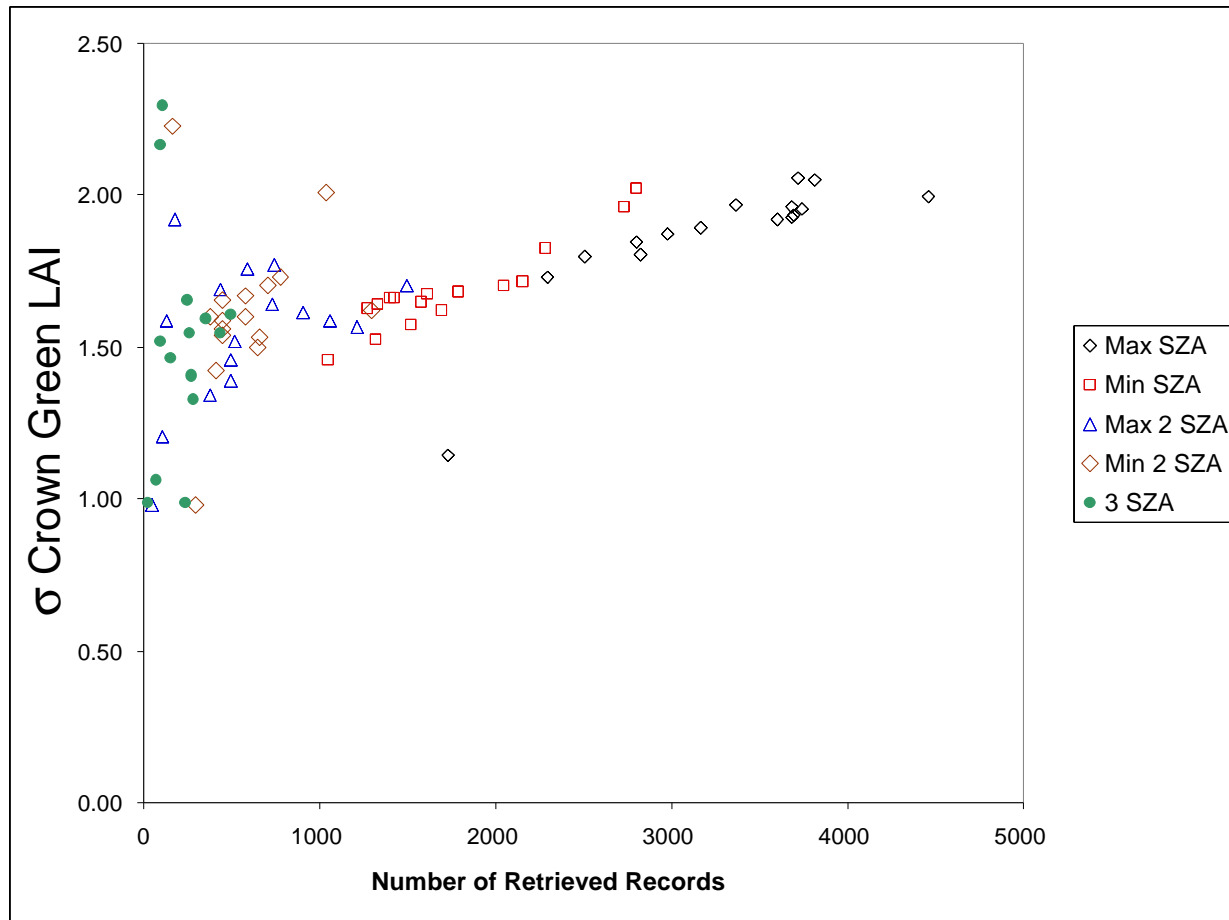
- Average Retrieved Crown Green LAI varied little using 2 or more SZAs.

