



OFFICE OF RESEARCH & DEVELOPMENT

**2012** R&D  
REVIEW

# Development of a Rail Temperature Prediction Model Based on Heat Transfer Principles



U.S. Department  
of Transportation  
Federal Railroad  
Administration

LEITH AL-NAZER  
Office of Research and Development  
Office of Railroad Policy and Development

# Program Area & Risk Matrix

Development of Rail Temperature Prediction Model Based on Heat Transfer Principles

Program Areas	Risk Factors	Trespass	Grade Crossing	Derailment	Train Collision	All Other Safety Hazards
Railroad Systems Issues						
Human Factors						
<b>Track &amp; Structures</b>				<b>X</b>		
Track & Train Interaction						
Facilities & Equipment						
Rolling Stock & Components						
Hazardous Materials						
Train Occupant Protection						
Train Control & Communications						
Grade Crossings & Trespass						

# Outline

- Background
  - Current industry practices
- Research motivation and objectives
- Research approach and implementation
- Lessons Learned
- Current status
  - Web-based application

# Background

- Due to thermal expansion, risk of buckling exists on track with continuously welded rail (CWR)
- 2005 to 2009: 124 incidents, \$43 million in damages
- Railroads issue slow speed orders on hot days in order to prevent and minimize damages due to buckling incidents

# Background



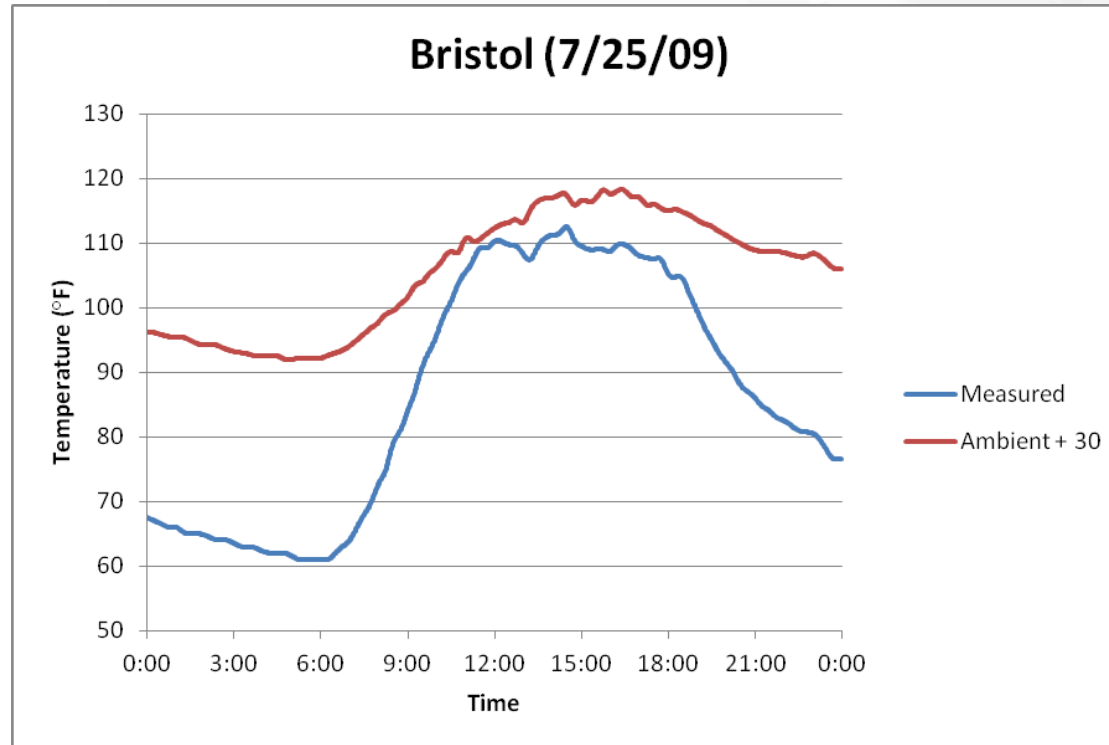
# Current Industry Practice

- Empirical model currently used by railroad industry to predict maximum daily rail temperature:

$$\left( \begin{array}{c} \text{Max Forecasted} \\ \text{Rail Temp.} \end{array} \right)^{\circ\text{F}} = \left( \begin{array}{c} \text{Max Forecasted} \\ \text{Ambient Temp.} \end{array} \right)^{\circ\text{F}} + \text{Constant}$$

- Constant is above equation, may vary but is typically 30 °F
- Model is accurate only for predicting the maximum daily rail temperature

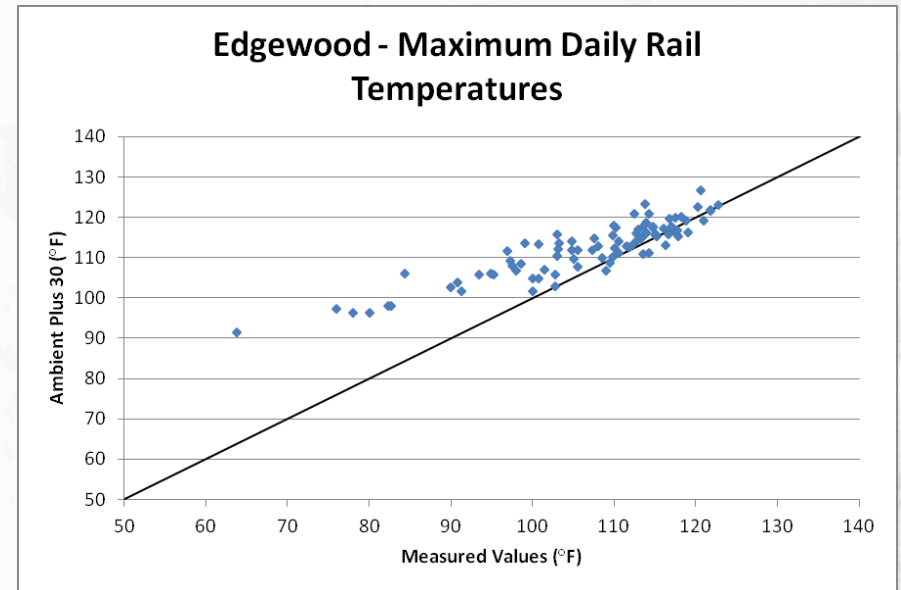
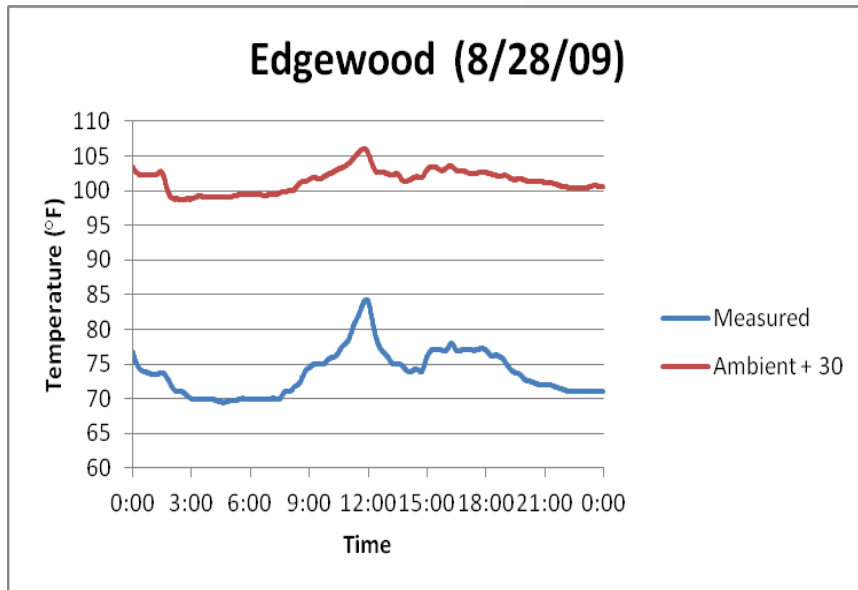
# Research Motivation



