





California State University MONTEREY BAY Extraordinary Opportunity

### Satellite Mapping of Crop Coefficients and Crop Water Requirements in California

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### Crop Coefficients: FAO-56









# **Annual Crops**





## Ground Cover: LIRF Corn 7/31/2008









Irrig Sci (2009) 28:17-34 DOI 10.1007/s00271-009-0182-z

ORIGINAL PAPER

#### Estimating crop coefficients from fraction of ground cover and height

Richard G. Allen · Luis S. Pereira

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**Abstract** The FAO-56 procedure for estimating the crop coefficient  $K_c$  as a function of fraction of ground cover and crop height has been formalized in this study using a density coefficient  $K_d$ . The density coefficient is multiplied by a basal  $K_c$  representing full cover conditions,  $K_{cb}$  full, to produce a basal crop coefficient that represents actual conditions of ET and vegetation coverage when the soil surface is dry.  $K_{cb}$  full is estimated primarily as a function of crop height.  $K_{cb}$  full can be adjusted for tree crops by multiplying by a reduction factor ( $F_x$ ) estimated using a mean leaf stomatal resistance term. The estimate for basal

#### Introduction

The two-step crop coefficient  $(K_c) \times$  reference evapotranspiration (ET<sub>ref</sub>) method has been a successful and dependable means to estimate evapotranspiration (ET) and crop water requirements. The method utilizes weather data to estimate ET for a reference condition and multiplies that estimate by a crop coefficient that represents the relative rate of ET from a specific crop and condition to that of the reference. The reference condition is generally ET from a clipped\_cool season\_well-watered\_grass (ET\_) or from a





#### Development of Reflectance-Based Crop Coefficients for Corn

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#### ABSTRACT

noncurrent measurements of reflected canopy radia-Ution and the basal crop coefficient (K<sub>cb</sub>) for corn were conducted throughout a season in order to develop a reflectance-based crop coefficient model. Reflectance was measured in Landsat Thematic Mapper bands TM3 (0.63 - 0.69 um) and TM4 (0.76 - 0.90 um) and used in the calculation of a vegetation index called the normalized difference (ND). A linear transformation of the ND was used as the reflectance-based crop coefficient (K<sub>cr</sub>). The transformation equates the ND for dry bare soil and the ND at effective cover, to the basal crop coefficient for dry soil evaporation and at effective cover, respectively. Basal crop coefficient values for corn were obtained from daily evapotranspiration measurements of corn and alfalfa, using hydraulic weighing lysimeters. The Richards growth curve function was fitted to both sets of data. The K<sub>cb</sub> values were determined to be within -2.6% and 4.7% of the K<sub>er</sub> values. The date of effective cover obtained from the K<sub>cb</sub> data was within four days of the date on which the ND curve reached its maxima according to the Richards function. A comparison of the K<sub>er</sub> with basal crop curves from the literature for several years of data indicated good agreement. Reflectancebased crop coefficients are sensitive to periods of slow and fast growth induced by weather conditions, resulting in a real time coefficient, independent from the traditional time base parameters based on the day of planting and effective cover.

#### INTRODUCTION

Crop coefficients and calculated reference crop

soil conditions from those under which they were developed. However, these coefficients provide an inexpensive and practical method for estimating actual crop ET throughout a growing season with reasonable accuracy for scheduling irrigations.

The use of reflected canopy radiation to obtain a real time reflectance-based basal crop coefficient for corn has previously been studied by Neale and Bausch (1983) and Bausch and Neale (1987). Those studies compared Wright's (1982) basal crop coefficient for corn with a transformation of the normalized difference (Deering et al., 1975), calculated using remotely sensed reflected canopy radiation data obtained throughout a growing season. One hypothesis in those comparisons was that the normalized difference curve reached its maxima close to the date of effective cover on the basal crop coefficient curve. However, no evapotranspiration data were available at that time to estimate basal crop coefficients.

The objective of this paper is to show the feasibility of using remote sensing methodology to estimate crop coefficients for corn by comparing simultaneous field estimates of corn basal crop coefficients and reflectancebased crop coefficients. A comparison of the reflectancebased crop coefficients with coefficients from the literature will also be made for several years of data.

