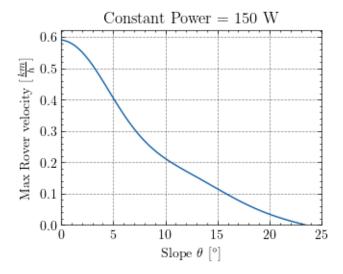
Terrain Simulation

March 12, 2024

```
[]: import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     import scienceplots
     plt.style.use(['science', 'grid']) # , 'nature'
     #plt.rcParams['figure.dpi'] = 600
     plt.rcParams['savefig.dpi'] = 1200
     n_wheel = 4 # number of wheels
     diameter = 0.8
     radius = diameter/2
     b width = 0.35 # wheel width in m
     h_grouser = 0.03 # grouser height
     n_grouser = 8 # number of grousers
     eta = 0.7 # gearbox efficiency
     power_constant = 150*eta # constant power for the whole locomotion subsystem
     v_rover_constant = 0.05 # km/h
     c_b = 170 \# N/m^2 coefficient of soil/wheel cohesion
     n = 1 # exponent of soil deformation
     k c = 1400 \# N/m^2 cohesive modulus of soil deformation
     k_phi = 830000 # N/m^3 frictional modulus of soil deformation
     K = 0.018 # coefficient of soil slip in m
     gamma = 2470 # N/m3 lunar soil density
     phi = 33 # deg soil internal friction angle
     phi_rad = np.deg2rad(phi) # rad soil internal friction angle
     mu = np.tan(phi_rad) # static friction coefficient
     N_q = 32.23
     N_c = 48.09
     N_{gamma} = 33.27
     K_c = 33.37
     K_{phi} = 72.77
     K_gamma = 81.93
     mass = 1800 # mass of the rover
     gravity = 1.62 # gravitational constant of the moon
     slope = np.linspace(0, 25, 100)
     \#slope = 0
```

```
slip = 0.0353 + 1/(1+67*np.exp(-0.32*slope))
slip\_old = 0.1 + (0.01 * slope) # Adjust formula to fitting experimental wheel_{\square}
\rightarrowparameters
\#slip = 1
k = (k_c / b_width) + k_phi
c f = 0.05 # internal rolling friction coefficient (bearings)
W_vehicle = mass * gravity
W_vehicle_normal = W_vehicle * np.cos(np.deg2rad(slope))
W_wheel = mass * gravity / n_wheel
W_wheel_normal = W_wheel * np.cos(np.deg2rad(slope))
z_{sinkage} = (3 * W_wheel_normal / ((3 - n) * (k_c + b_width * k_phi) * np.
\hookrightarrowsqrt(diameter))) ** (2/(2*n+1))
pressure = k * z_sinkage ** n
theta_1 = np.arccos(1 - z_sinkage / radius)
theta_m = (0.45 + 0.24 * slip) * theta_1
lever_e = radius * np.sin(theta_1 - theta_m)
lever_l = radius * np.cos(theta_1 - theta_m)
1_contact = (diameter / 2) * np.arccos(1 - (2 * z_sinkage / diameter))
A = b_width * l_contact
alpha = np.arccos(1 - 2 * z_sinkage / diameter) # angle of approach
1_0 = z_sinkage * np.tan(np.pi / 4 - phi_rad / 2) ** 2 # distance of rupture
R_{\text{compression}} = ((b_{\text{width}} * k) / (n+1)) * z_{\text{sinkage}} ** (n+1)
R bulldozing = (b width*np.sin(alpha+phi rad) / (2*np.sin(alpha)*np.
→cos(phi_rad))) * (2*z_sinkage*c_b*K_c + gamma*K_gamma*z_sinkage**2) +
((gamma*1_0**3)/3) * (np.pi/2 - phi_rad) + (c_b*1_0**2) * (1+np.tan(np.pi/4_0))
 →+ phi_rad/2))
R_gravitational = W_wheel * np.sin(np.deg2rad(slope))
R_rolling = W_wheel_normal * c_f # internal (bearings)
sum_R = 4* R_compression + 2* R_bulldozing + 4* R_gravitational + 4* R_rolling
n_grouser_contact = 2 * np.pi * radius / (n_grouser * l_contact)
sum_H_grouser = 4* (A*c_b*(1+2*h_grouser/b_width)*n_grouser_contact +_
 →W_wheel_normal*mu*(1 + (0.64*h_grouser/b_width) * np.arctan(b_width/
→h_grouser))) * (1-np.exp(-slip * l_contact / K))
W_{\text{wheel}} = A * (c_b*N_c + gamma*z_sinkage*N_q + gamma*b_width*N_gamma/2)
pressure_soil_safe = W_wheel_safe / A
pressure_real = W_wheel / A
pressure_margin = pressure_real / pressure_soil_safe
weight_margin = W_wheel / W_wheel_safe
drawbar_pull = sum_H_grouser - sum_R
```

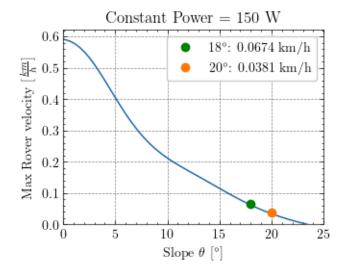
```
plt.plot(slope, v_rover)
plt.xlabel('Slope $\\theta$ [$^\circ$]')
plt.ylabel('Max Rover velocity [$\\frac{km}{h}$]')
plt.xlim(0, 25)
plt.ylim(0,)
plt.title('Constant Power = {:.0f} W'.format(power_constant/eta))
plt.savefig("max_rover_velocity_slope.pdf")
plt.savefig("max_rover_velocity_slope.png")
plt.show()
```



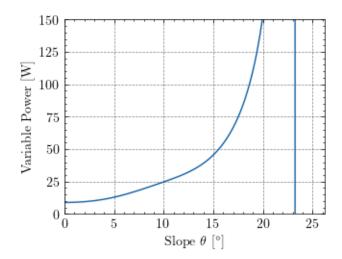
```
[]: plt.plot(slope, v_rover)
   plt.xlabel('Slope $\\theta$ [$^\circ$]')
   plt.ylabel('Max Rover velocity [$\\frac{km}{h}$]')
   plt.xlim(0, 25)
   plt.ylim(0,)
   plt.title('Constant Power = {:.0f} W'.format(power_constant/eta))

# Mark points at 18 and 20 degrees
```

