



UNIVERSITY of  
**LOUISIANA**  
L A F A Y E T T E \*

# Lecture 5

## Introduction of WRF-HRRR and Mesonet Data

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# **WRF-The Weather Research and Forecasting**

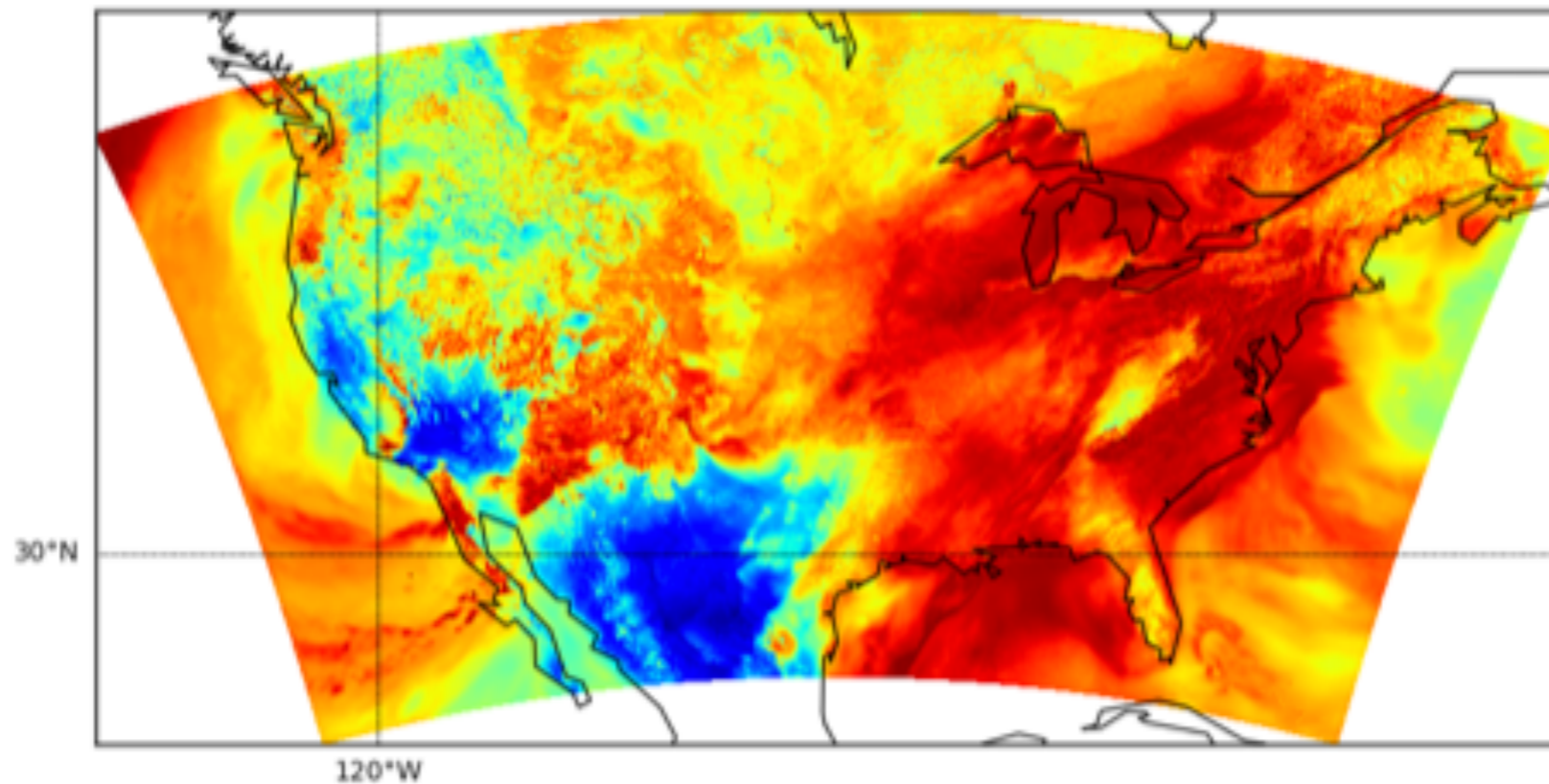
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- **Next generation mesoscale numerical weather prediction system**
- **It can produce simulations based on actual atmospheric conditions (i.e., from observations and analysis)**
- **It has been developed since the later of 1990's**