#### BACKGROUND

#### **Agronomic Variables and Evapotranspiration**

Specific factors that influence evapotranspiration of plants include plant species, percent canopy cover, plant population, row spacing and orientation, rooting depth and extent, stage of growth, light reflection, and soil moisture availability (Gates and Hanks, 1967). Penman

# Satellite Irrigation Management Support Project: TOPS - SIMS

#### **Processing Steps**

- At sensor radiance
- LEDAPS
- Surface reflectance
- NDVI
- Fractional ground cover
- K<sub>cb</sub>
- $ET_{cb} = K_{cb} * ET_{ref}$











Landsat (TM / ETM+ / OLI) 30m / 0.25 acres Overpass every 8-16 days Sentinel-2A 20m / 0.1 acres Sentinel 2B launch in 2017

# Satellite Irrigation Management Support Project: TOPS - SIMS

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### **ET**<sub>ref</sub>: CIMIS / Spatial CIMIS



CIMS Station Data, 1982 to present

Spatial CIMIS, 2003 to present

# Satellite Irrigation Management Support Project: TOPS - SIMS

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### Satellite Irrigation Management Support (SIMS) Web Services





#### **TOPS Satellite Irrigation Management Support**



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# Satellite Irrigation Management Support (SIMS) Web Services





#### **Accuracy Assessment Field Campaign**





Inflow

Soil Water

Deep Perc

Water Balance and SEB 15 crops 30 sites



**Eddy Covariance** 

#### **SIMS Accuracy Assessment**





- Field validation campaign completed in partnership with partner growers, CA DWR, CSU Fresno, USDA ARS and UC Davis.
- Data collected for 15 crops at 30 sites using eddy covariance, surface energy balance residual, soil moisture sensor networks.
- Results highly encouraging for seasonal and daily comparisons.
  - 11% mean absolute error for seasonal ETcb vs measured ETa (via eddy covariance / surface energy balance residual; n = 12)
  - 14.2% mean absolute error for seasonal ETcb vs measure ETa (via soil water balance; n = 23) → reduced to 8.7% for well-irrigated crops only (n = 16)

### Average Observed ET<sub>rf</sub> by Crop Type for CA Delta



- SIMS close to median of model ensemble for most crop types
- SIMS close to mid-point between two versions of METRIC run by two different expert groups (ITRC and UC Davis)

# **Monitoring Crop Development**



NASA Official: Ramakrishna R.Nemani

Curator: Forrest Melton

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#### Southern San Joaquin Valley, CA, Field Boundary Polygons



#### Application Program Interface (API) for Integration with Other Irrigation Management Software

CropM	lanage					
Planting Home	Ranch Home	Edit Ranch	Ranch List	Site Administration -	Help	
Planting: romai	CCE Ranch 3, Lot ne 2, 10.0 acres 2 row, 40 inch bec					

#### Irrigation Summary

Show / H	lide Columns				Reset Column O	rder	Show Previ	ous Columns	Show Next Columns
Water Date	Irrigation Method	Recommended Irrigation Interval (days)	Recommended Irrigation Amount (Inches)	Recommended Irrigation Time (hours)	Irrigation Water Applied (inches)	Ke	Canopy Cover (%)	Average Reference ET (inches/day)	Total Crop ET (inches)
6/4/15	Germination Sprinkler	N/A.	NA	NGA	0.75 m	0.00	0	0.00	0.00
6/5/13	Germination Sprinkler	1.6	0.22 in	0.72 hrs	0.45 in	1.00	0	0.14	0.14
6/7/13	Germination Sprinkler	1.9	0.36 in	1.18 hrs	0.30 in	0.70	0	0.17	0.23
6/9/13	Germination Sprinkler	1.7	0.39 in	1.29 hrs	0.45 in	0.70	0	0.18	0.25
6/12/13	Sprinkler	3.1	0,28 in	0.95 hrs	0.30 in	0.48	t.	0.15	0.21
6/16/13	Sprinkter	2.9	0.40 in	1.33 hrs	0.45 in	0.37	t	0.20	0.30
Totals			1.64 in	5.47 hrs	2.70 in				1.13 in

# **SIMS Earth Engine Implementation**

#### Advantages:

- Leverage satellite image archives and data services on Google Earth Engine
- Reduce operational costs
- Increase scalability to other regions





# Limitations of the SIMS / Reflectance Approach

- Requires local ET<sub>ref</sub> (Weather station network)
- Provides ET<sub>cb</sub> must add in soil evaporation to get ET<sub>p</sub>
- Must estimate crop stress to get ET<sub>a</sub> (e.g., via soil water balance)

### **Advantages of the SIMS / Reflectance Approach**

- Reflectance data freely available from multiple satellites (e.g., Landsat 7/8, Sentinel-2A) → <u>operational reliability & data continuity</u>
- Reflectance data available at field scales (30 m)
- Relatively simple calculations fully automated computations
- NDVI/Fc/K<sub>cb</sub> interpolates well between point measurements
- Extensible framework for satellite data processing
- ET<sub>cb</sub> represents biological demand for water by the plant

Combination of energy balance (e.g., METRIC) and reflectance (e.g., SIMS) approaches provides robust, long-term strategy for sustaining operational use.



#### TOPS Satellite Irrigation Management Support



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