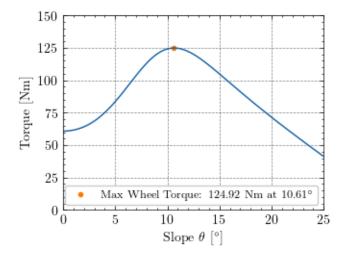
```
plt.plot(18, v_rover[np.where(np.round(slope) == 18)][0], 'o', color='green', \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \
```



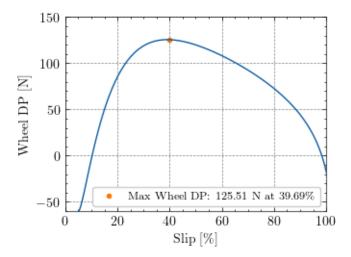
```
[]: plt.plot(slope, power_variable)
  plt.xlabel('Slope $\\theta$ [$^\circ$]')
  plt.ylabel('Variable Power [W]')
  plt.xlim(0,)
  plt.ylim(0,150)
  plt.savefig("power_variable_slope.pdf")
  plt.savefig("power_variable_slope.png")
  plt.show()
```

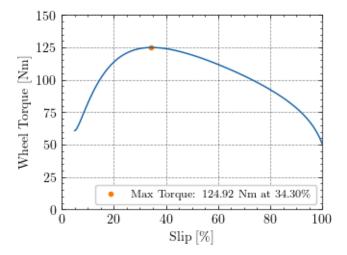


```
[]: # Find the index of the maximum torque
     max_torque_index = np.argmax(torque)
     max_torque_value = torque[max_torque_index]
     max_torque_theta = slope[max_torque_index]
     plt.plot(max_torque_theta, max_torque_value, 'o', color=(255/255, 120/255, 0/
      \hookrightarrow255), markersize=3)
     plt.plot(slope, torque)
     plt.xlabel('Slope $\\theta$ [$^\circ$]')
     plt.ylabel('Torque [Nm]')
     plt.xlim(0, 25)
     plt.ylim(0,150)
     plt.legend([f'Max Wheel Torque: {max_torque_value:.2f} Nm at {max_torque_theta:.
      →2f}$^\circ$'], loc='lower right', fontsize='small')
     plt.savefig("max_torque_slope.pdf")
     plt.savefig("max_torque_slope.png")
     plt.show()
```

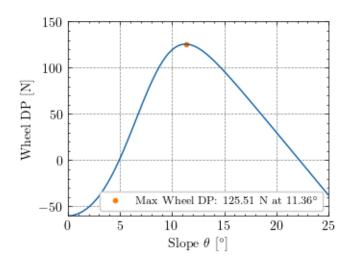


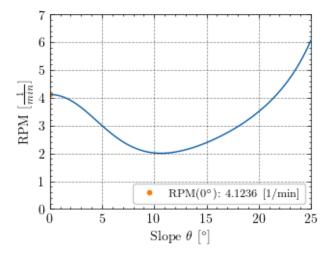
```
[]: # Find the index of the maximum torque
     max_dp_index = np.argmax(drawbar_pull)
     max_dp_value = drawbar_pull[max_dp_index]/n_wheel
     max_dp_slip = slip[max_dp_index]*100
     plt.plot(max_dp_slip, max_dp_value, 'o', color=(255/255, 120/255, 0/255),__
      →markersize=3)
    plt.plot(slip*100, drawbar_pull/n_wheel)
     plt.xlabel('Slip [\%]')
     plt.ylabel('Wheel DP [N]')
     plt.xlim(0, 100)
     plt.ylim(min(drawbar_pull/n_wheel), 150)
     plt.legend([f'Max Wheel DP: {max_dp_value:.2f} N at {max_dp_slip:.2f}$\%$'],_
      →loc='lower right', fontsize='small')
     plt.savefig("max_drawbar_pull_slip.pdf")
     plt.savefig("max_drawbar_pull_slip.png")
     plt.show()
```

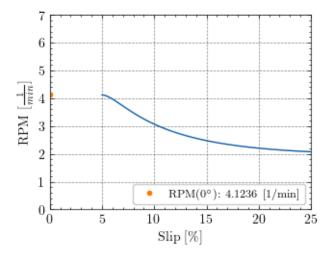




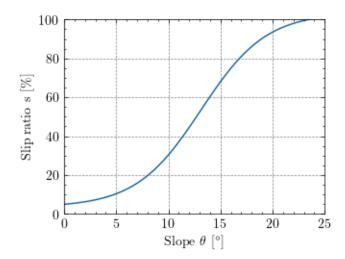
```
[]: # Find the index of the maximum torque
    max_dp_index = np.argmax(drawbar_pull)
    max_dp_value = drawbar_pull[max_dp_index]/n_wheel
    max_dp_theta = slope[max_dp_index]
    plt.plot(max_dp_theta, max_dp_value, 'o', color=(255/255, 120/255, 0/255),__
     →markersize=3)
    plt.plot(slope, drawbar_pull/n_wheel)
    plt.xlabel('Slope $\\theta$ [$^\circ$]')
    plt.ylabel('Wheel DP [N]')
    plt.xlim(0, 25)
    plt.ylim(min(drawbar_pull/n_wheel), 150)
    plt.legend([f'Max Wheel DP: {max_dp_value:.2f} N at {max_dp_theta:.
     plt.savefig("max_drawbar_pull_slope.pdf")
    plt.savefig("max_drawbar_pull_slope.png")
    plt.show()
```