# WRF-HRRR

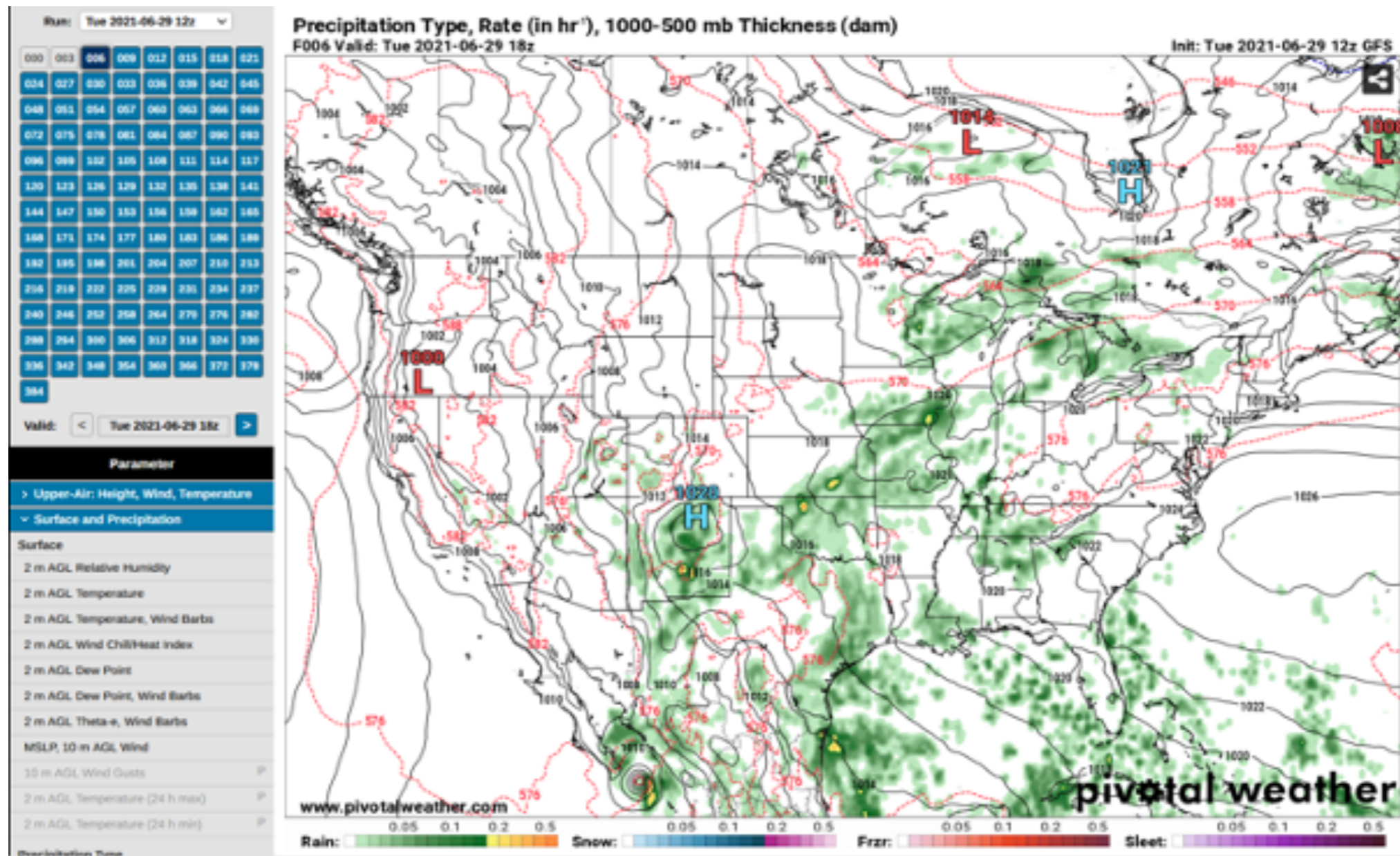
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The Weather Research and Forecasting Model with High-resolution Rapid Refresh