# Average Retrieved Crown Cover Fraction

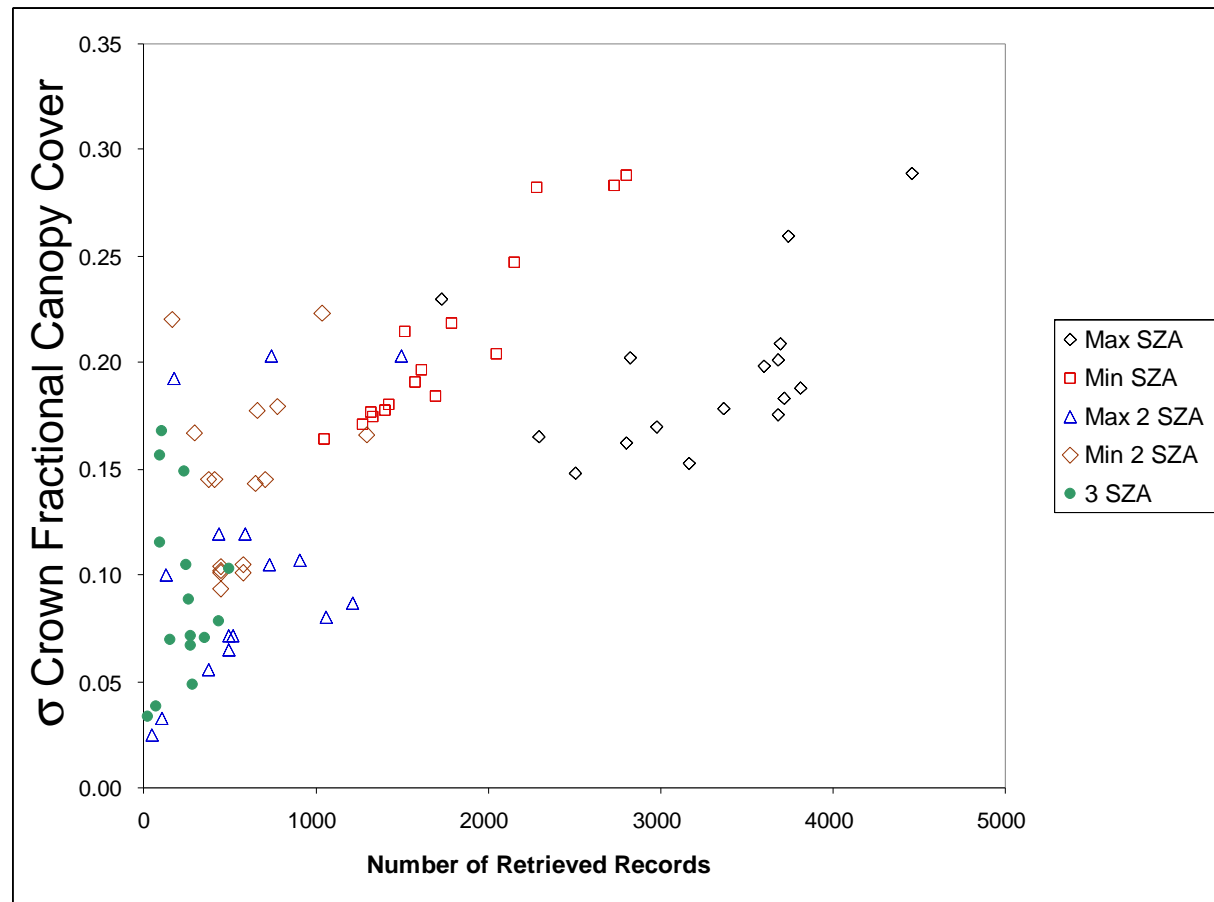


- Average Retrieved Crown Cover Fraction Sensitive to SZA

# Retrieval Precision vs Number of Records Retrieved-Crown Green LAI