**Empirical model used by industry does not provide information on how rail temperatures change throughout an entire day**

# Research Motivation

- Empirical model used by industry is only useful for predicting the maximum daily rail temperature on hot days
- Over-predicts on cool / overcast days



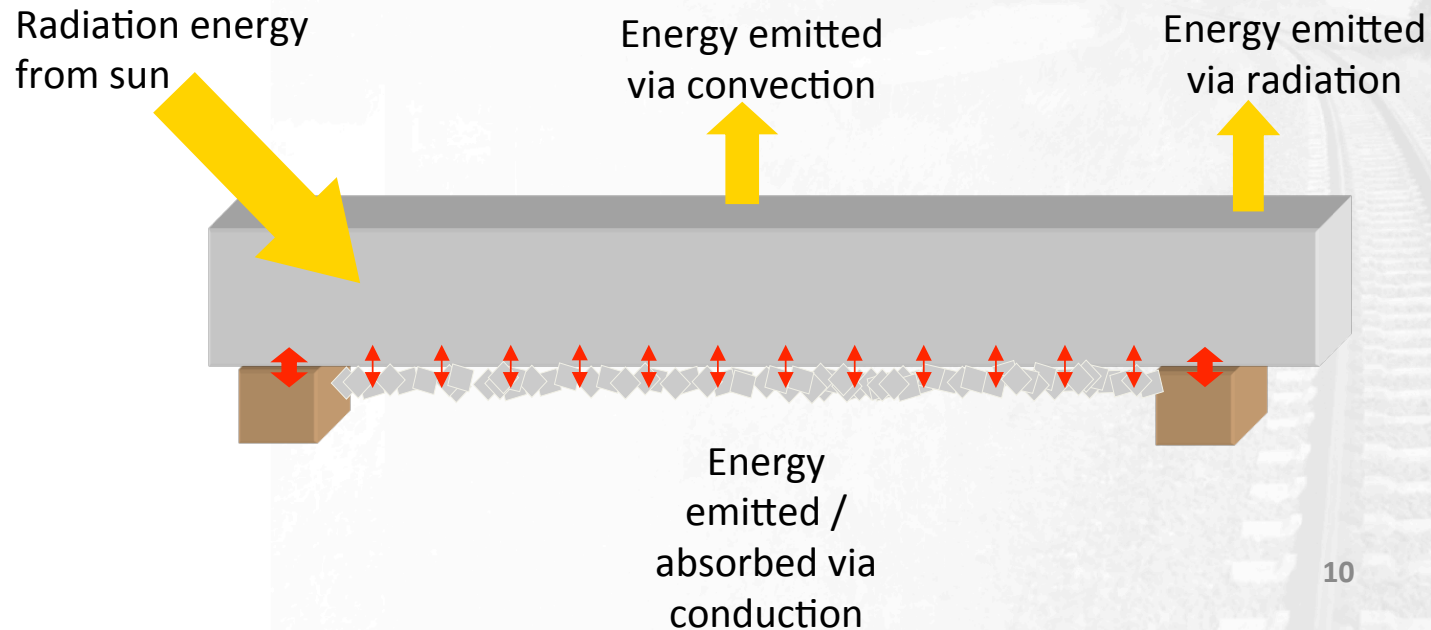


# Research Objectives

- Assist in more objective and reliable issuance of slow orders
- Develop a model that accurately predicts rail temperatures at all times throughout the day
  - Predicts rail temperatures up to nine hours in advance
- Incorporate predictions into a web-based application to be utilized by industry for operations and research purposes

# Research Approach

- Develop physical model based on heat transfer principles
- Diagram of physical heat transfer model parameters:



# Rail Temperature Model

Heat transfer equation

$$k\alpha_s A_s G_s \cos(\theta) - \left[ h_{conv} A_c (T_r - T_\infty) + \epsilon\sigma A_r (T_r^4 - T_{sky}^4) \right] = \rho c V \frac{dT_r}{dt}$$

Solar Radiation

Convection

Sky Radiation

Qualitative explanation of equation

$$\left( \begin{array}{c} \text{Energy Absorbed} \\ \text{Via Solar Radiation} \end{array} \right) - \left[ \left( \begin{array}{c} \text{Energy Emitted} \\ \text{Via Convection} \end{array} \right) + \left( \begin{array}{c} \text{Energy Emitted} \\ \text{Via Sky Radiation} \end{array} \right) \right] = \left( \begin{array}{c} \text{Rate of Change} \\ \text{of Rail Temperature} \end{array} \right)$$

# Data Collection

- Set up weather station with short section of rail (including ties and ballast)
  - Verify accuracy of predicted weather data (solar radiation, ambient temperature, and wind speed)
  - Verify accuracy of rail temperature model predictions