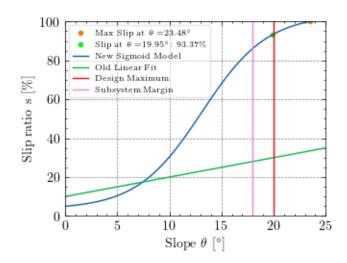
```
[]: plt.plot(slope, slip*100)
  plt.xlabel('Slope $\\theta$ [$^\circ$]')
  plt.ylabel('Slip ratio s [\%]')
  plt.xlim(0, 25)
  plt.ylim(0, 100)
  plt.savefig("sigmoid_slip.pdf")
  plt.savefig("sigmoid_slip.png")
  plt.show()
```



```
[]: max_s_index = np.argmax(np.clip(slip, 0, 1))
      max s value = slip[max s index]
      max_s_slope = slope[max_s_index]
      plt.plot(max_s_slope, max_s_value*100, 'o', color=(255/255, 120/255, 0/255),__
        markersize=3, label=f'Max Slip at $\\theta=${max_s_slope:.2f}$^\circ$')
      # Find the index of the slope closest to 20 degrees
      design s index = np.abs(slope - 20).argmin()
      design s slope = slope[design s index]
      design_s_value = slip[design_s_index]
      plt.plot(design_s_slope, design_s_value*100, 'o', color='lime', markersize=3,__
        Garage Salabel=f'Slip at $\\theta=${design_s_slope:.2f}$^\circ$: {design_s_value*100:.

      plt.plot(slope, slip*100, label='New Sigmoid Model')
      plt.plot(slope, slip_old*100, label='Old Linear Fit')
      plt.plot(np.linspace(20, 20, len(slope)), np.linspace(0, 100, len(slope)),
       ⇔color='red', label='Design Maximum')
      plt.plot(np.linspace(18, 18, len(slope)), np.linspace(0, 100, len(slope)),

¬color='violet', label='Subsystem Margin')
      plt.xlabel('Slope $\\theta$ [$^\circ$]')
      plt.ylabel('Slip ratio s [\%]')
      plt.xlim(0, 25)
      plt.ylim(0, 100)
      plt.legend(fancybox=True, framealpha=0.5, loc='best', fontsize='6')
      plt.savefig("sigmoid_slip_comparison.pdf")
      plt.savefig("sigmoid_slip_comparison.png")
      plt.show()
```



```
[]: def func(slope, v_rover_constant, mass, diameter, power_constant, b_width,_
      ⇒gravity, h_grouser, n_grouser):
         n_wheel = 4 # number of wheels
         \#diameter = 0.8
         radius = diameter/2
         \#b\_width = 0.35 \# wheel width in m
         #h_grouser = 0.03 # grouser height
         #n_grouser = 8 # number of grousers
         eta = 0.7 # gearbox efficiency
         #power constant = 150*eta # constant power for the whole locomotion
      ⇔subsystem
         #v\_rover\_constant = 0.05 # km/h
         c_b = 170 \# N/m^2 coefficient of soil/wheel cohesion
         n = 1 # exponent of soil deformation
         k_c = 1400 \# N/m^2 cohesive modulus of soil deformation
         k_phi = 830000 # N/m^3 frictional modulus of soil deformation
         K = 0.018 # coefficient of soil slip in m
         gamma = 2470 # N/m3 lunar soil density
         phi = 33 # deg soil internal friction angle
         phi_rad = np.deg2rad(phi) # rad soil internal friction angle
         mu = np.tan(phi_rad) # static friction coefficient
         N_q = 32.23
         N c = 48.09
         N_{gamma} = 33.27
         K c = 33.37
         K_{phi} = 72.77
         K_gamma = 81.93
         #mass = 1800 # mass of the rover
         #gravity = 1.62 # gravitational constant of the moon
         \#slope = np.linspace(0, 25, 100)
```

```
\#slope = 0
  slip = 0.0353 + 1/(1+67*np.exp(-0.32*slope))
  \#slip = 1
  k = (k_c / b_width) + k_phi
  c_f = 0.05 # internal rolling friction coefficient (bearings)
  W_vehicle = mass * gravity
  W_vehicle_normal = W_vehicle * np.cos(np.deg2rad(slope))
  W_wheel = mass * gravity / n_wheel
  W_wheel_normal = W_wheel * np.cos(np.deg2rad(slope))
  z_{sinkage} = (3 * W_{wheel_normal} / ((3 - n) * (k_c + b_{width} * k_phi) * np.
\rightarrowsqrt(diameter))) ** (2/(2*n+1))
  pressure = k * z_sinkage ** n
  theta_1 = np.arccos(1 - z_sinkage / radius)
  theta_m = (0.45 + 0.24 * slip) * theta_1
  lever_e = radius * np.sin(theta_1 - theta_m)
  lever_l = radius * np.cos(theta_1 - theta_m)
  1_contact = (diameter / 2) * np.arccos(1 - (2 * z_sinkage / diameter))
  A = b width * l contact
  alpha = np.arccos(1 - 2 * z_sinkage / diameter) # angle of approach
  1_0 = z_sinkage * np.tan(np.pi / 4 - phi_rad / 2) ** 2 # distance of rupture
  R_{\text{compression}} = ((b_{\text{width}} * k) / (n+1)) * z_{\text{sinkage}} ** (n+1)
  R_bulldozing = (b_width*np.sin(alpha+phi_rad) / (2*np.sin(alpha)*np.
→cos(phi_rad))) * (2*z_sinkage*c_b*K_c + gamma*K_gamma*z_sinkage**2) +
((gamma*l_0**3)/3) * (np.pi/2 - phi_rad) + (c_b*l_0**2) * (1+np.tan(np.pi/4)
→+ phi_rad/2))
  R_gravitational = W_wheel * np.sin(np.deg2rad(slope))
  R_rolling = W_wheel_normal * c_f # internal (bearings)
  sum_R = 4* R_compression + 2* R_bulldozing + 4* R_gravitational + 4*_
→R rolling
  n_grouser_contact = 2 * np.pi * radius / (n_grouser * l_contact)
  sum_H_grouser = 4* (A*c_b*(1+2*h_grouser/b_width)*n_grouser_contact +__
www.wheel_normal*mu*(1 + (0.64*h_grouser/b_width) * np.arctan(b_width/
→h_grouser))) * (1-np.exp(-slip * l_contact / K))
  W_wheel_safe = A * (c_b*N_c + gamma*z_sinkage*N_q + gamma*b_width*N_gamma/2)
  pressure_soil_safe = W_wheel_safe / A
  pressure_real = W_wheel / A
  pressure_margin = pressure_real / pressure_soil_safe
  weight_margin = W_wheel / W_wheel_safe
  drawbar_pull = sum_H_grouser - sum_R
```

```
torque = (((drawbar_pull * lever_l) + (W_vehicle_normal * lever_e))/

on_wheel) + 28

omega = power_constant / (torque * n_wheel)

rpm = omega * 60/ (2 * np.pi)

v_rover = omega * radius * 3.6 * (1-slip)

power_variable = (v_rover_constant * torque * n_wheel) / (radius * 3.6 * \( \frac{1}{2} \) - slip))

power_variable_step = np.clip(50 * np.ceil(np.clip(power_variable,30,100) / \( \frac{1}{2} \) - 30), 50, 100)

return v_rover, power_variable, power_variable_step, drawbar_pull, torque, \( \frac{1}{2} \) - rpm, pressure, pressure_soil_safe, pressure_real
```

```
[]: data = pd.read_csv('Pfad_komplett_Steigung_V2.csv')
     distance = data['position']
     longitude = -data['lon']
     latitude = -data['lat']
     slope = data['TerrainSlope']
     data['position_difference'] = data['position'].diff()
     data['slope_difference'] = data['TerrainSlope'].diff()
     solution = func(slope, v_rover_constant, mass, diameter, power_constant,_
      ⇒b_width, gravity, h_grouser, n_grouser)
     data['velocity'] = solution[0]
     data['weighted_velocity'] = data['position_difference'] * data['velocity']
     data['weighted_slope'] = data['position_difference'] * data['TerrainSlope']
     mean velocity = data['weighted velocity'].sum() / data['position_difference'].
      ⇒sum()
     mean_slope = data['weighted_slope'].sum() / data['position_difference'].sum()
     end distance = data['position'].iloc[-1]
     data['time'] = data['position'].diff().sum() / (data['weighted_velocity'].
     ⇒sum() / data['position_difference'].sum())
     total_time = end_distance / mean_velocity
     print("Mean Velocity: {:.4f} km/h".format(mean_velocity))
     print("Mean Slope: {:.4f} ".format(mean_slope))
     print("End Distance: {:.4f} km".format(end_distance))
     print("Total Time: {:.4f} h".format(total_time))
```

