

# MAPPING THREATS TO AGRICULTURE IN EAST AFRICA: PERFORMANCE OF MODIS DERIVED LST FOR FROST IDENTIFICATION IN KENYA'S TEA PLANTATIONS

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# INTRODUCTION

- Increased prevalence of weather related hazards in eastern Africa including drought, floods, hail and frost threatening agricultural productivity.
- Kenya is heavily dependent on agriculture for economic growth (FAO 2013)
  - Agriculture contributed 23.5% and 21.5% of GDP in 2009 and 2010 respectively
  - Employment to half a million households of smallholders and 150,000 on large tea estates
- Tea growing in Kenya depends on stability of the weather (Cheserek et al 2015)
  - Weather is unpredictable
  - Frost has contributed 30% of tea leaf losses
  - Drought has contributed 14-30%
  - The losses are experienced between January and march -  
frost and dry season

**“Inadequate knowledge and lack of early warning systems”**

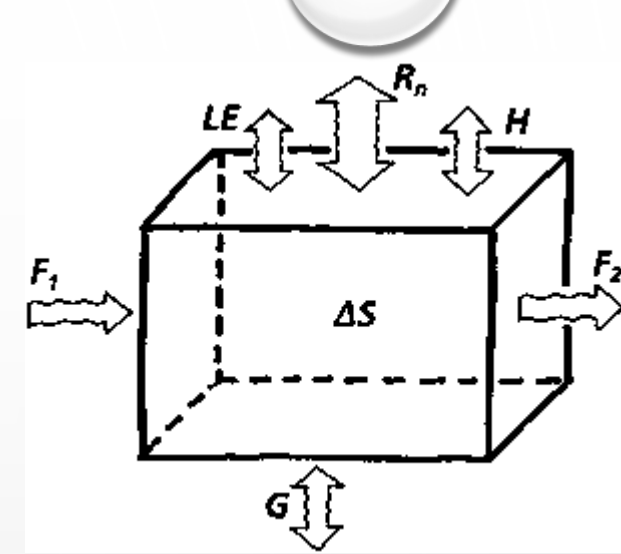
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# INTRODUCTION

- ❖ Frost has widely been defined in relation to the occurrence of temperatures of 273K (0°C) or below, as measured within an appropriate weather shelter at a height of between 1.25 and 2.0 m from the ground surface (Snyder, R. L. et al 2005)
- ❖ Casually used to describe a meteorological condition involving low temperatures when plants are damaged
- ❖ Two categories of frost (Richard L. Snyder, 2000)
  - ❖ Advective
  - ❖ Radiative
- ❖ Radiative frost is most common in Kenya

# FROST MITIGATION

- During a radiative frost night, falling temperatures result from:
    - Decreasing sensible heat of the air when:
      - The sum of
        - Sensible heat transfer downward from the air,
        - Soil heat flux upward to the soil surface, and
        - Transfer of heat within the vegetation to the plant surface
      - Is insufficient to
        - Replace the sensible heat content losses resulting from net radiation energy losses
    - Therefore,
    - Goal of any frost damage prevention mechanism is to manipulate energy transfer processes of:
      - Conduction through the soil to the surface,
      - Horizontal convection through the air and
      - Downward radiation from the sky to reduce the net energy loss
- “Any factor that affects these processes affects frost occurrence”**



net radiation ( $R_n$ ),  
sensible heat flux ( $H$ ),  
latent heat flux ( $LE$ ),  
soil heat flux or conduction ( $G$ ),  
sensible and latent energy  
advection in ( $F_1$ ) and out ( $F_2$ ),  
and energy storage in the crop  
( $\Delta S$ )

# FROST ASSESSMENT IN KENYA

- Based on mapping and classifying MODIS LST values into levels to represent light to severe frost risk.
  - Proven useful in indicating areas affected by frost
  - Concern on underestimation of lighter frosts
- Basing on MODIS LST alone, frost tend to occur at higher LST ranging 260K to 288K (-13<sup>0</sup>C to 15<sup>0</sup>C).
- While this range may be calibrated to indicate frost occurrences, the occurrence of these temperatures does not always correspond to frost occurrence and crop damage.
- Studies have indicated **frost occurrence to be a combination of microclimatic factors including topography, weather conditions and land surface characteristics** affecting night minimum LST which is the major determinant (Blennow, K, 1998; Lindkvist, L 2000).