Predict hourly weather parameters covering US continent

# WRF-HRRR: High-resolution Rapid Refresh



Source: <https://www.pivotalweather.com/model.php>

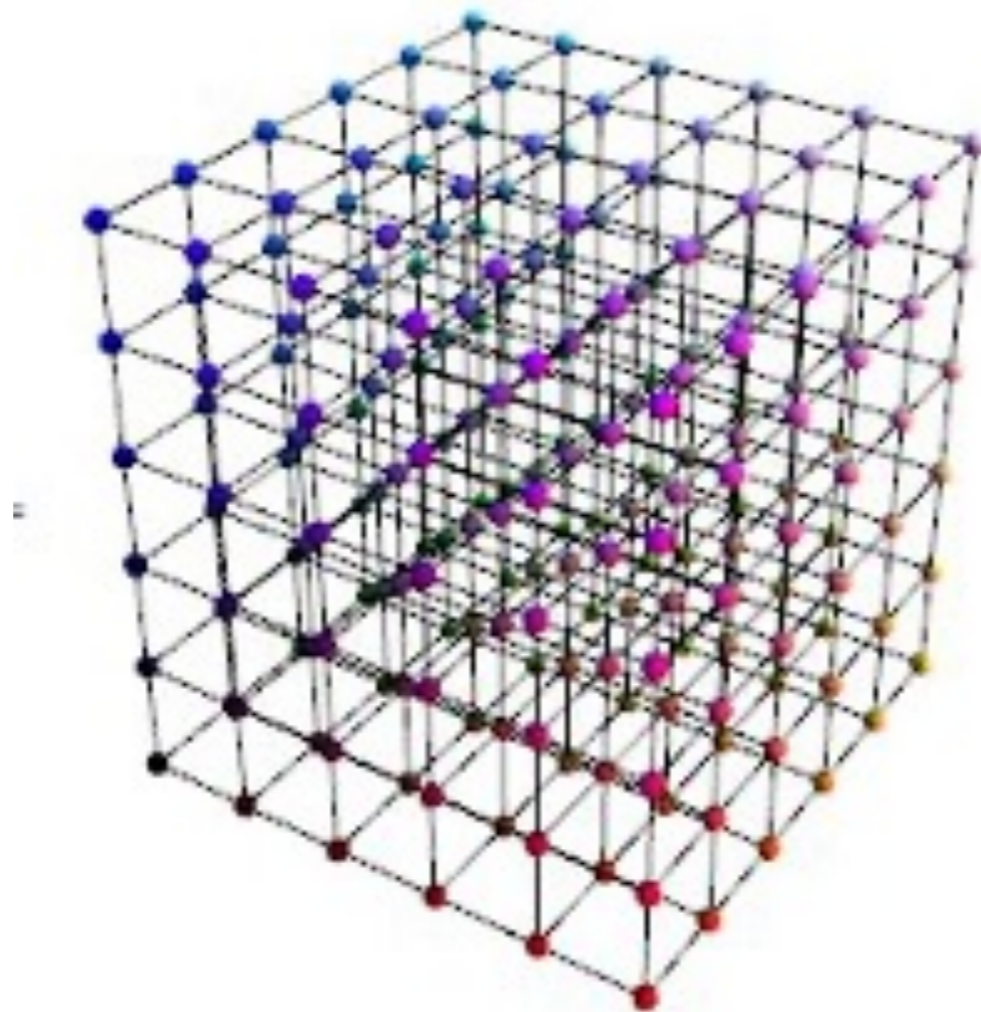




# WRF-HRRR Data Format

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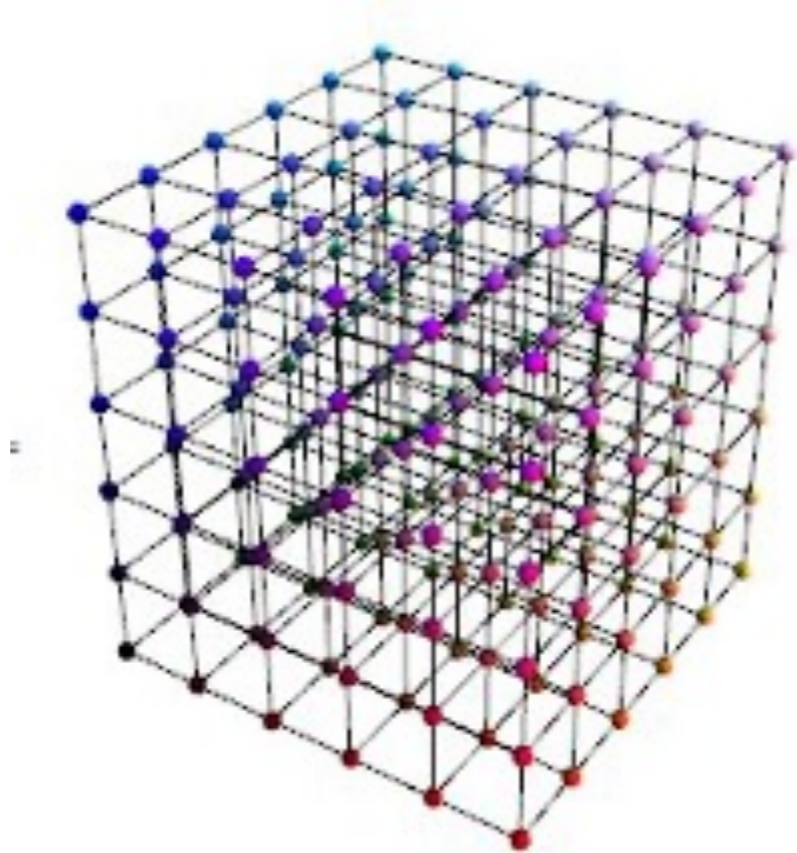
- **HRRR models store data in GRIB format (i.e., 3-D grid), which is a compressed format**
- **Each Grid is of fixed size, 3km x 3km**
  - ▶ Covering the United States continent: 1059 x1799 geo-grids



# WRF-HRRR Data Format

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- **Each layer in a GRIB file represents one feature (e.g., temperature), spanning throughout United States**
  - ▶ Horizon represents locations and vertical represents features
- **So all vertically aligned grid points represents the set of features for a particular location**
  - ▶ The latitude and longitude information are encoded in the GRIB file



148 Parameters = 148 Layers

## Examples for Features at Some Layers

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Layer	Feature
1	Maximum/Composite radar reflectivity
8	Wind speed (gust)
11	U component of wind
12	V component of wind
13	Geopotential Height
14	Temperature
15	Dew point temperature
57	Surface pressure
66	2 meter temperature
71	10 meter U wind component
72	10 meter V wind component
73	10 meter wind speed



# Extracting Weather Conditions at A Location

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- **The latitude and longitude of the UL Lafayette (ULL) is 30.2126 and -92.0193, respectively**
  - ▶ How to get the weather conditions at ULL?
- **We can fetch the latitude and longitude matrix from GRIB file**
  - ▶ Find the grid point that has the closest distance to ULL