# Data Collection

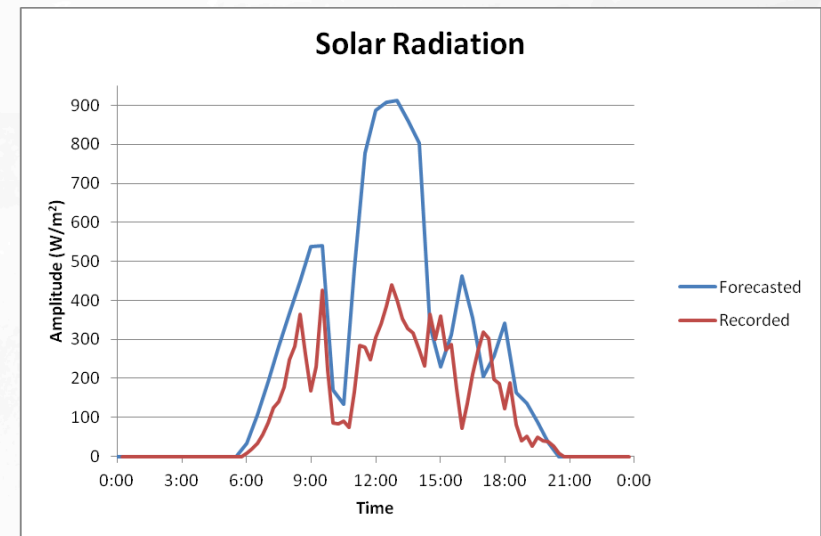
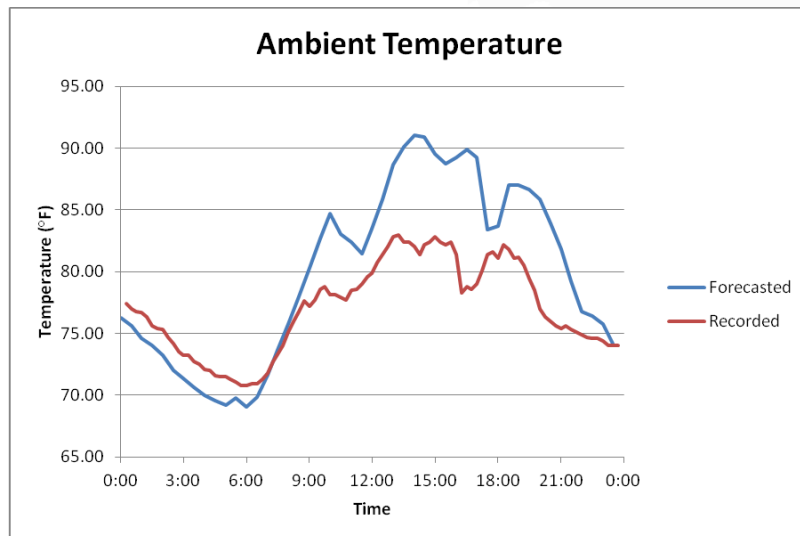
Received recorded weather data and rail temperatures from several wayside weather stations on Amtrak's Northeast Corridor

- Data collected at 15 minute intervals
- Stations installed approximately 30 feet above ground



# Initial Lessons Learned

- Forecast weather data sometimes is not accurate
- Example of inaccurate ambient temperature and solar radiation forecast



# Model Modifications and Improvements

- Utilized measured weather data as inputs into model
  - Allows for isolation of model errors by eliminating input uncertainties due to weather forecast errors
- Based on this approach, several improvements were made to the model
  - Rail orientation / Solar angle

# Data Collection



Installed two ground-level weather stations at a couple locations on Amtrak's Northeast Corridor