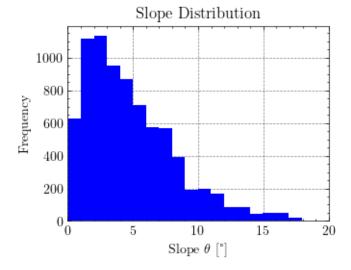
Mean Velocity: 0.4048 km/h Mean Slope: 5.3480 ° End Distance: 128.4204 km Total Time: 317.2371 h

```
[]: v_rover_constant = distance.iloc[-1] / ((0.5 - 0.25) * 365 * 24) v_rover_constant
```

[]: 0.058639446575342466

```
[]: plt.hist(slope, bins=18, density=False, color='b')

# Add labels and title
plt.xlabel('Slope $\\theta$ [°]')
plt.ylabel('Frequency')
plt.xlim(0, 20)
plt.title('Slope Distribution')
plt.savefig("slope_histogram.pdf")
plt.savefig("slope_histogram.png")
# Show the plot
plt.show()
```



```
[]: from matplotlib import colors
from matplotlib.ticker import PercentFormatter

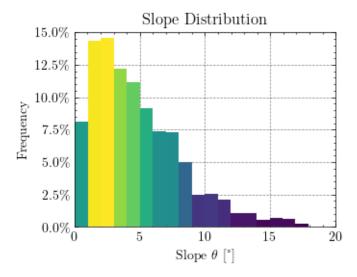
# Number of bins
n_bins = 18 #*2

# First subplot: Color-coded histogram
N, bins, patches = plt.hist(slope, bins=n_bins, density=True)

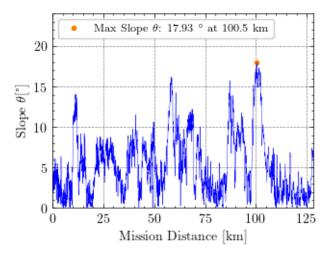
# Normalize the data for color-coding
fracs = N / N.max()
norm = colors.Normalize(fracs.min(), fracs.max())
```

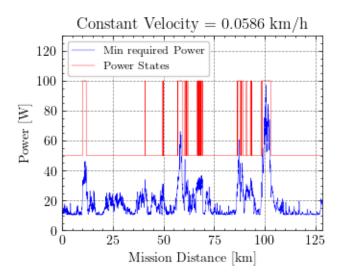
```
# Loop through patches and set color based on height
for thisfrac, thispatch in zip(fracs, patches):
    color = plt.cm.viridis(norm(thisfrac))
    thispatch.set_facecolor(color)

# Add labels and title
plt.xlabel('Slope $\\theta$ [°]')
plt.ylabel('Frequency')
plt.xlim(0, 20)
plt.ylim(0, 0.15)
plt.title('Slope Distribution')
plt.gca().yaxis.set_major_formatter(PercentFormatter(1))
plt.savefig("slope_histogram_color.pdf")
plt.savefig("slope_histogram_color.png")
plt.show()
```

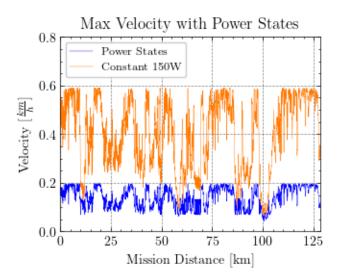


```
plt.savefig("mission_distance_slope.pdf")
plt.savefig("mission_distance_slope.png")
plt.show()
```