# MAIN GOAL AND RESEARCH QUESTIONS

➤ The main purpose of this research is to establish how well MODIS derived LST data, specifically the MYD11A1 product, performs in **discriminating** between frost and no frost locations.

The research questions are:

1. What is the **accuracy** of the currently operational frost identification model and the LST thresholds in defining frost occurrence LST in the Kenyan tea zone?
2. What is the **overall performance** of MODIS derived LST product in discriminating between frost and no frost zones within the Kenyan tea zone?
3. What is the **role** of remotely sensed data in facilitating agricultural monitoring in regions with **sparse network** of ground observation stations?



# STUDY AREA

- Focus on the **tea growing** regions of Kenya.
- **Sampling frame** - all major tea plantations
- As of the year 2010;
  - ☐ Total tea growing area 171,916 hectares
    - Large plantation farms 56,893 hectares
    - Smallholder farms 115,023 hectares
- The sampled farms are about;
  - 30 % of all the tea growing area
  - 60% of the sampling frame.

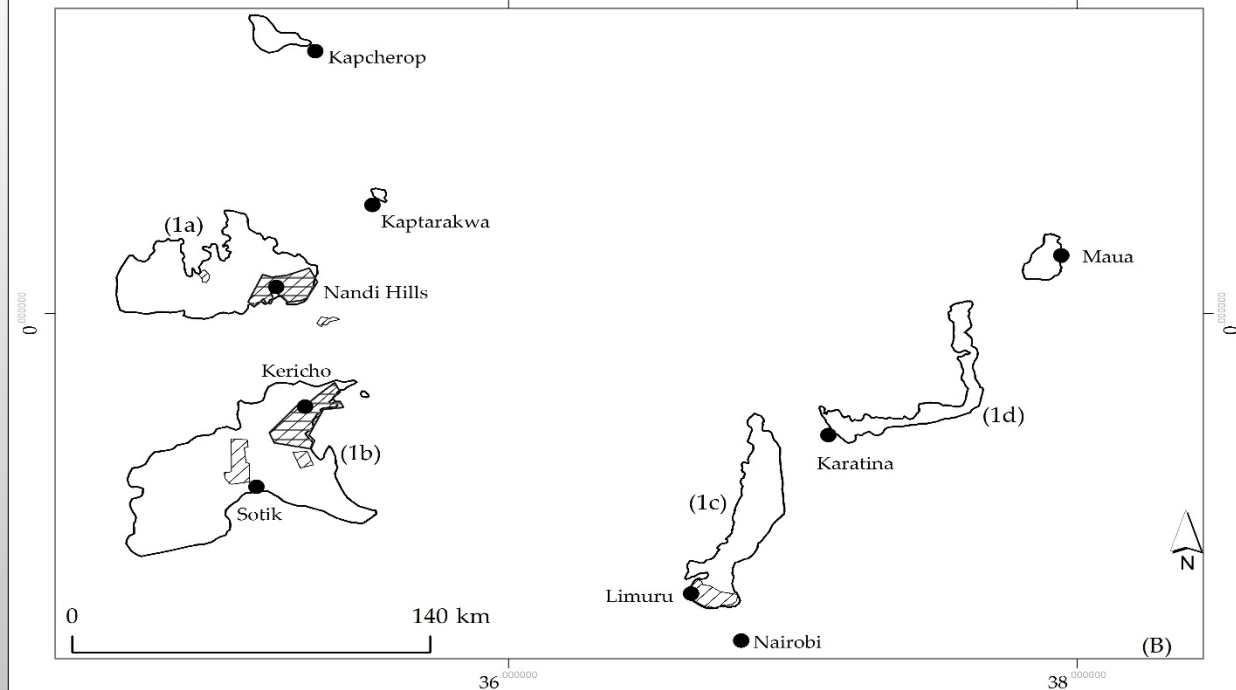
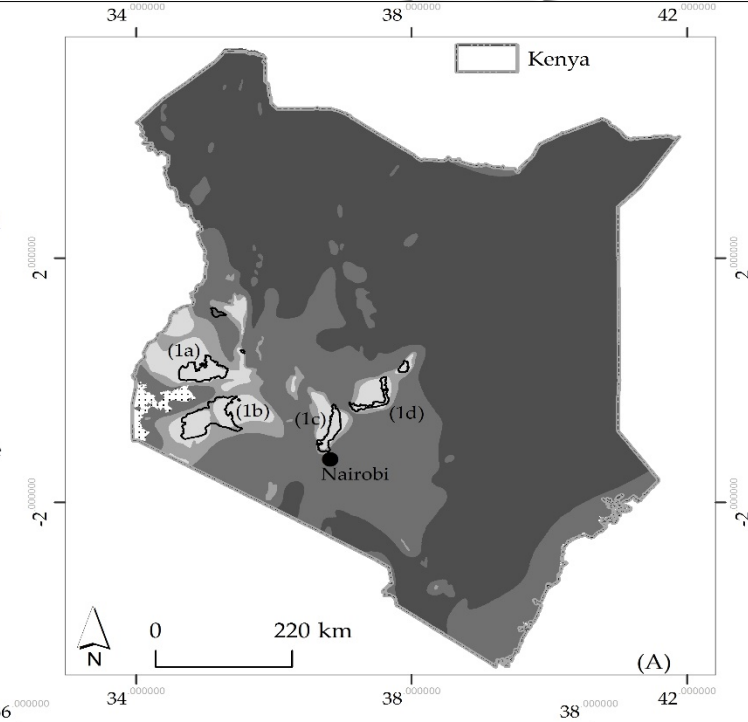
Agro-ecological zones