# Extracting Weather Conditions at ULL

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```
lt_ULL = 30.2126
ln_ULL = -92.0193
gr = pygrib.open('path/to/grib/file')           # open file
msg = gr [1]                                    # get layer-1 message (any layer no. works here).
lt, ln = msg.latlons()                          # extract GPS coordinate
dis_mat = (lt-lt_ULL)**2+(ln-ln_ULL)**2         # compute distance between each grid point and ULL
p_lt, p_ln = numpy.unravel_index(dis_mat.argmin(), dis_mat.shape) # pick smallest distance index
data = msg.values
feature_ULL = data[p_lt,p_ln]
```

# Mesonet

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- **Comprising a set of automated weather stations located at some specific area in the USA**
  - ▶ Each station monitors tens of of atmospheric measurements, like temperature, rainfall, wind speed, etc., once per minute
  - ▶ South Alabama Mesonet includes a network of 26 weather stations, maintained by Dr. Sytske Kimball, Co-PI of our project
  - ▶ Kentucky Mesonet is led by Dr. Eric Rappin, Co-PIs of our project

# South Alabama Mesonet

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- **Data is publicly available at: [http://chiliweb.southalabama.edu/archived\\_data.php](http://chiliweb.southalabama.edu/archived_data.php)**
  - ▶ A combination of selectable features for a given range of date is available for downloading
  - ▶ Dataset includes 60 features, excluding time, date, and location
  - ▶ Data are in CSV format



# South Alabama Mesonet



# South Alabama Mesonet

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# South Alabama Mesonet

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**Temperature** (soil - 5 depths and above the surface at 1.5, 2, 9.5, and 10 m).

**Relative Humidity** (above the surface at 2 and 10 m).

**Horizontal Wind Speed and Direction** (2 and 10 m).

**Vertical Wind Speed** (10 m).

**Atmospheric Pressure.**

**Rainfall.**

**Solar Radiation** (Total Radiation and PAR).

# Example

Select Meteorological Data to Download

Begin Date:  End Date:  Station:  Format:  CSV  Fixed

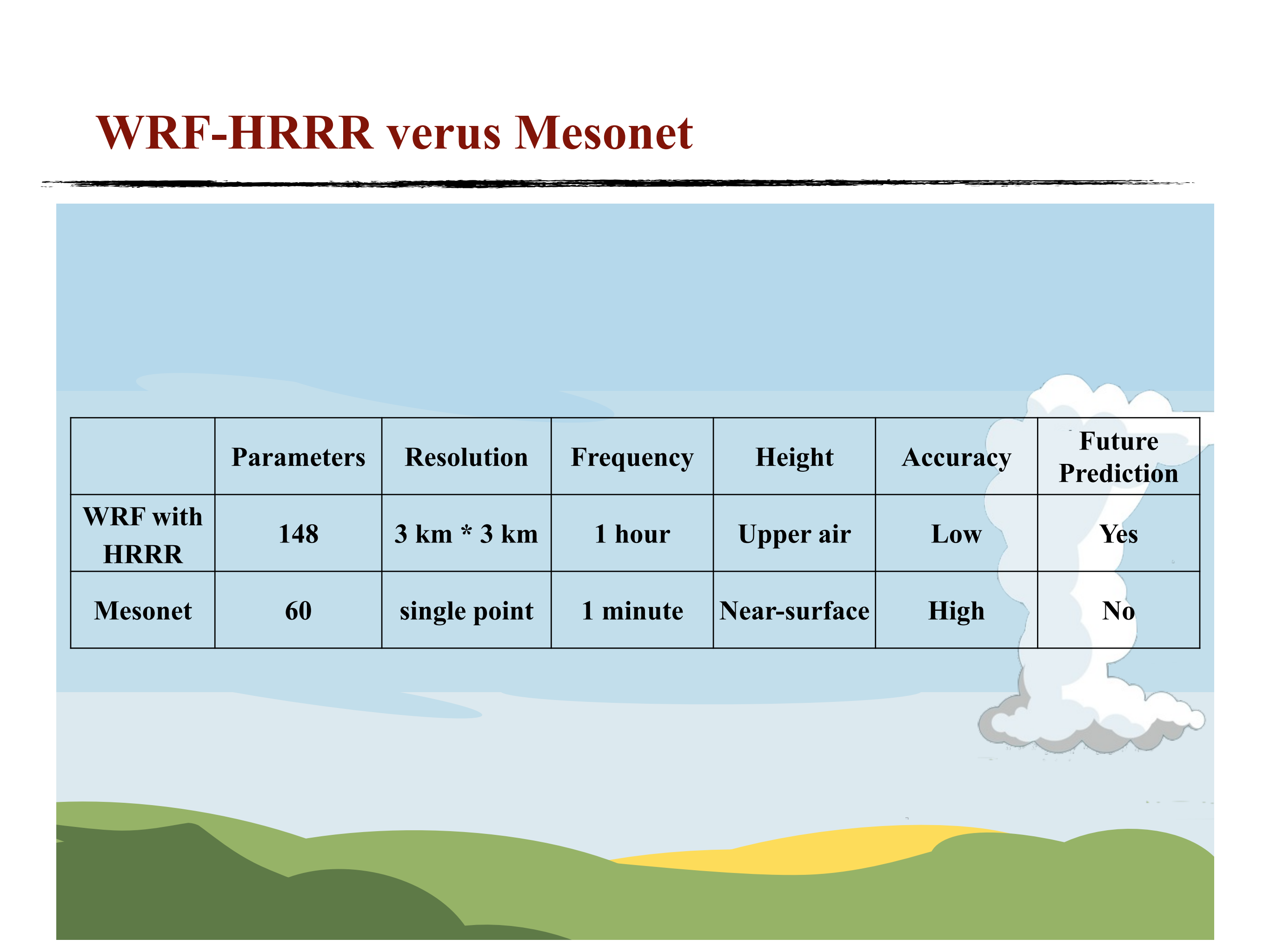
<input checked="" type="checkbox"/>	Select/Deselect All
<input checked="" type="checkbox"/>	Record Id
<input checked="" type="checkbox"/>	Table Code
<input checked="" type="checkbox"/>	Year
