## GDD\_Calculation\_and\_Drying\_Algorithm

May 6, 2025

```
[1]: from google.colab import output
    output.enable_custom_widget_manager()
[2]: !pip install ipywidgets plotly --quiet
                          1.6/1.6 MB
   8.6 MB/s eta 0:00:00
   Model 1: GDD Calculation
[4]: import urllib.parse
    import urllib.request
    from io import BytesIO
    import pandas as pd
    import plotly.express as px
    import ipywidgets as widgets
    from ipywidgets import VBox, HBox, Layout, Button, Output
    from IPython.display import display
    import math
    # 1) Function to Fetch Weather Data from Visual Crossing
    def fetch_weather_data_from_api(api_key, city, state, start_date, end_date,
                                unit_group="us",_
     ⇔elements="datetime,tempmin,tempmax"):
       Fetches weather data from Visual Crossing between start_date and end_date
       for a given city, state (e.g. city='Ithaca', state='NY').
       Returns a DataFrame with columns ['datetime', 'tempmin', 'tempmax',...].
       location_string = f"{city}, {state}"
        encoded_location = urllib.parse.quote(location_string)
```

```
base_url = "https://weather.visualcrossing.com/VisualCrossingWebServices/
 →rest/services/timeline"
   url = (
       f"{base_url}/{encoded_location}/{start_date}/{end_date}"
       f"?unitGroup={unit_group}&include=days&elements={elements}"
       f"&key={api key}&contentType=csv"
   )
   try:
       response = urllib.request.urlopen(url)
       df_api = pd.read_csv(BytesIO(response.read()))
       return df_api
   except urllib.error.HTTPError as e:
       print("HTTP Error:", e.code, e.read().decode())
   except urllib.error.URLError as e:
       print("URL Error:", e.reason)
   return pd.DataFrame() # Return empty if error
# 2) GDD Calculation Functions
def calc_average_gdd(t_min, t_max, T_base):
   return max(((t_max + t_min) / 2) - T_base, 0)
def calc_single_sine_gdd(t_min, t_max, T_base):
   T_{mean} = (t_{max} + t_{min}) / 2
   A = (t_max - t_min) / 2
   if t_max <= T_base:</pre>
       return 0.0
   if t min >= T base:
       return (T_mean - T_base)
   alpha = (T_base - T_mean) / A
   theta = math.acos(alpha)
   dd = ((T_mean - T_base)*(math.pi - 2*theta) + A*math.sin(2*theta)) / math.pi
   return dd
def calc_single_triangle_gdd(t_min, t_max, T_base):
   if t_max <= T_base:</pre>
       return 0.0
   if t_min >= T_base:
       return ((t_max + t_min)/2 - T_base)
   proportion_of_day = (t_max - T_base) / (t_max - t_min)
   avg\_above = ((t\_max + T\_base) / 2) - T\_base
   dd = proportion_of_day * avg_above
   return max(dd, 0)
```

```
def calc double sine gdd(t min_today, t max_today, t min_tomorrow, T_base):
    seg1 = calc_single_sine_gdd(t_min_today, t_max_today, T_base) * 0.5
    seg2 = calc single_sine_gdd(t_min_tomorrow, t_max_today, T_base) * 0.5
    return seg1 + seg2
def calc_double_triangle_gdd(t_min_today, t_max_today, t_min_tomorrow, T_base):
    seg1 = calc_single_triangle_gdd(t_min_today, t_max_today, T_base) * 0.5
    seg2 = calc_single_triangle_gdd(t_min_tomorrow, t_max_today, T_base) * 0.5
    return seg1 + seg2
def calculate_daily_gdd(row, df,
                        method="average",
                        T_base=50.0,
                        use_modified=False,
                        T_lower=50.0,
                        T_upper=86.0):
    idx = row.name
    t max = row['tmax']
    t_min = row['tmin']
    if use_modified:
        t_max = min(t_max, T_upper)
        t_min = max(t_min, T_lower)
    if method == "average":
        return calc_average_gdd(t_min, t_max, T_base)
    elif method == "sine":
        return calc_single_sine_gdd(t_min, t_max, T_base)
    elif method == "triangle":
        return calc_single_triangle_gdd(t_min, t_max, T_base)
    elif method == "double_sine":
        if idx < len(df) - 1:
            t_min_next = df.loc[idx+1, 'tmin']
            if use_modified:
                t_min_next = max(t_min_next, T_lower)
            return calc double sine gdd(t min, t max, t min next, T base)