- I Humid
- II Sub-humid
- V/IV/III Semi-humid to Semi-arid
- VI/VII Semi-arid, arid to Very-arid

- All tea growing zones
- Large scale tea plantations
- Sampled plantations

All tea growing zones includes both large scale plantations and smallholder farms. Sampled plantations are within the large scale plantations

- Towns
- Lake Victoria



# DATA

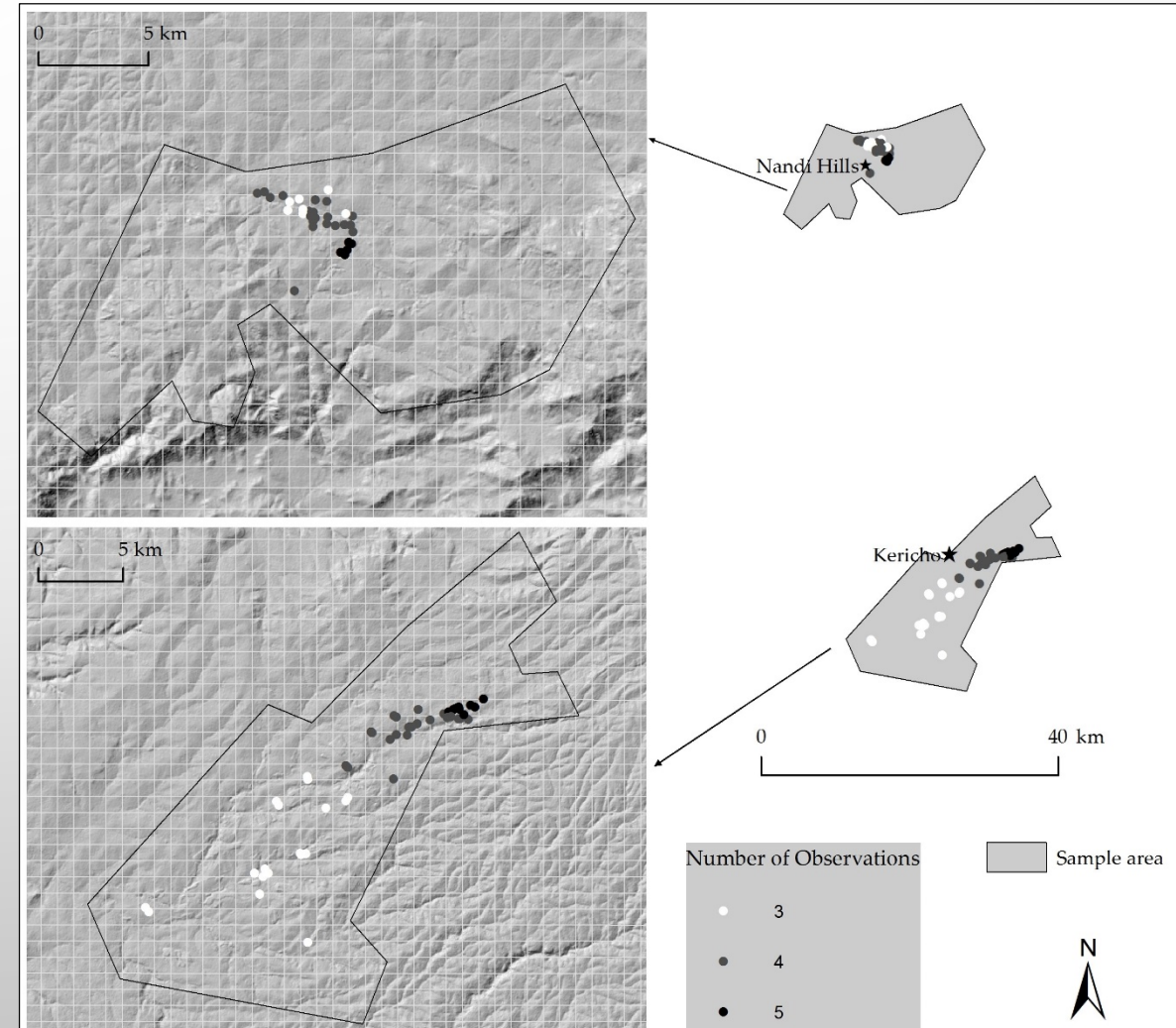
## 1. Frost and No frost observations

Not all collected points were used

	Total Observations	Used Observations
Frost	215	190
No frost	121	92
	336	282

## 2. MODIS derived LST

- Aqua satellite
- Daily night product MYD11A1 LST
- 1km spatial resolution
- 1:30 am equatorial overpass time





# METHODS

- Frost likelihood categories were reclassified into two groups:
  - Frost
  - No frost.
- 6 different cutoff thresholds were tested.
- Depending on the cutoff threshold, frost locations were coded as 1 and no frost was coded 0.

		Reclassified thresholds					
MODIS LST	Frost Likelihood	Baseline	1	2	3	4	5
>288K	No Frost	No Frost	No Frost	No Frost	No Frost	≤273K=Frost >273K=No Frost	≤278.5K=Frost >278.5K=No Frost
284-288K	Minor Frost Pockets	Frost					