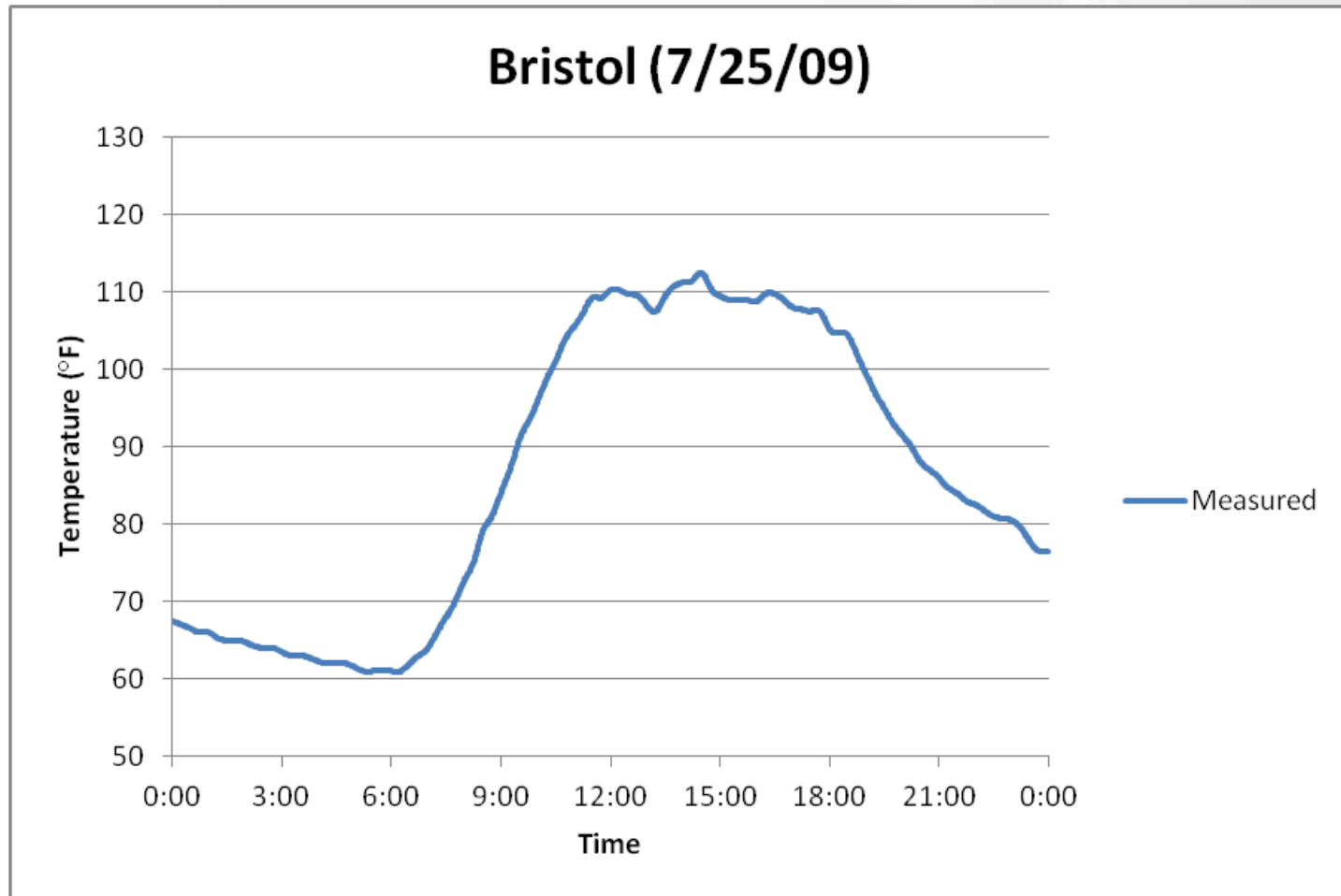


# Data Collection

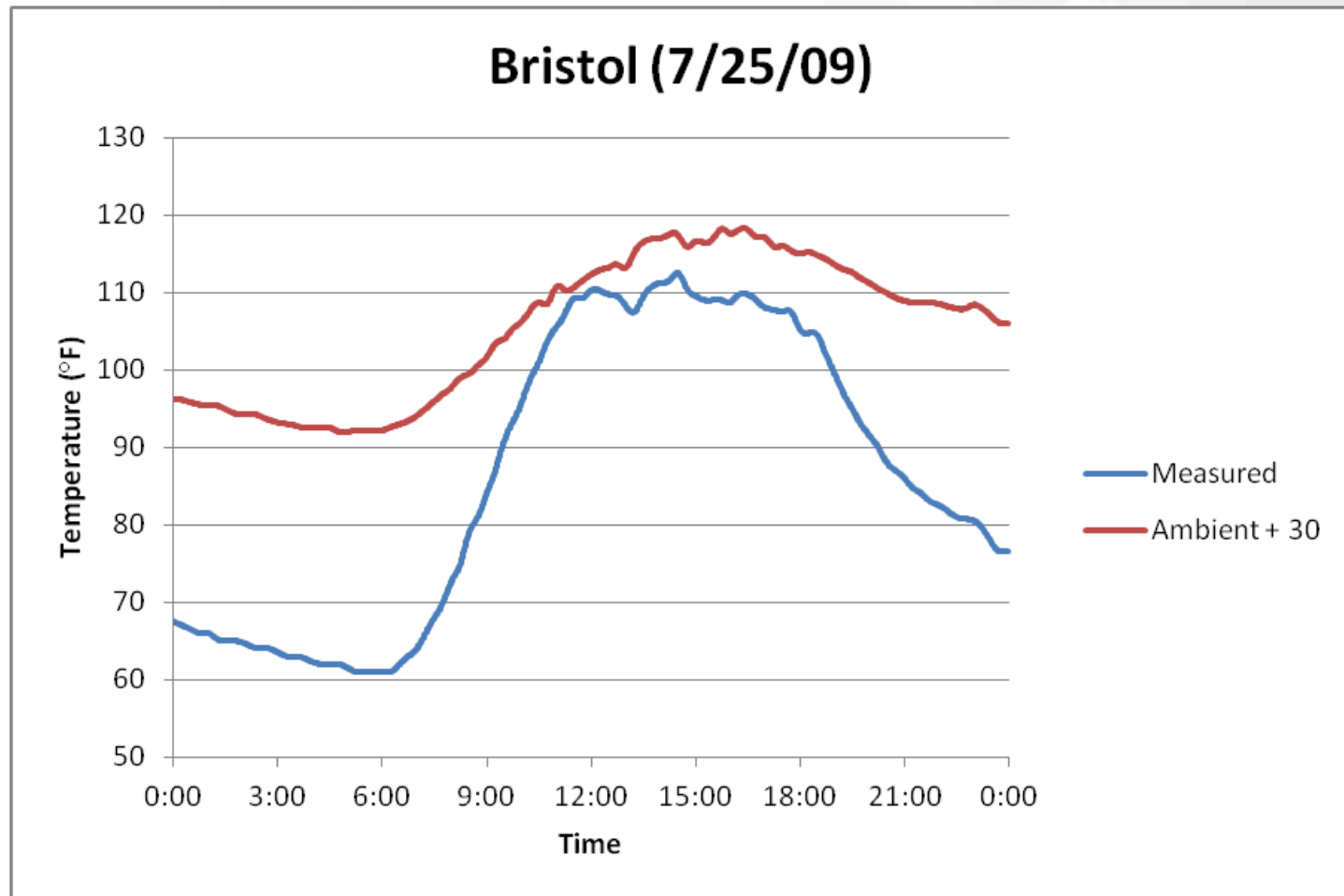
- Ambient temperature at ground level typically 1 – 2 °F higher than ambient temperature measured 30 feet above ground level
- Wind speed at ground level typically 25% - 50% less than wind speed measured 30 feet above ground level