        else:
            return calc_single_sine_gdd(t_min, t_max, T_base)
    elif method == "double_triangle":
        if idx < len(df) - 1:
            t_min_next = df.loc[idx+1, 'tmin']
            if use modified:
                t_min_next = max(t_min_next, T_lower)
            return calc_double_triangle_gdd(t_min, t_max, t_min_next, T_base)
        else:
            return calc_single_triangle_gdd(t_min, t_max, T_base)
    else:
        raise ValueError(f"Unknown method: {method}")
```

```
def calculate_gdds_for_df(df,
                       start_date,
                       end_date,
                       method="average",
                       T_base=50.0,
                       use modified=False,
                       T_lower=50.0,
                       T_upper=86.0):
   df = df.copy()
   df['date'] = pd.to_datetime(df['date'])
   df.sort_values('date', inplace=True)
   df['daily_gdd'] = df.apply(
       lambda row: calculate_daily_gdd(row, df,
                                    method=method,
                                    T_base=T_base,
                                    use_modified=use_modified,
                                    T_lower=T_lower,
                                    T_upper=T_upper),
       axis=1
   start_date = pd.to_datetime(start_date)
   end_date = pd.to_datetime(end_date)
   df.loc[df['date'] < start date, 'daily gdd'] = 0</pre>
   df.loc[df['date'] > end_date, 'daily_gdd'] = 0
   df['cumulative_gdd'] = df['daily_gdd'].cumsum()
   return df
# 3) Interactive UI
def interactive_gdd_api_viewer(api_key):
   Provides a UI with separate city/state inputs, and improved layout
   to avoid overlapping or scrunching of widgets.
   11 11 11
   city_widget = widgets.Text(
       value="Ithaca",
       description="City:",
       layout=Layout(width="200px")
   state_widget = widgets.Text(
       value="NY",
       description="State:",
       layout=Layout(width="150px")
   )
```

```
start_date_widget = widgets.Text(
      value="2025-01-01",
      description="Start:",
      layout=Layout(width="180px")
  end_date_widget = widgets.Text(
      value="2025-03-31",
      description="End:",
      layout=Layout(width="180px")
  method_widget = widgets.Dropdown(
      options=["average", "sine", "triangle", "double_sine", u

¬"double_triangle"],
      value="average",
      description="Method:",
      layout=Layout(width="200px")
  )
  tbase_widget = widgets.FloatText(
      value=50.0,
      description="T_base:",
      layout=Layout(width="150px")
  )
  modified_widget = widgets.Checkbox(
      value=False,
      description="Use Modified?"
  tlower_widget = widgets.FloatText(
      value=50.0,
      description="T_lower:",
      layout=Layout(width="150px", display="none")
  tupper_widget = widgets.FloatText(
      value=86.0,
      description="T_upper:",
      layout=Layout(width="150px", display="none")
  )
  update_button = widgets.Button(
      description="Get Data & Calculate",
      button_style="primary",
      layout=Layout(width="170px")
  plot_output = Output()
  def on_modified_change(change):
      if modified_widget.value:
          tlower_widget.layout.display = "block"
          tupper_widget.layout.display = "block"
```

```
else:
        tlower_widget.layout.display = "none"
        tupper_widget.layout.display = "none"
modified_widget.observe(on_modified_change, names="value")
def on_update_button_click(b):
    with plot_output:
        plot_output.clear_output(wait=True)
        city_val = city_widget.value
        state_val = state_widget.value
        sdate = start_date_widget.value
        edate = end_date_widget.value
        df_api = fetch_weather_data_from_api(
            api_key=api_key,
            city=city_val,
            state=state_val,
            start_date=sdate,
            end_date=edate,
            unit_group="us",
            elements="datetime,tempmin,tempmax"
        )
        if df_api.empty:
            print("No data returned from API or an error occurred.")