<input checked="" type="checkbox"/>	Month
<input checked="" type="checkbox"/>	Day of Month
<input checked="" type="checkbox"/>	Day of Year
<input checked="" type="checkbox"/>	Hour
<input checked="" type="checkbox"/>	Minute
<input checked="" type="checkbox"/>	Station Id
<input checked="" type="checkbox"/>	Latitude
<input checked="" type="checkbox"/>	Longitude
<input checked="" type="checkbox"/>	Elevation
<input checked="" type="checkbox"/>	Sign
<input checked="" type="checkbox"/>	Door open indicator
<input checked="" type="checkbox"/>	Battery Voltage
<input checked="" type="checkbox"/>	Observations in the last minute
<input checked="" type="checkbox"/>	Precipitation over the last minute (TB3)
<input checked="" type="checkbox"/>	Precipitation over the last minute (TX)
<input checked="" type="checkbox"/>	Precipitation since midnight (TB3)

- ✓ Agricola
- Andalusia
- Ashford
- Ashford North
- Atmore
- Bay Minette
- Bay Minette FS
- Bayou La Batre
- Castleberry
- Dauphin Island
- Dixie
- Elberta
- Fairhope
- Floral
- Foley
- Gasque
- Geneva
- Grand Bay
- Jay
- Kinston
- Leakesville
- Loxley
- Mobile (Dog River)
- Mobile (USA Campus)
- Mobile (USA Campus West)
- Mount Vernon
- Pascagoula
- Poarch Creek
- Robertsdale
- Saraland
- Walnut Hill

Source: [http://chiliweb.southalabama.edu/archived\\_data.php](http://chiliweb.southalabama.edu/archived_data.php)



# WRF-HRRR versus Mesonet



	<b>Parameters</b>	<b>Resolution</b>	<b>Frequency</b>	<b>Height</b>	<b>Accuracy</b>	<b>Future Prediction</b>
<b>WRF with HRRR</b>	<b>148</b>	<b>3 km * 3 km</b>	<b>1 hour</b>	<b>Upper air</b>	<b>Low</b>	<b>Yes</b>
<b>Mesonet</b>	<b>60</b>	<b>single point</b>	<b>1 minute</b>	<b>Near-surface</b>	<b>High</b>	<b>No</b>

## We would like to ...

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By incorporating the two datasets, we develop Deep Learning approach to predict the future weather conditions.

The good thing here is that you don't need to label the data.

# Comparing to Twitter Data

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- **Twitter Data**

- ▶ Unstructured
- ▶ Classification on purpose
- ▶ Classification based on spam patterns: feature extraction
- ▶ No ground truth
- ▶ Binary classification

- **Weather Data**

- ▶ Structured
- ▶ Prediction on purpose
- ▶ All features (weather parameters) have been provided
- ▶ No ground truth
- ▶ Time-series Prediction