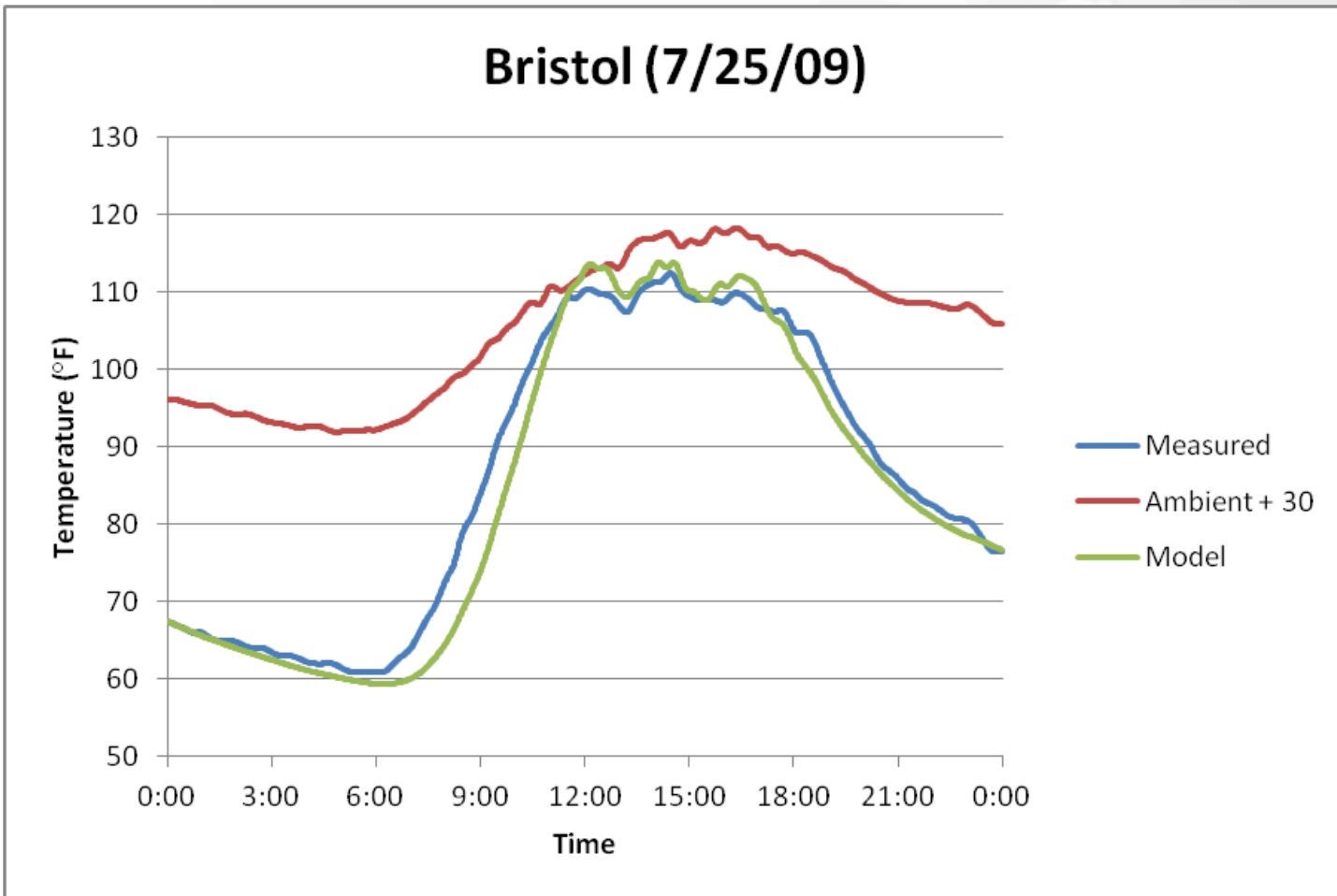
# Example of Model Output



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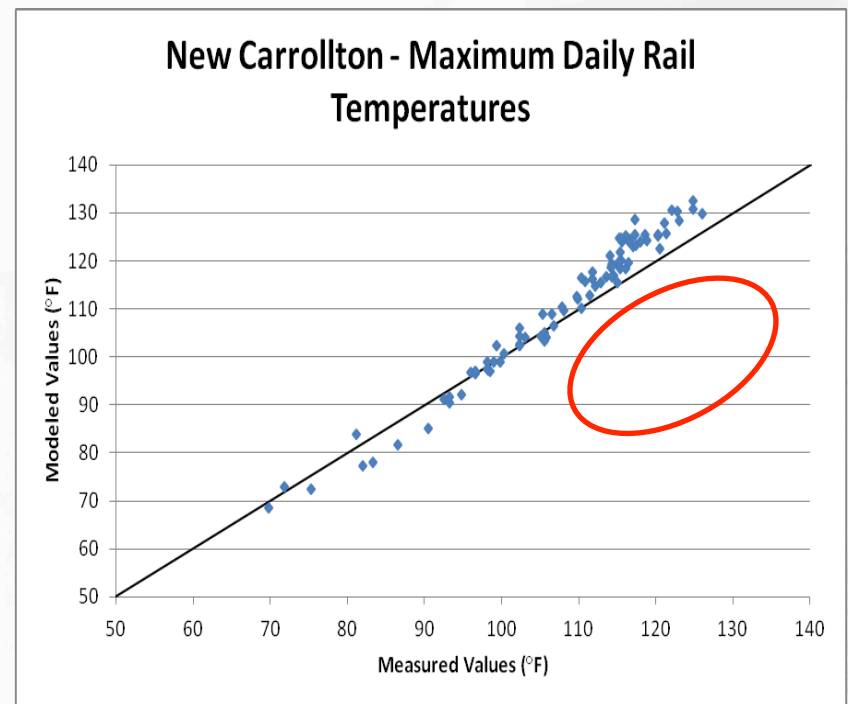
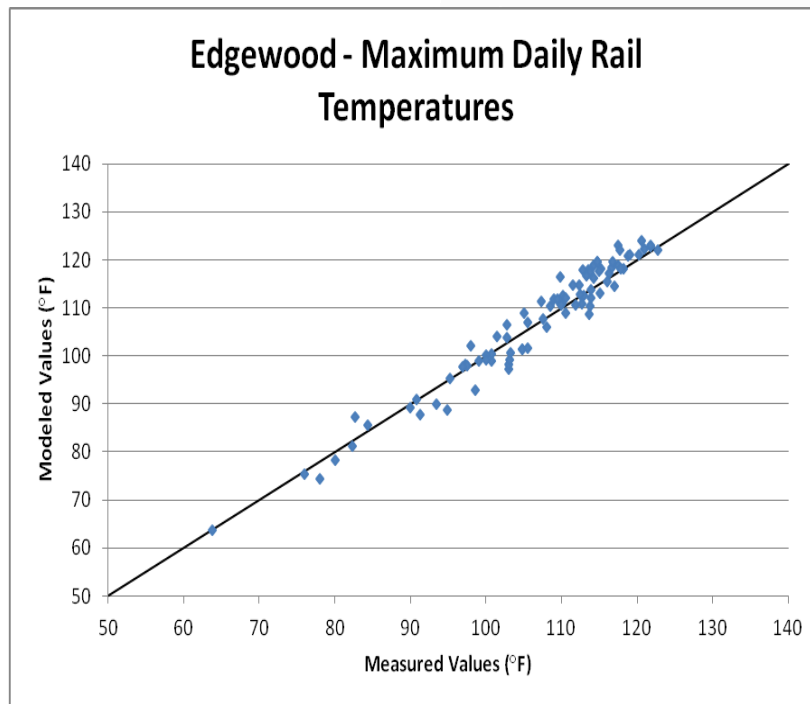