        df_renamed = df_api.copy()
        df_renamed.rename(columns={
            "datetime": "date",
            "tempmax": "tmax",
            "tempmin": "tmin"
        }, inplace=True)
        df_calc = calculate_gdds_for_df(
            df=df_renamed,
            start_date=sdate,
            end date=edate,
            method=method_widget.value,
            T_base=tbase_widget.value,
            use_modified=modified_widget.value,
            T_lower=tlower_widget.value,
            T_upper=tupper_widget.value
        )
        location_str = f"{city_val}, {state_val}"
```

```
fig = px.line(
               df_calc,
               x="date",
               y="cumulative_gdd",
               title=f"Cumulative GDD ({method_widget.value}) for_u
 →{location_str}"
            fig.update_layout(xaxis_title="Date", yaxis_title="Cumulative GDD")
            fig.show()
   update_button.on_click(on_update_button_click)
    # -----
    # Lay out the widgets
    # -----
    # row1: city, state
   row1 = HBox([city_widget, state_widget], layout=Layout(padding="5px"))
    # row2: start_date, end_date
   row2 = HBox([start_date_widget, end_date_widget],__
 →layout=Layout(padding="5px"))
    # row3: method, T_base, Use Modified?
   row3 = HBox([method_widget, tbase_widget, modified_widget],__
 ⇔layout=Layout(padding="5px"))
    # row4: T_lower, T_upper, button
   row4 = HBox([tlower_widget, tupper_widget, update_button],__
 ⇔layout=Layout(padding="5px"))
   ui = VBox([row1, row2, row3, row4, plot_output],_
 ⇔layout=Layout(padding="10px"))
    # Initialize the check logic
   on_modified_change(None)
   display(ui)
# API Input
if __name__ == "__main__":
   API_KEY = "BPFV62G5X9SUQB8HLPPFHNYFV"
    interactive_gdd_api_viewer(API_KEY)
```

```
VBox(children=(HBox(children=(Text(value='Ithaca', description='City:',⊔ ⇔layout=Layout(width='200px')), Text(va...
```

Model 2: Drying calculation Algorithm

```
[5]: import urllib.parse import urllib.request
```

```
from io import BytesIO
import numpy as np
import pandas as pd
import plotly.express as px
import ipywidgets as widgets
from ipywidgets import VBox, HBox, Layout, Button, Output
from IPython.display import display
from datetime import datetime
from dateutil.relativedelta import relativedelta
# 1) Function to Fetch Weather Data from Visual Crossing
# -----
def fetch_drying_weather_data_from_api(api_key, city, state, start_date,_
 ⇔end_date,
                               unit_group="us"):
   location_string = f"{city}, {state}"
   encoded_location = urllib.parse.quote(location_string)
   base url = "https://weather.visualcrossing.com/VisualCrossingWebServices/
 ⇔rest/services/timeline"
   daily_url = (
       f"{base_url}/{encoded_location}/{start_date}/{end_date}"
 -unitGroup={unit group}&include=days&elements=datetime,temp,dew,soilmoisturevol01"
       f"&key={api_key}&contentType=csv"
   )
   try:
       response = urllib.request.urlopen(daily_url)
       df_days = pd.read_csv(BytesIO(response.read()))
   except urllib.error.HTTPError as e:
       print("HTTP Error:", e.code, e.read().decode())
   except urllib.error.URLError as e:
       print("URL Error:", e.reason)
   hourly_url = (
       f"{base_url}/{encoded_location}/{start_date}/{end_date}"
 -unitGroup={unit_group}&include=hours&elements=datetime,solarradiation"
       f"&key={api_key}&contentType=csv"
   )
   try:
       response = urllib.request.urlopen(hourly_url)
```

```
df_hours = pd.read_csv(BytesIO(response.read()))
   except urllib.error.HTTPError as e:
       print("HTTP Error:", e.code, e.read().decode())
   except urllib.error.URLError as e:
       print("URL Error:", e.reason)
   try:
       return df_days, df_hours
   except:
       return pd.DataFrame() # Return empty if error
# -----
# 2) Interactive UI to Get Weather Data
def interactive_weather_data_viewer(api_key):
   city_widget = widgets.Text(value="San Joaquin Valley", description="City:", u
 ⇔layout=Layout(width="200px"))
   state_widget = widgets.Text(value="CA", description="State:",_
 ⇔layout=Layout(width="150px"))
   start_date_widget = widgets.Text(value="2024-10-25", description="Harvest_
 →Date:", layout=Layout(width="180px"))
   #moisture_widget = widgets.FloatText(value=.80, description="Starting_
 →Moisture:", layout=Layout(width="180px"))
   #end_date_widget = widgets.Text(value="2025-06-31", description="End:",__