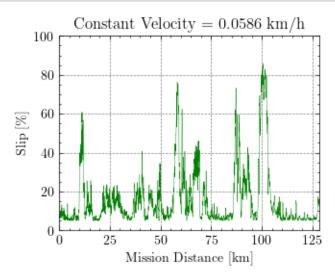




```
[]: power_states = func(slope, v_rover_constant, 1800, 0.8, 150*0.7, 0.35, 1.62, 0.
     ⇔03, 8)[2]
     plt.plot(distance, func(slope, v_rover_constant, 1800, 0.8, power_states*0.7, 0.
      →35, 1.62, 0.03, 8)[0], color='b', linewidth=0.3, label='Power States')
     plt.plot(distance, func(slope, v_rover_constant, 1800, 0.8, 150*0.7, 0.35, 1.
      →62, 0.03, 8)[0], color=(255/255, 120/255, 0/255), linewidth=0.3,
      ⇔label='Constant 150W')
     plt.xlabel('Mission Distance [km]')
     plt.ylabel('Velocity [$\\frac{km}{h}$]')
     plt.xlim(0, distance.iloc[-1])
     plt.ylim(0, 0.8)
     plt.title('Max Velocity with Power States')
     plt.legend(loc='upper left', fontsize='small')
     plt.savefig("power_states_over_mission_distance.pdf")
     plt.savefig("power_states_over_mission_distance.png")
     plt.show()
```



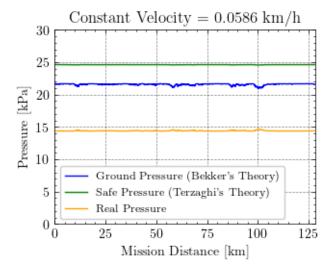
```
[]: slip = 0.0353 + 1/(1+67*np.exp(-0.32*slope))
   plt.plot(distance, slip*100, linewidth=0.3, color='green')
   plt.xlabel('Mission Distance [km]')
   plt.ylabel('Slip [\%]')
   plt.xlim(0, distance.iloc[-1])
   plt.ylim(0, 100)
   plt.title('Constant Velocity = {:.4f} km/h'.format(v_rover_constant))
   plt.savefig("slip_over_mission_distance.pdf")
   plt.savefig("slip_over_mission_distance.png")
   plt.show()
```



```
[]: plt.plot(distance, func(slope, v_rover_constant, 1800, 0.8, 150*0.7, 0.35, 1.
      462, 0.03, 8)[6]/1000, color='blue', linewidth=1, label='Ground Pressure⊔

    Gekker\'s Theory)')

    plt.plot(distance, func(slope, v_rover_constant, 1800, 0.8, 150*0.7, 0.35, 1.
      →62, 0.03, 8)[7]/1000, color='green', linewidth=1, label='Safe Pressure
      plt.plot(distance, func(slope, v_rover_constant, 1800, 0.8, 150*0.7, 0.35, 1.
      -62, 0.03, 8)[8]/1000, color='orange', linewidth=1, label='Real Pressure')
    plt.xlabel('Mission Distance [km]')
    plt.ylabel('Pressure [kPa]')
    plt.xlim(0, distance.iloc[-1])
    plt.ylim(0, 30)
    plt.title('Constant Velocity = {:.4f} km/h'.format(v_rover_constant))
    plt.legend(fontsize='small')
    plt.savefig("pressure_over_mission_distance.pdf")
    plt.savefig("pressure_over_mission_distance.png")
    plt.show()
```



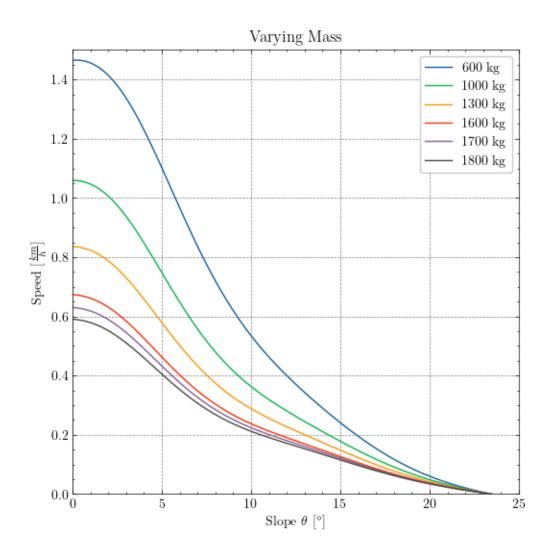
```
[]: slope = np.linspace(0, 25, 100) # range of slope values

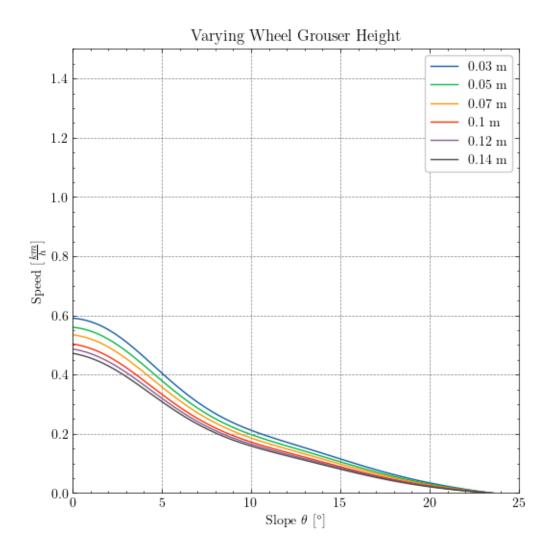
# Define values for mass, slip, radius, and power
mass_values = [600, 1000, 1300, 1600, 1700, 1800]
h_grouser_values = [0.03, 0.05, 0.07, 0.1, 0.12, 0.14]
diameter_values = [0.4, 0.6, 0.8, 1]
power_values = [50, 100, 150, 200, 250]
width_values = [0.1, 0.2, 0.3, 0.4, 0.5]
gravity_values = [1.62, 3.71, 9.81]
gravity_name_values = ['Moon', 'Mars', 'Earth']
n_grouser_values = [8, 16, 24, 32]
```