# Example of Model Output



# Lessons Learned – Role of Local Conditions

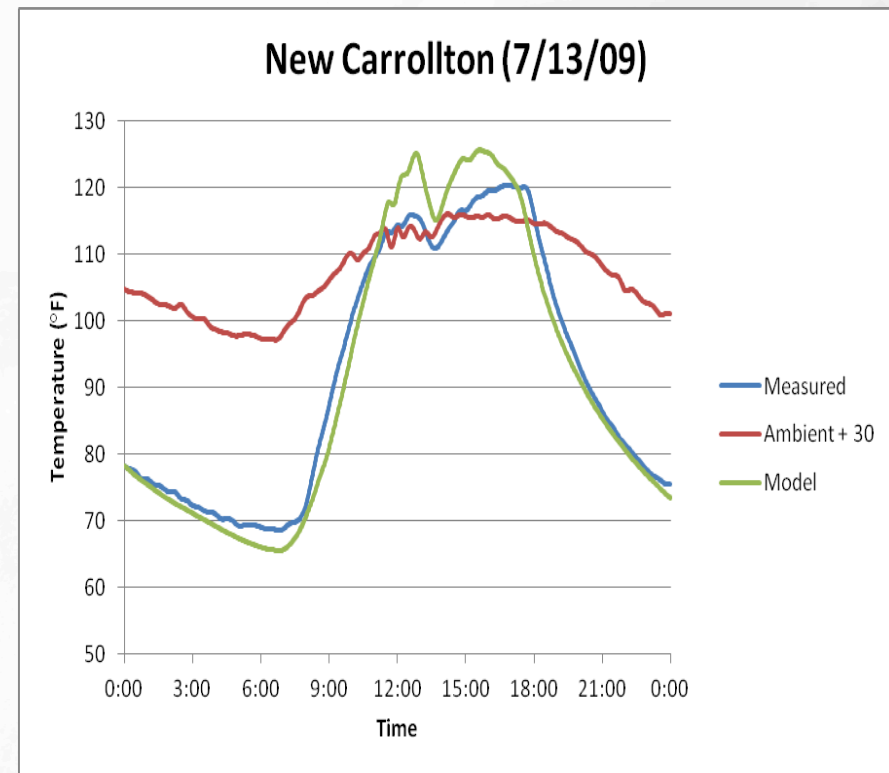
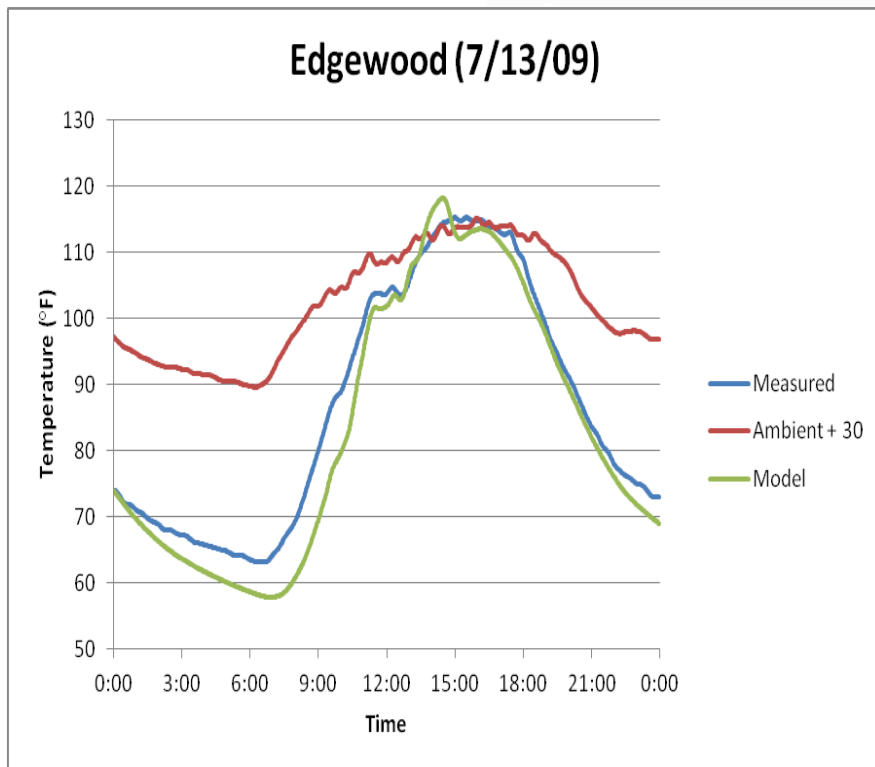
## Local conditions affect rail temperature



New Carrollton data shows offset on hot days

# Lessons Learned – Role of Local Conditions

## Local conditions affect rail temperature



# Lessons Learned – Role of Local Conditions

Solar absorptivity coefficient

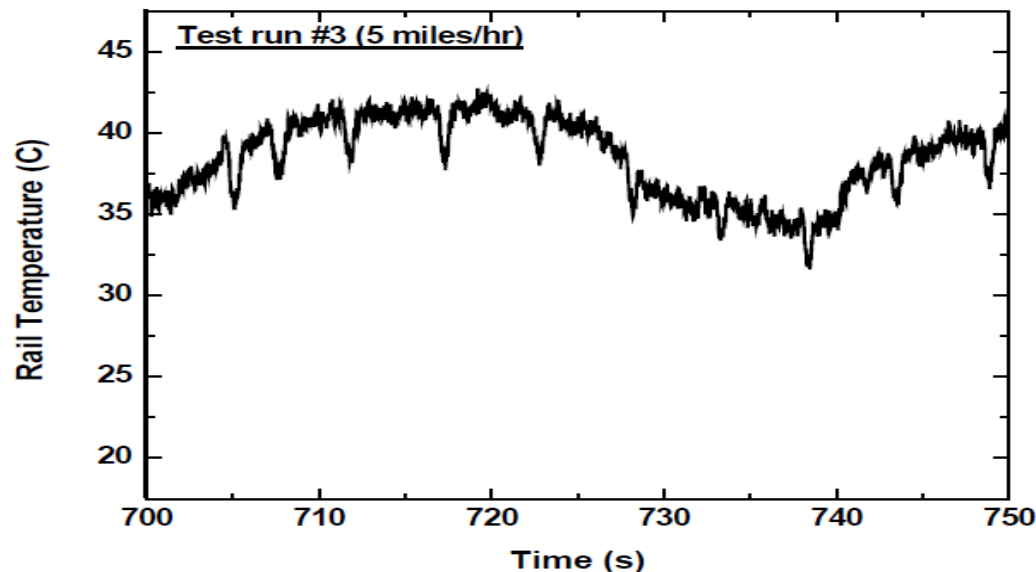
$$k \boxed{\alpha_s} A_s G_s \cos(\theta) - \left[ h_{conv} A_c (T_r - T_\infty) + \varepsilon \sigma A_r (T_r^4 - T_{sky}^4) \right] = \rho c V \frac{dT_r}{dt}$$

Convection coefficient

$$k \alpha_s A_s G_s \cos(\theta) - \left[ \boxed{h_{conv}} A_c (T_r - T_\infty) + \varepsilon \sigma A_r (T_r^4 - T_{sky}^4) \right] = \rho c V \frac{dT_r}{dt}$$

# Lessons Learned – Role of Local Conditions

- More mass can affect rate of increase / decrease of rail temperature
- Data from rail temperature measurement sensor is currently being developed