 → layout=Layout (width="180px"))
   update_button = widgets.Button(description="Estimate Drying",__
 ⇔button_style="primary", layout=Layout(width="170px"))
   plot_output = Output()
   def on_update_button_click(b):
       with plot_output:
           plot_output.clear_output(wait=True)
           city = city_widget.value
           state = state_widget.value
           sdate = start_date_widget.value
           sdate dt = datetime.strptime(sdate, "%Y-%m-%d")
           edate_dt = sdate_dt + relativedelta(months=2)
           edate = edate_dt.strftime("%Y-%m-%d")
           d_df, h_df = fetch_drying_weather_data_from_api(
               api_key=api_key,
               city=city,
               state=state,
               start date=sdate,
               end_date=edate
```

```
if d_df.empty or h_df.empty:
                                      print("No data returned from API or an error occurred.")
                                      return
                            df = merge_dfs(daily_df=d_df, hourly_df=h_df)
                            df = predict_moisture_content(df, sdate)
                            df.rename(columns={"datetime": "date"}, inplace=True)
                            df["date"] = pd.to datetime(df["date"])
                            location_str = f"{city}, {state}"
                             # Quick plot of max temp as an example
                            fig = px.line(df, x="date", y="predicted_moisture", __
   بtitle=f"Expected Alfalfa Moisture Content for {location_str} - Harvest Date: المارة 

√{sdate}")
                            fig.update_layout(xaxis_title="Date", yaxis_title="Moisture Contentu
  (%) ")
                            fig.show()
         update_button.on_click(on_update_button_click)
         # Layout
         row1 = HBox([start_date_widget, city_widget, state_widget],__
   →layout=Layout(padding="5px"))
         #row2 = HBox([start_date_widget, end_date_widget],__
  → layout=Layout (padding="5px"))
         row3 = HBox([update_button], layout=Layout(padding="5px"))
         ui = VBox([row1, row3, plot_output], layout=Layout(padding="10px"))
         display(ui)
# 3) Drying calculation Algorithm
# -----
def merge_dfs(daily_df, hourly_df):
         Extracts the peak solar radiation for each day from the
         hourly df and adds it to the daily df.
         111
         d_df = daily_df
         h_df = hourly_df
         # Step 1: Convert datetime columns to datetime objects
         h_df['datetime'] = pd.to_datetime(h_df['datetime'])
```

```
d_df['datetime'] = pd.to_datetime(d_df['datetime'])
    # Step 2: Extract date from hourly timestamps
   h_df['date'] = h_df['datetime'].dt.date
    # Step 3: Group by date and get max solar radiation
   daily_peaks = h_df.groupby('date')['solarradiation'].max().reset_index()
   daily_peaks.rename(columns={'solarradiation': 'peak_solarradiation'},__
 →inplace=True)
   # Step 4: Merge with daily dataframe
   # Convert daily_df datetime to just date for merging
   d_df['date'] = d_df['datetime'].dt.date
   merged_df = pd.merge(d_df, daily_peaks, on='date', how='left')
   # (Optional) Drop the 'date' column if not needed
   merged df.drop(columns='date', inplace=True)
   return merged_df
def calculate vapor pressure deficit(df):
    '''Takes weather df (with temp and dew in {}^{\circ}F) and adds vapor pressure_{\sqcup}
 ⇔deficit column in kPa'''
    # Convert temperature and dew point from Fahrenheit to Celsius
   df['temp_C'] = (df['temp'] - 32) / 1.8
   df['dew C'] = (df['dew'] - 32) / 1.8
   # Calculate saturation and actual vapor pressure
   df['saturation_vapor_pressure'] = 0.6108 * np.exp((17.27 * df['temp_C']) /
 df['actual_vapor_pressure'] = 0.6108 * np.exp((17.27 * df['dew_C']) / _ _
 \hookrightarrow (df['dew_C'] + 237.3))
    # Calculate VPD
   df['vapor_pressure_deficit'] = df['saturation_vapor_pressure'] -__

→df['actual_vapor_pressure']
   return df
def swath_density_conversion(plants_per_sqft, g_per_plant=25):
    '''convert swath density from plants/ft^2 to g/m^2'''
   plants_per_sqm = plants_per_sqft * 10.764
   return plants_per_sqm * g_per_plant # returns g/m2
```

```
def predict_moisture_content(df, startdate, swath_density=25,_
 ⇔starting_moisture=0.80, application_rate=0):
    Simulate daily drying and return DataFrame with moisture content \Box
 \hookrightarrow predictions.