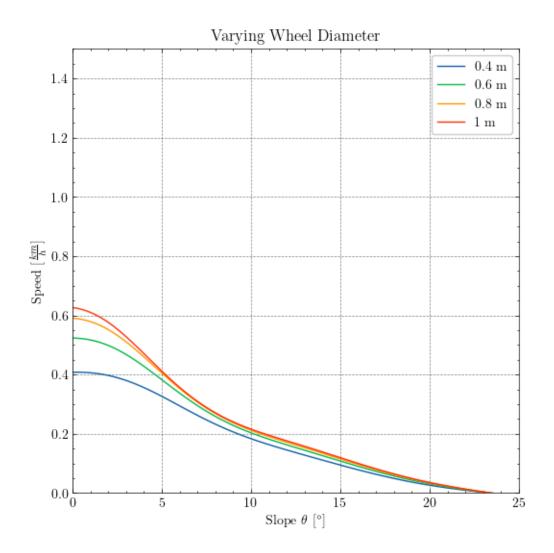
```
# Create subplots for Varying Mass
fig1, ax1 = plt.subplots(figsize=(6, 6))
for mass_n in mass_values:
    ax1 plot(slope, func(slope, v_rover_constant, mass n, diameter, __
 ⇒power_constant, b_width, gravity, h_grouser, n_grouser)[0], label=f"{mass_n}_∪
⊸kg")
ax1.legend()
ax1.set_title("Varying Mass")
ax1.set_xlabel('Slope $\\theta$ [$^\circ$]')
ax1.set_ylabel('Speed [$\\frac{km}{h}$]')
ax1.set_xlim(0, 25)
ax1.set_ylim(0, 1.5)
plt.savefig("tire_sim_varying_mass.pdf")
plt.savefig("tire_sim_varying_mass.png")
plt.show()
# Create subplots for Varying Grouser Height
fig2, ax2 = plt.subplots(figsize=(6, 6))
for h_grouser_n in h_grouser_values:
    ax2.plot(slope, func(slope, v rover constant, mass, diameter,
⇒power_constant, b_width, gravity, h_grouser_n, n_grouser)[0],
 ⇒label=f"{h_grouser_n} m")
ax2.legend()
ax2.set_title("Varying Wheel Grouser Height")
ax2.set_xlabel('Slope $\\theta$ [$^\circ$]')
ax2.set ylabel('Speed [$\\frac{km}{h}$]')
ax2.set xlim(0, 25)
ax2.set_ylim(0, 1.5)
plt.savefig("tire_sim_varying_grouser_height.pdf")
plt.savefig("tire_sim_varying_grouser_height.png")
plt.show()
# Create subplots for Varying Radius
fig3, ax3 = plt.subplots(figsize=(6, 6))
for diameter n in diameter values:
    ax3.plot(slope, func(slope, v_rover_constant, mass, diameter_n,_
 ⇒power_constant, b_width, gravity, h_grouser, n_grouser)[0], ⊔
 ⇔label=f"{diameter_n} m")
ax3.legend()
ax3.set_title("Varying Wheel Diameter")
ax3.set xlabel('Slope $\\theta$ [$^\circ$]')
ax3.set_ylabel('Speed [$\\frac{km}{h}$]')
ax3.grid(True)
ax3.set_xlim(0, 25)
ax3.set_ylim(0, 1.5)
plt.savefig("tire_sim_varying_diameter.pdf")
```

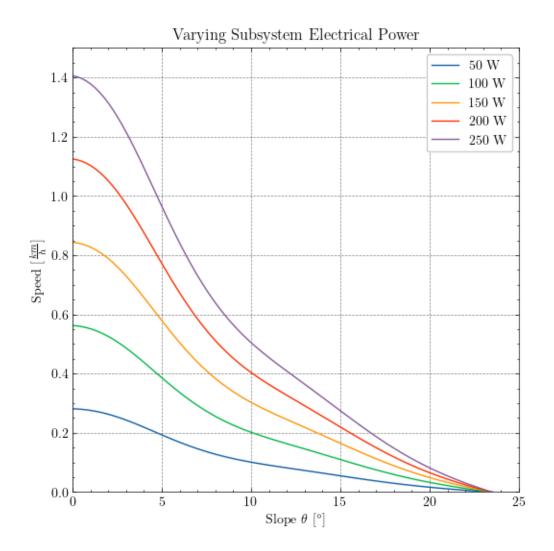
```
plt.savefig("tire_sim_varying_diameter.png")
plt.show()
# Create subplots for Varying Power
fig4, ax4 = plt.subplots(figsize=(6, 6))
for power_n in power_values:
   ax4.plot(slope, func(slope, v_rover_constant, mass, diameter, power_n,_u
⇒b_width, gravity, h_grouser, n_grouser)[0], label=f"{power_n} W")
ax4.legend()
ax4.set_title("Varying Subsystem Electrical Power")
ax4.set_xlabel('Slope $\\theta$ [$^\circ$]')
ax4.set_ylabel('Speed [$\\frac{km}{h}$]')
ax4.grid(True)
ax4.set_xlim(0, 25)
ax4.set vlim(0, 1.5)
plt.savefig("tire_sim_varying_power.pdf")
plt.savefig("tire sim varying power.png")
plt.show()
# Create subplots for Varying Width
fig5, ax5 = plt.subplots(figsize=(6, 6))
for width n in width values:
    ax5.plot(slope, func(slope, v_rover_constant, mass, diameter,_
 ⇒power_constant, width_n, gravity, h_grouser, n_grouser)[0], ⊔
 ⇔label=f"{width_n} m")
ax5.legend()
ax5.set title("Varying Wheel Width")
ax5.set xlabel('Slope $\\theta$ [$^\circ$]')
ax5.set_ylabel('Speed [$\\frac{km}{h}$]')
ax5.set xlim(0, 25)
ax5.set_ylim(0, 1.5)
plt.savefig("tire sim varying width.pdf")
plt.savefig("tire_sim_varying_width.png")
plt.show()
# Create subplots for Varying Planet
fig6, ax6 = plt.subplots(figsize=(6, 6))
for gravity val, gravity name in zip(gravity values, gravity name values):
   ax6.plot(slope, func(slope, v_rover_constant, mass, diameter,__
power_constant, b_width, gravity_val, h_grouser, n_grouser)[0],
→label=f"{gravity_name}")
ax6.legend()
ax6.set title("Varying Astronomical Body")
ax6.set_xlabel('Slope $\\theta$ [$^\circ$]')
ax6.set_ylabel('Speed [$\\frac{km}{h}$]')
ax6.set_xlim(0, 25)
```