# Lessons Learned – Role of Local Conditions

More mass can affect rate of increase / decrease of rail



# Lessons Learned – Role of Local Conditions

Area to volume ratio can affect rate of increase of rail temperature

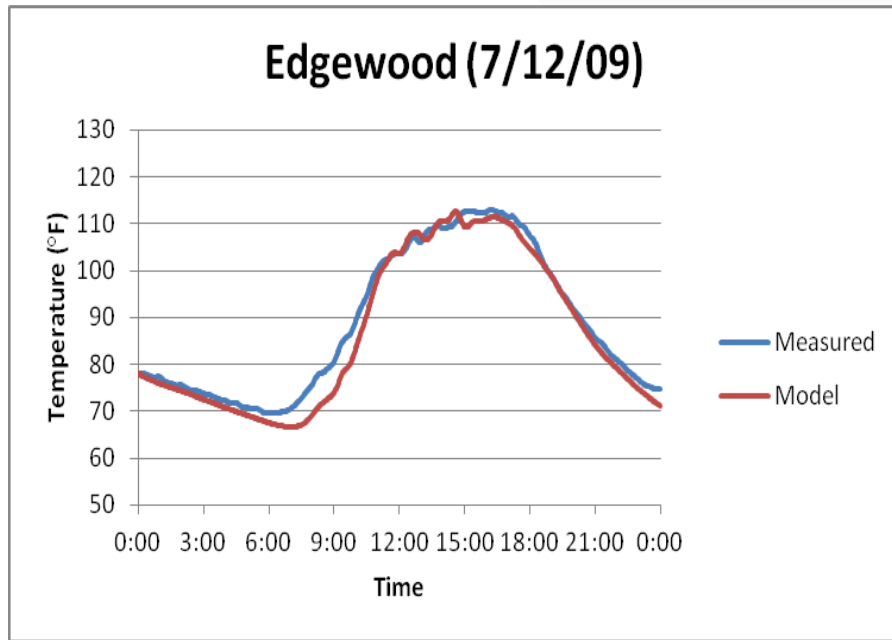
$$k\alpha_s A_s G_s \cos(\theta) - \left[ h_{conv} A_c (T_r - T_\infty) + \epsilon\sigma A_r (T_r^4 - T_{sky}^4) \right] = \rho c V \frac{dT_r}{dt}$$

As a result, rail sections with different area to volume ratios may heat up at different rates

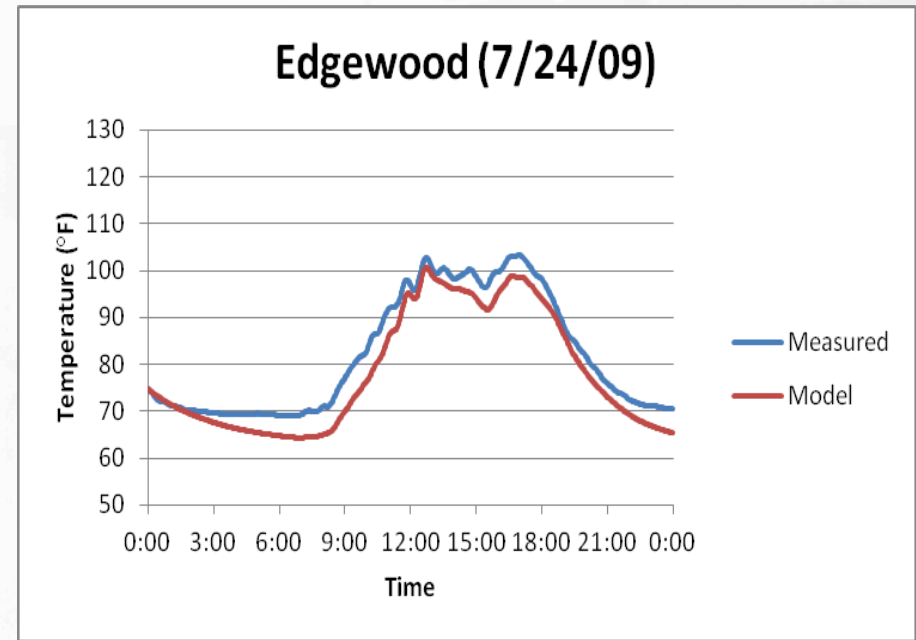
# Lessons Learned – Intermittent Offset Issue

Continuous offset is sometimes observed

No offset



Continuous offset



# Lessons Learned – Potential Role of Sky Temperature

- Offset most likely not due to local conditions
  - Only observed periodically on certain days
- May be result of the sky temperature

$$k\alpha_s A_s G_s \cos(\theta) - \left[ h_{conv} A_c (T_r - T_\infty) + \varepsilon \sigma A_r (T_r^4 - T_{sky}^4) \right] = \rho c V \frac{dT_r}{dt}$$

# Lessons Learned – Potential Role of Sky Temperature

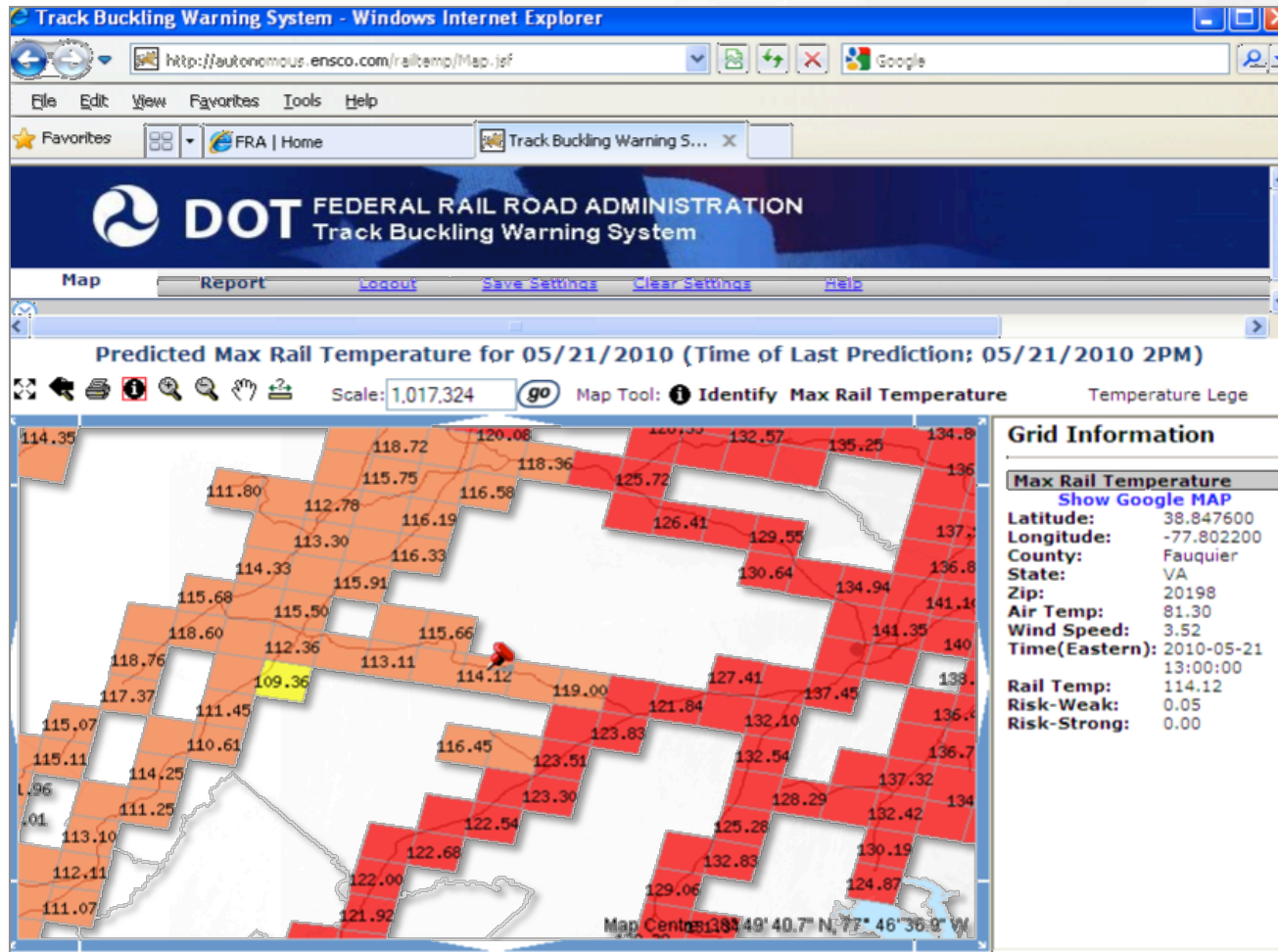
Measured and predicted sky temperatures are not readily available