281-284K	Moderate Frost		Frost				
277-281K	Severe Frost		Frost				
<277K	Very Severe Frost		Frost				

# ACCURACY ASSESSMENT

## Overall accuracy of each threshold

- "OVERALL ACCURACY =  $(TP + TN) / (TP + TN + FP + F) = (TP + TN) / N$

## Overall accuracy of MODIS LST

### ■ RECEIVERS OPERATING CHARACTERISTICS (ROC) CURVE

- Allows the **threshold to vary** and measures the changes in true and false positive fractions .
- A roc curve is used as an estimate of **overall performance** of the predictor (MODIS derived LST) to **discriminate** between occurrence and non-occurrence and is described by the true positive rate (TPR) and the false positive rates (FPR)

- **True Positive Rate (Sensitivity) =  $TP / TP + FN$**
- **Specificity =  $TN / FP + TN$**
- **False Positive Rate (1 – Specificity) =  $FP / FP + TN$**

# RESULTS

## Overall Accuracy of each Threshold

Threshold	Cutoff point (K)	True Positives	False Positives	True Negatives	False Negatives	Overall Accuracy
<b>Baseline</b>	288	190	92	0	0	67.38%
<b>1</b>	284	190	92	0	0	67.38%
<b>2</b>	281	190	98	4	0	68.79%
<b>3</b>	277	108	30	62	82	60.28%
<b>4</b>	273	84	26	66	106	53.19%
<b>5</b>	278.5	156	45	47	34	71.99%