# Retrieval Precision vs Number of Records Retrieved-Crown Cover



# Distribution of LAI Retrievals

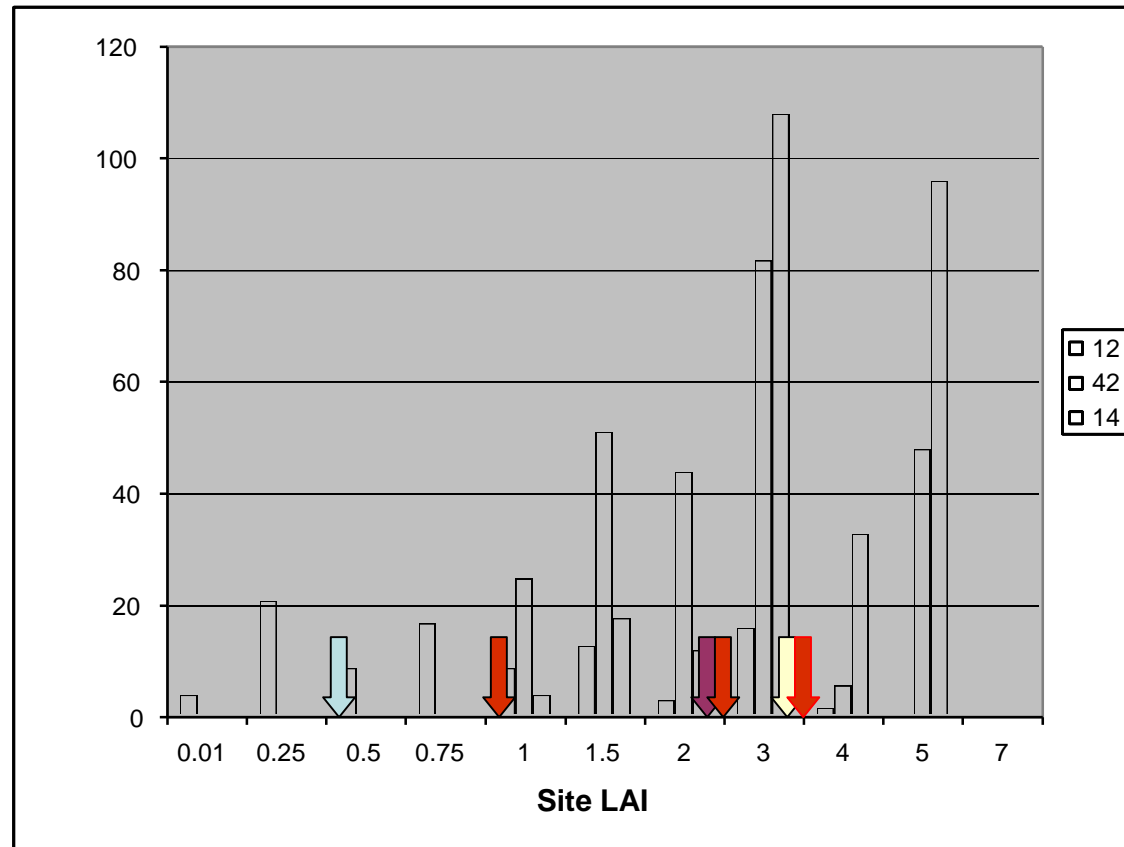
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# Retrieval Results