- Sky temperature can vary from 0 °F to 140 °F below ambient air temperature
- Some empirical models based on dew point temperature are available but their accuracy is not known
- Currently assume a sky temperature at 68 °F (20 °C) below ambient temperature

# Web-Based Application

- Receiving real-time weather forecast data in 9x9 km grids
- Predicting rail temperature 12 hours in advance; updated at 3 hour intervals
- Only generating predictions in grids that have railroad tracks
  - Using National Transportation Atlas Data (NTAD) to determine rail line locations

# Web-Based Application



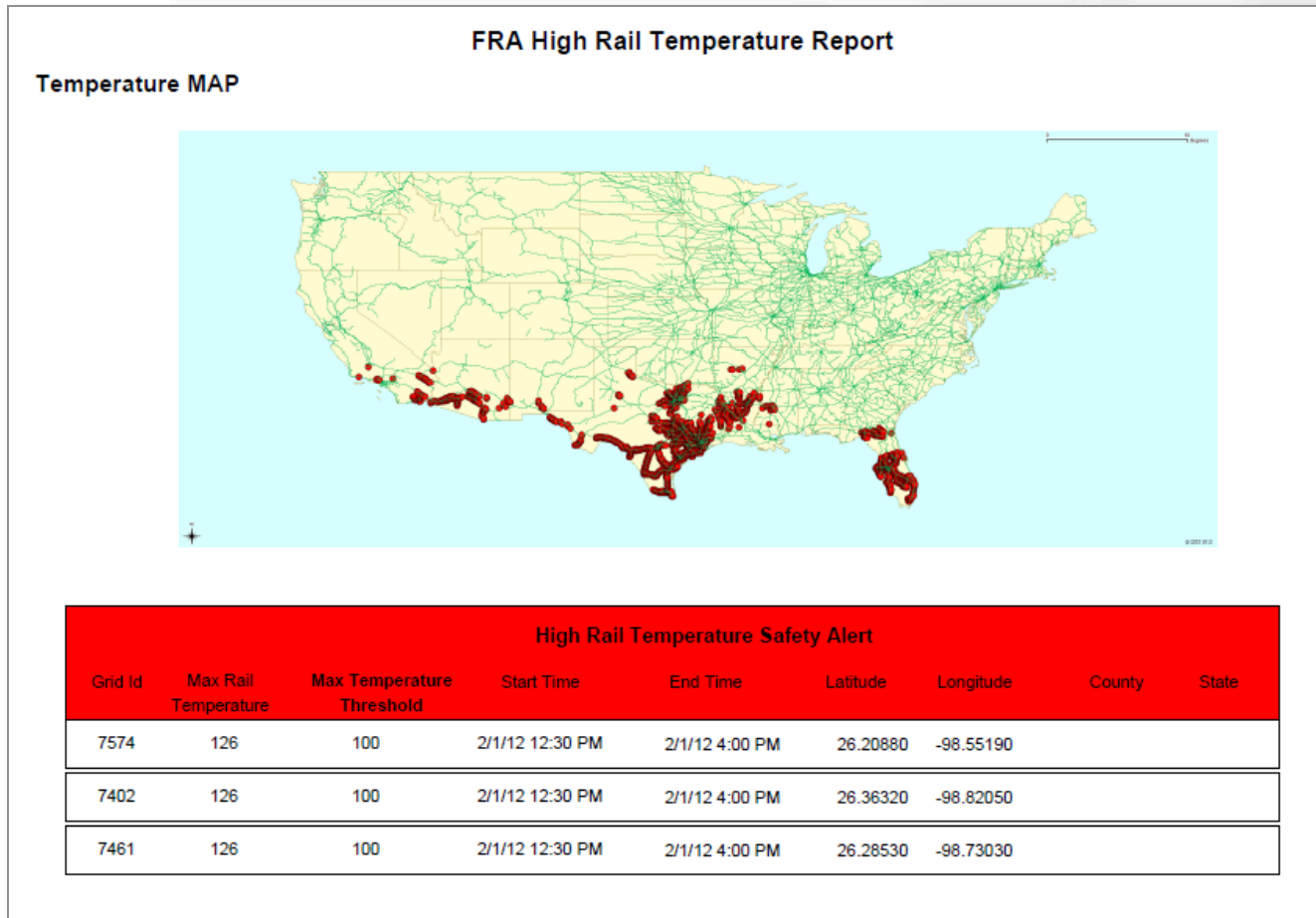
\*Example screenshot

# Improvements to Web-Based Application

- Customizing user logons
  - Currently Amtrak has the only user logon, and only the Northeast Corridor is shown
- Developing automated customizable reports
  - Report can be sent directly to user's email
  - User will not have to logon to website



# Web-Based Application



*\*Example Report*

# Acknowledgements

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