# RESULTS

## Overall Accuracy of MODIS derived LST

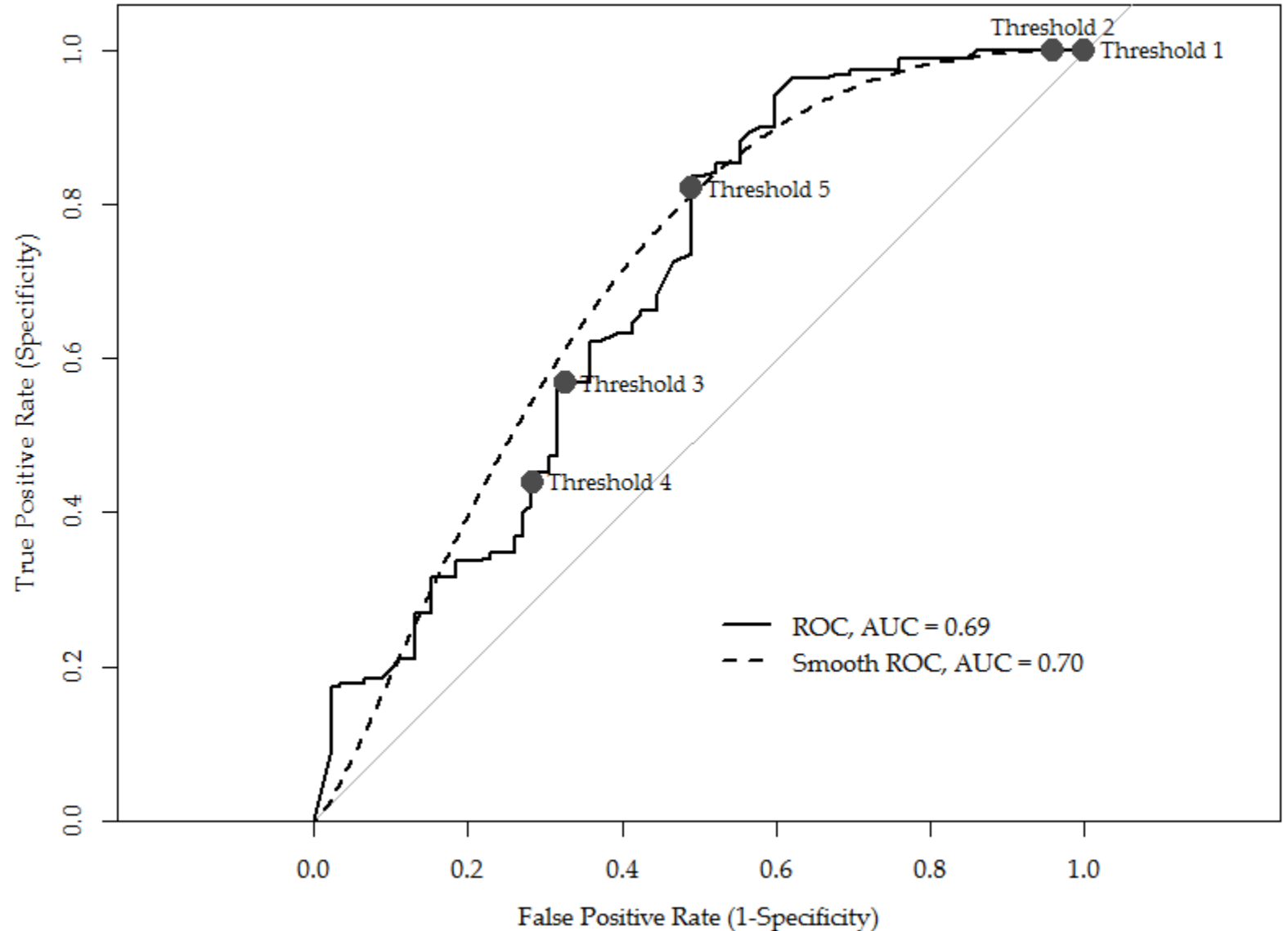
Threshold	Cutoff point (K)	Predicted Frost events (Count)	Predicted No Frost events (Count)	Sensitivity (True Positive Rate)	Specificity	1-Specificity (False Positive Rate)
Baseline	288	282	0	1	0	1
1	284	282	0	1	0	1
2	281	278	4	1	0.04	0.96
3	277	138	144	0.568	0.674	0.326
4	273	110	172	0.44	0.717	0.283
5	278.5	201	81	0.821	0.511	0.489

# RESULTS

## Overall Accuracy of MODIS LST

- Performance of **0.69** on a scale of 0.5 to 1
- A **poor Performance**
  - 1 - perfect performance
  - 0.5 - worst performance
- At 95% confidence level, an
  - AUC confidence interval of 0.63-0.77