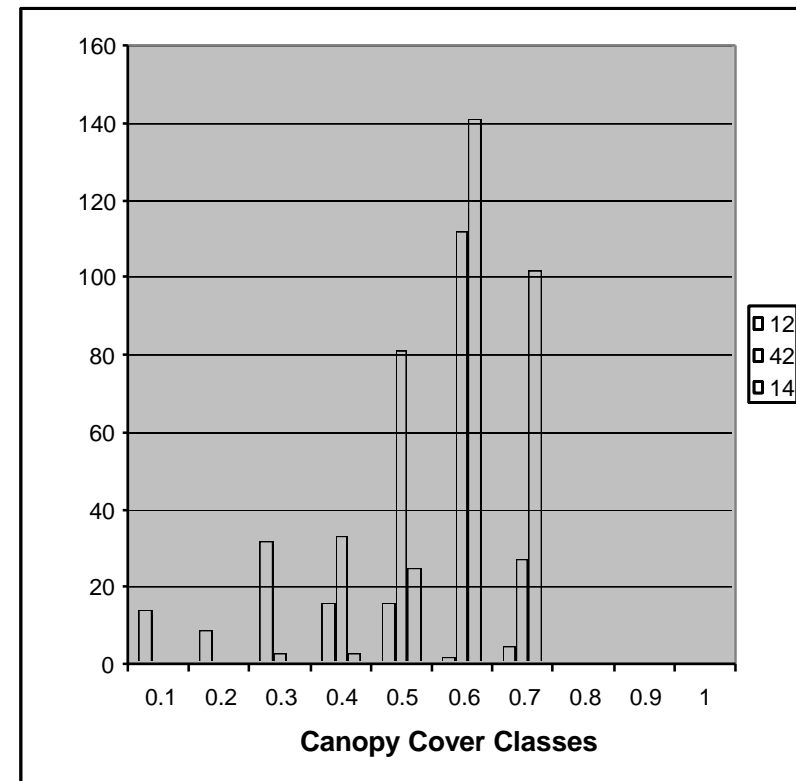
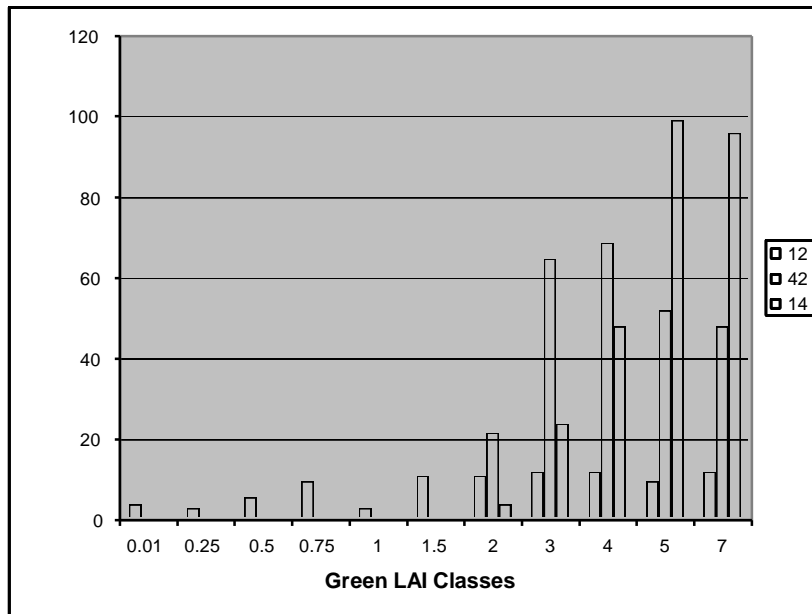
## Distribution of Site LAI Retrievals





# Retrieval Results

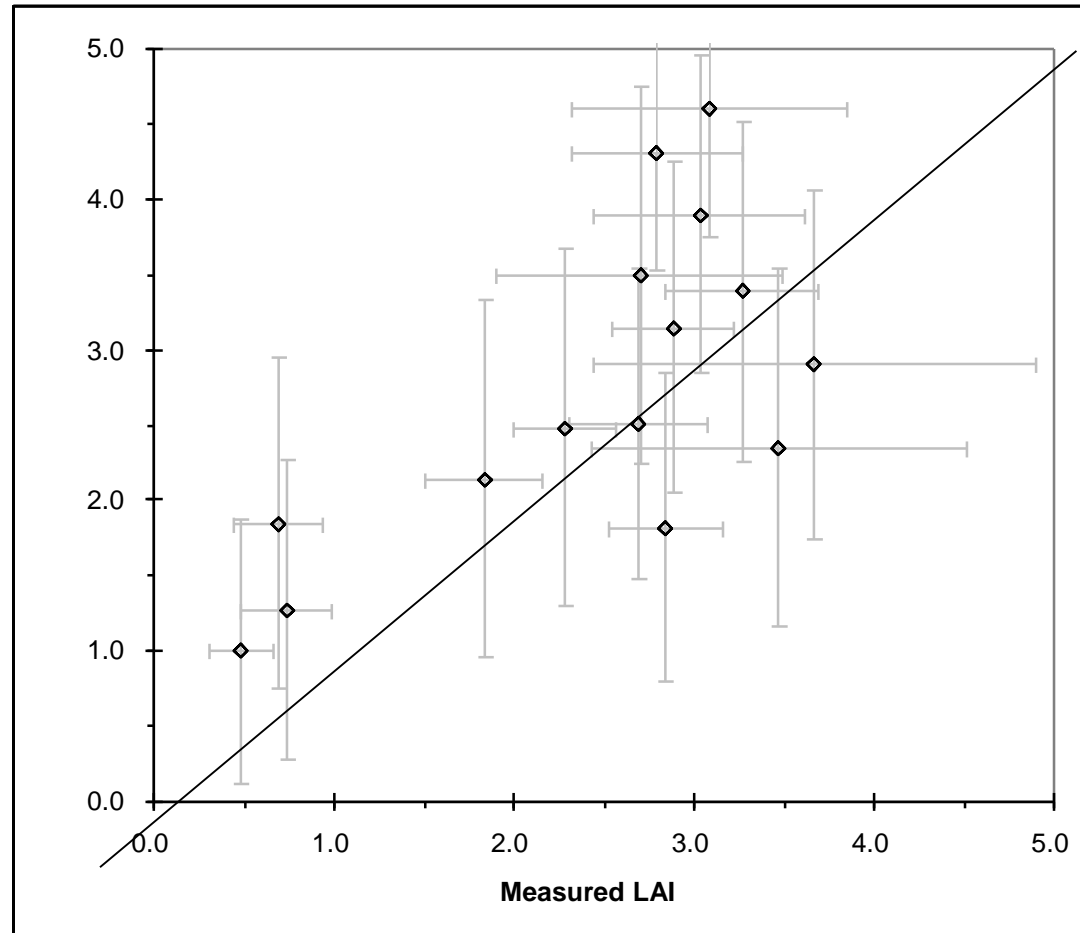
## Distribution of Canopy Component Retrievals - Crown LAI and CC