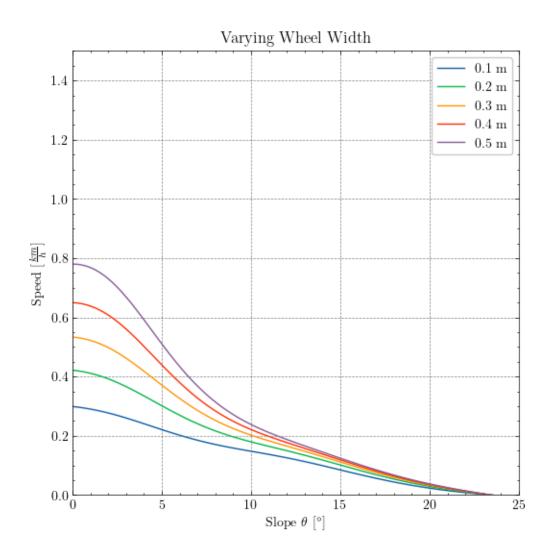
```
ax6.set_ylim(0, 1.5)
plt.savefig("tire_sim_varying_gravity.pdf")
plt.savefig("tire_sim_varying_gravity.png")
plt.show()
# Create subplots for Varying Number of Grousers
fig5, ax7 = plt.subplots(figsize=(6, 6))
for n_grouser_n in n_grouser_values:
   ax7.plot(slope, func(slope, v_rover_constant, mass, diameter,_
⇔power_constant, b_width, gravity, h_grouser, n_grouser_n)[0], ___
→label=f"{n_grouser_n}")
ax7.legend()
ax7.set_title("Varying Number of Grousers")
ax7.set_xlabel('Slope $\\theta$ [$^\circ$]')
ax7.set_ylabel('Speed [$\\frac{km}{h}$]')
ax7.set_xlim(0, 25)
ax7.set_ylim(0, 1.5)
plt.savefig("tire_sim_varying_number_of_grousers.pdf")
plt.savefig("tire_sim_varying_number_of_grousers.png")
plt.show()
```

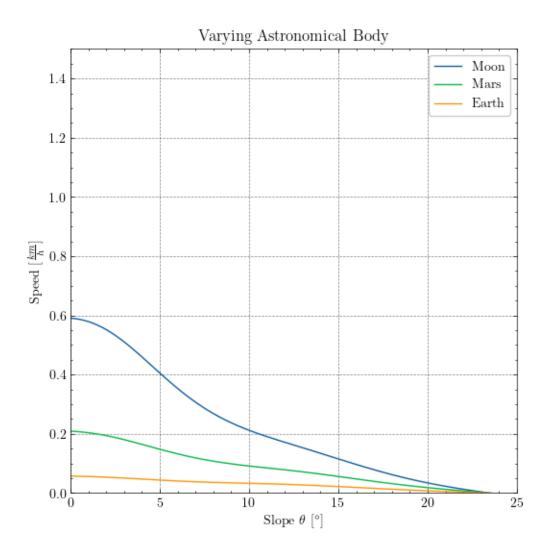


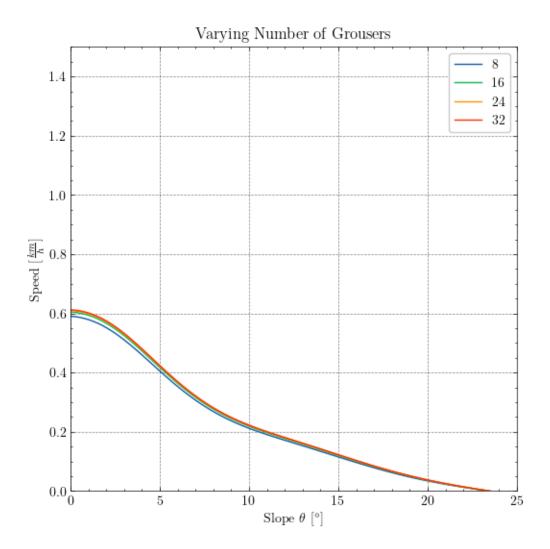












```
# Print optimized results with \n after each value
     print(f"slope min: {slope min: .2f}\nslope mean: {slope min: .2f}\nslope max:__
      \rightarrow{slope_max:.2f}\n"
           f"drawbar_pull_min: {drawbar_pull_min:.2f}\ndrawbar_pull_mean:__
      →{drawbar_pull_mean:.2f}\ndrawbar_pull_max:.{drawbar_pull_max:.2f}\n"
           f"v_min: \{v_min: .4f\}\nv_mean: \{v_mean: .4f\}\nv_max: \{v_max: .4f\}\n"
           f"torque_min: {torque_min:.2f}\ntorque_mean: {torque_mean:.

¬2f}\ntorque_max: {torque_max:.2f}\n"

           f"P_min: \{P_min: .2f\} \\ nP_mean: \{P_mean: .2f\} \\ nP_max: \{P_max: .2f\} \\ n"
           f"rpm_min: {rpm_min:.4f}\nrpm_mean: {rpm_mean:.4f}\nrpm_max: {rpm_max:.

4f}")
    slope_min: 10.00
    slope_mean: 10.00
    slope_max: 20.00
    drawbar_pull_min: -241.65
    drawbar_pull_mean: 204.78
    drawbar_pull_max: 502.09
    v_min: 0.0343
    v_mean: 0.2629
    v max: 0.5907
    torque_min: 60.79
    torque mean: 95.06
    torque_max: 124.91
    P_min: 10.42
    P_mean: 42.84
    P_max: 179.56
    rpm_min: 2.0067
    rpm_mean: 2.7850
    rpm_max: 4.1236
[]: from scipy.interpolate import griddata
     data = pd.read_csv('Pfad_komplett_Steigung_V2.csv')
     distance = data['position']
     longitude = -data['lon']
     latitude = -data['lat']
     slope = data['TerrainSlope']
     # Create a grid spherical
     #xi_sphere = np.linspace(min(longitude), max(longitude), len(longitude))
     #yi sphere = np.linspace(min(latitude), max(latitude), len(latitude))
     #xi_sphere, yi_sphere = np.meshgrid(xi_sphere, yi_sphere)
     #zi sphere = griddata((longitude, latitude), slope, (xi sphere, yi sphere), u
      ⇔method='nearest')
```