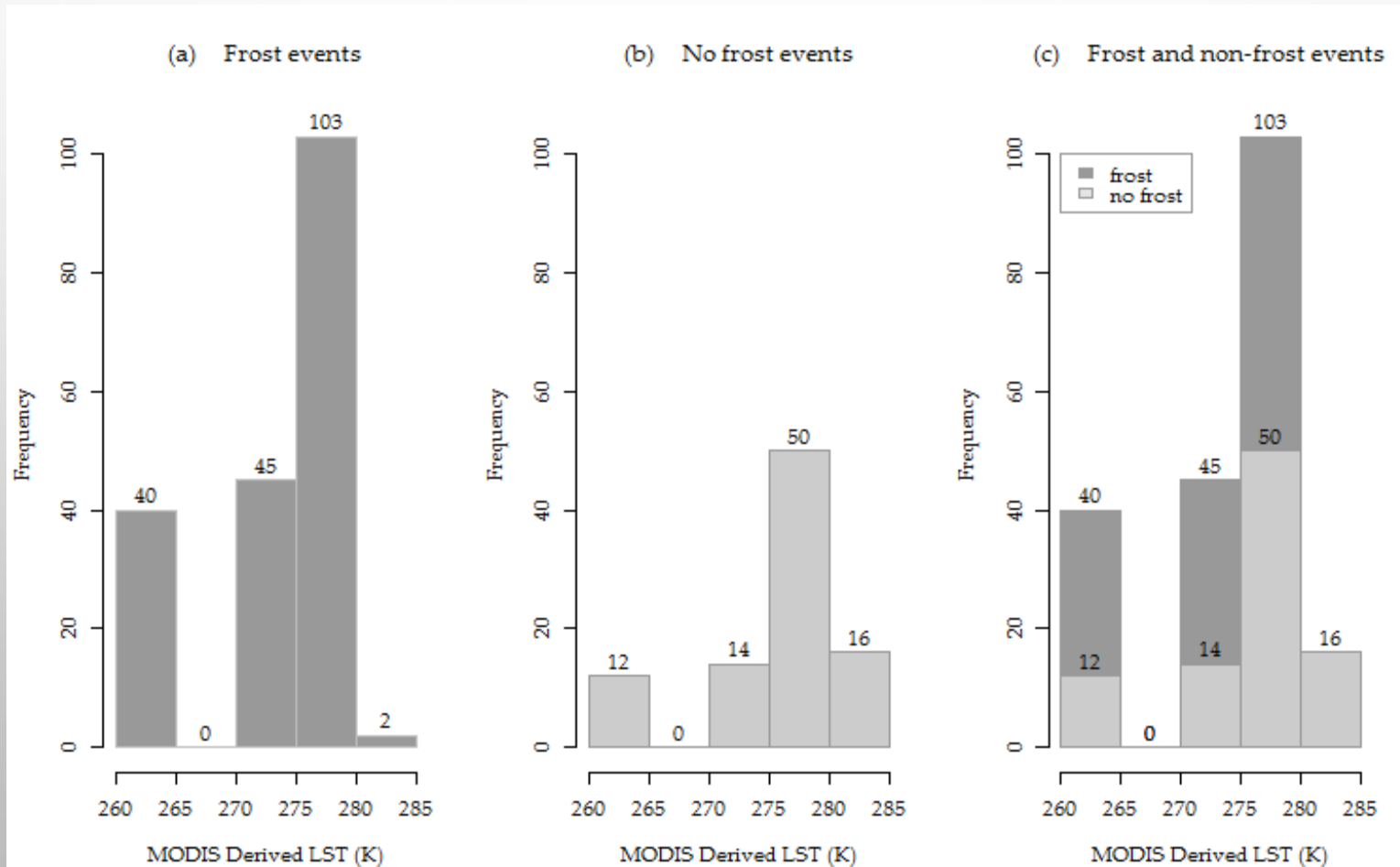
ROC Curve for MODIS LST Model





# SUMMARY OF FINDINGS

- Frost can occur at LST values higher than 273K
- MODIS LST may be useful in indicating frost zones
- MODIS LST cannot discriminate between frost and no frost with high accuracy



# IMPLICATIONS

- MODIS derived LST may be used in conjunction with other variables to guide low cost farmers' decisions on:
  - Pruning tea bushes to avert further crop damage
  - Picking ready tea leaves to cut down on losses
- The findings inform further research on frost and tea
- Use of remote sensing based data in agricultural monitoring is illustrated.

# POSSIBLE SOURCES OF UNCERTAINTIES

- Other factors affecting frost occurrence
- Possible sample and sampling errors
- Spatial variation of LST in mountainous areas
- Frost and no frost observations within single MODIS pixel
- Minimum LST and MODIS overpass times
- MODIS LST representation of on the ground LST

# CONCLUSIONS

- MODIS LST alone is not a good predictor of frost occurrence
- Further research on characterization of frost occurrence in this region is strongly recommended in order to improve the accuracy of frost prediction.

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**THANK YOU**