    n n n
    # Ensure datetime and sort
    df['datetime'] = pd.to_datetime(df['datetime'])
    df = df[df['datetime'] >= pd.to_datetime(startdate)].copy()
    df.sort_values('datetime', inplace=True)
    # Calculate VPD
    df = calculate_vapor_pressure_deficit(df)
    # Initialize columns
    moisture_contents = [starting_moisture]
    drying_rates = []
    current_moisture = starting_moisture
    for idx, row in df.iterrows():
        day_number = len(moisture_contents) - 1
        DAY = 1 if day_number == 0 else 0
        SI = row['peak_solarradiation']
        VPD = row['vapor_pressure_deficit']
        SM = 100 * row['soilmoisturevol01']
        SD = 450 #swath density conversion(swath density)
        AR = 0 #application rate
        k = calculate_drying_rate_constant(SI, VPD, DAY, SM, SD, AR)
        # Update moisture content
        current_moisture *= np.exp(-k)
        moisture contents.append(current moisture)
        drying_rates.append(k)
        # Stop if moisture is below 0.5
        if current_moisture <= 0.08:</pre>
            break
    # Create result DataFrame
    result_df = df.iloc[:len(moisture_contents)-1].copy()
    result_df['drying_rates'] = drying_rates
    result_df['predicted_moisture'] = moisture_contents[:-1]
    result_df = result_df.dropna(subset=['predicted_moisture'])
```

```
return result_df

def calculate_drying_rate_constant(SI, VPD, DAY, SM, SD, AR=0):

SI = solar insolation, W/m^2

VPD = vapor pressure deficit, kPA

DAY = 1 for first day, 0 otherwise

SM = soil moisture content, % dry basis

SD = swath density, g/m^2

AR = application rate of chemical solution, g_solution/g_dry-matter

'''

drying_rate = ((SI * (1. + 9.03*AR)) + (43.8 * VPD)) / ((61.4 * SM) + SD * (1.

482 - 0.83 * DAY) * ((1.68 + 24.8 * AR)) + 2767)

return drying_rate

# API Key

if __name__ == "__main__":

API_KEY = "BPFV62G5X9SUQB8HLPPFHNYFV"

interactive_weather_data_viewer(API_KEY)
```

## 0.1 Scratch work and testing

```
[6]: d_df, h_df = fetch_drying_weather_data_from_api("BPFV62G5X9SUQB8HLPPFHNYFV", □

→"San Joaquin Valley", "CA", "2024-06-01", "2024-08-01", unit_group="us")

merged_df = merge_dfs(d_df, h_df)

df2 = predict_moisture_content(merged_df, "2024-06-01")

df2
```

```
[6]:
         datetime temp
                         dew soilmoisturevol01 peak_solarradiation
                                                                      temp_C \
    0 2024-06-01 63.4
                        56.0
                                         0.153
                                                            413.0 17.444444
    1 2024-06-02 63.9 55.8
                                         0.150
                                                            589.0 17.722222
    2 2024-06-03 65.2 55.7
                                                            805.0
                                                                  18.44444
                                         0.148
    3 2024-06-04 65.9
                        56.8
                                         0.148
                                                            850.0 18.833333
    4 2024-06-05 64.8 58.2
                                         0.148
                                                            517.0 18.222222
    5 2024-06-06 64.8 59.5
                                         0.145
                                                            381.0 18.222222
    6 2024-06-07 64.7 58.9
                                         0.143
                                                            335.0 18.166667
    7 2024-06-08 65.5 57.5
                                                            619.0 18.611111
                                         0.144
    8 2024-06-09 65.3 56.9
                                         0.143
                                                            896.0 18.500000
    9 2024-06-10 67.0 57.7
                                         0.141
                                                            816.0 19.444444
    10 2024-06-11 65.9 59.3
                                         0.139
                                                            304.0 18.833333
    11 2024-06-12 66.2 58.7
                                         0.140
                                                            907.0 19.000000
```

```
12 2024-06-13
                     66.1
                           57.1
                                              0.139
                                                                     879.0
                                                                            18.944444
                     67.1
                           57.7
     13 2024-06-14
                                              0.137
                                                                     894.0
                                                                            19.500000
     14 2024-06-15
                     66.9
                           59.7
                                               0.136
                                                                     869.0
                                                                            19.388889
     15 2024-06-16
                     67.4
                           58.2
                                               0.136
                                                                     864.0
                                                                            19.666667
     16 2024-06-17
                     65.7
                           55.7
                                               0.136
                                                                     819.0
                                                                            18.72222