$$LAI = LC_i \times CC_i$$

# Retrieval Results

## All SNF Spruce Sites



Accuracies roughly equivalent to empirical approach  
Error structure automatically retrieved

## Problems

**Algorithm Locality:** Spectral change with illumination view geometry, background, season etc.affects robustness.

**“Chicken & Egg Syndrome”:** Radiance correction for view and illumination angle to infer biophysical parameters depends on the parameters.

**Parameter Inconsistency:** Different algorithms for different parameters.

**Error structures:** Error, uncertainty characterization approximate, empirical.

**Model utility:** Inability to directly assimilate satellite radiances in CWE models.

## BIOPHYS

## Solutions

**Algorithm Locality:** Physically-based retrievals incorporate reflectance responses to exogenous variables.

**“Chicken & Egg Syndrome”:** BIOPHYS uses spectral dependence on view and illumination angle to improve retrieval accuracies.

**Parameter Inconsistency:** BIOPHYS Parameter retrievals from a single algorithm using common input data.

**Error structures:** Error, uncertainty an integral part of the retrieval.

**Model utility:** BIOPHYS structure can assimilate data directly from within CWE models.

# BIOPHYS FRAMEWORK & OSSEs

- OBSERVING SYSTEM SIMULATION EXPERIMENT (OSSE)
  - OSSEs compute sensor/platform reqts in terms of biophysical parameter reqts.
- BIOPHYS FRAMEWORK provides a physically-based framework for computing
  - Retrieval precision as a function of sensor S/N
  - Retrieval bias as a function of sensor cal bias
  - Retrieval error as a function of band selection
  - Retrieval error as a function of CRT models

# Finally

- BIOPHYS Framework Can Bridge the Chasm
- What's Next
  - Wrap up exploratory work
    - Investigate convergence of mean to actual
  - Select RT Model (GOMS, GORT ...). Build the LUTs
  - Complete MODIS BIOPHYS Algorithm
  - Evaluate MODIS Products and Iterate
- PRODUCE PROVISIONAL DATA SETS WITH BIOPHYS OVER SELECTED STUDY REGIONS.
- PLACE PROVISIONAL DATA SETS ONLINE; INITIATE USER EVALUATION.

# The Chasm



**CANOPY RT  
MODELER**

Circa 1972- Modeling

Circa 1980 - Inversion

Circa 1996 Physically -  
Based Classification/MFM  
For Landsat



**Climate, Carbon  
Modeler**

BIOPHYS

2000 MODIS LAI , FPAR