             dew C
                     saturation_vapor_pressure actual_vapor_pressure
     0
         13.333333
                                       1.992984
                                                                1.530743
     1
         13.22222
                                       2.028215
                                                                1.519682
     2
         13.166667
                                       2.122378
                                                                1.514178
     3
         13.777778
                                       2.174649
                                                                1.575697
     4
         14.555556
                                       2.093005
                                                                1.657158
     5
         15.277778
                                       2.093005
                                                                1.736084
     6
         14.944444
                                       2.085718
                                                                1.699256
     7
         14.166667
                                       2.144644
                                                                1.615977
     8
         13.833333
                                       2.129777
                                                                1.581397
     9
         14.277778
                                       2.259066
                                                                1.627650
     10
        15.166667
                                       2.174649
                                                                1.723731
     11
         14.833333
                                       2.197393
                                                                1.687134
     12
         13.944444
                                       2.189789
                                                                1.592851
     13
         14.277778
                                       2.266880
                                                                1.627650
     14
         15.388889
                                       2.251275
                                                                1.748514
     15
         14.555556
                                       2.290466
                                                                1.657158
         13.166667
     16
                                       2.159601
                                                                1.514178
         vapor_pressure_deficit
                                   drying rates
                                                  predicted moisture
     0
                        0.462241
                                       0.097252
                                                            0.800000
     1
                        0.508532
                                       0.120712
                                                            0.725862
     2
                        0.608200
                                       0.164628
                                                             0.643323
     3
                                       0.173455
                        0.598952
                                                             0.545673
     4
                        0.435847
                                       0.106122
                                                             0.458777
     5
                        0.356921
                                       0.078803
                                                             0.412585
     6
                        0.386462
                                       0.070092
                                                             0.381320
     7
                        0.528667
                                       0.127739
                                                             0.355508
     8
                        0.548381
                                       0.183236
                                                             0.312877
     9
                        0.631415
                                       0.168439
                                                             0.260492
                                       0.064797
     10
                        0.450918
                                                            0.220111
     11
                        0.510260
                                       0.185776
                                                            0.206301
     12
                        0.596938
                                       0.181160
                                                            0.171325
     13
                        0.639230
                                       0.184988
                                                             0.142936
     14
                        0.502760
                                       0.178993
                                                             0.118796
     15
                        0.633308
                                       0.179137
                                                             0.099327
     16
                        0.645422
                                       0.170204
                                                             0.083037
[7]: def swath_density_conversion(plants_per_sqft, g_per_plant=5):
          '''convert swath density from plants/ft^2 to g/m^2'''
         plants_per_sqm = plants_per_sqft * 10.764
```

```
return plants_per_sqm * g_per_plant # returns g/m²
out = swath_density_conversion(25)
print(out)
```

## 1345.499999999998

## 0.17755152174485223