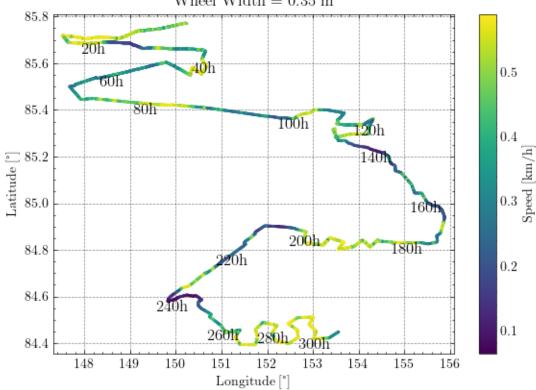
```
solution = func(slope, v_rover_constant, 1800, 0.8, 150*0.7, 0.35, 1.62, 0.03, 1.00 to 1.00 to
  ⇔8)
end_distance = distance.iloc[-1]
v min = np.min(solution[0])
v_max = np.max(solution[0])
data['position_difference'] = data['position'].diff()
data['velocity'] = solution[0]
data['weighted_velocity'] = data['position_difference'] * data['velocity']
mean velocity = data['weighted velocity'].sum() / data['position_difference'].
  ⇒sum()
slope mean = np.mean(data['TerrainSlope'])
end_distance = data['position'].iloc[-1]
total_time = end_distance / mean_velocity
fig, (ax1) = plt.subplots(1, 1, figsize=(6, 6))
#contour1 = ax1.contourf(xi_sphere, yi_sphere, zi_sphere, cmap='viridis')
#fig.colorbar(contour1, ax=ax1, label='Terrain Slope [°]')
scatter1 = ax1.scatter(longitude, latitude, c=solution[0], marker='.', s=5,__
  fig.colorbar(scatter1, ax=ax1, label='Speed [km/h]')
ax1.axis('auto')
margin = 0.2
ax1.set_xlim(min(longitude) - margin, max(longitude) + margin)
ax1.set_ylim(min(latitude) - margin/5, max(latitude) + margin/5)
# Add markers for every k hours
k = 20 \# hours
tolerance = 0.1 # Adjust this value based on your desired tolerance
marker_added = False # Reset the marker flag at the start
for i in range(len(distance)):
        if i > 0:
                v_mean_up_to = np.cumsum(data['weighted_velocity'] /__

data['position_difference'])[i] / i

                time_passed = distance[i] / v_mean_up_to
                # Check if the current time_passed is within a new k-hour interval
                if abs((time_passed % k) / k - 1) < tolerance and not marker_added:
                         label_text = '{}h'.format(round(time_passed / k) * k)
                         \#ax1.plot(longitude[i], latitude[i], 'o', color='white', 
  →markersize=6, markeredgecolor='black')
                         ax1.text(longitude[i], latitude[i], label_text, ha='center', ___
  ⇔va='top', color='black', fontsize=12, alpha=1)
                         marker_added = True # Set the flag to True after adding the marker_
  ⇔and label
```

```
elif abs((time_passed % k) / k - 1) >= tolerance:
            marker_added = False  # Reset the flag if outside the tolerance
ax1.set_xlabel('Longitude [°]')
ax1.set_ylabel('Latitude [°]')
ax1.set_title('Mission Path Map\n'
              'Mean Velocity = \{:.4f\} km/h\n'
              'Min Velocity = \{:.4f\} km/h\n'
              'Max Velocity = \{:.4f\} km/h\n'
              'Min Mission Length = {:.2f} h\n'
              'Total Gearbox and Motor efficiencies $\\eta$ = {:.2f}\n'
              'Constant Electrical Power = {:.0f} W\n'
              'RTS Mass = \{:.0f\} kg\n'
              'Wheel Diameter = {:.2f} m\n'
              'Wheel Width = {:.2f} m'
              .format(mean_velocity, v_min, v_max, total_time, eta,__
→power_constant/eta, mass, diameter, b_width))
plt.tight_layout()
plt.savefig("map_constant_power_variable_velocity.pdf")
plt.savefig("map_constant_power_variable_velocity.png")
plt.show()
```

```
\begin{array}{c} {\rm Mission~Path~Map} \\ {\rm Mean~Velocity} = 0.4048~{\rm km/h} \\ {\rm Min~Velocity} = 0.0634~{\rm km/h} \\ {\rm Max~Velocity} = 0.5906~{\rm km/h} \\ {\rm Min~Mission~Length} = 317.24~{\rm h} \\ {\rm Total~Gearbox~and~Motor~efficiencies~} \eta = 0.70 \\ {\rm Constant~Electrical~Power} = 150~{\rm W} \\ {\rm RTS~Mass} = 1800~{\rm kg} \\ {\rm Wheel~Diameter} = 0.80~{\rm m} \\ {\rm Wheel~Width} = 0.35~{\rm m} \\ \end{array}
```



```
[]: from scipy.interpolate import griddata
  data = pd.read_csv('Pfad_komplett_Steigung_V2.csv')

distance = data['position']
  longitude = -data['lon']
  latitude = -data['lat']
  slope = data['TerrainSlope']

# Create a grid spherical
  #xi_sphere = np.linspace(min(longitude), max(longitude), len(longitude))
  #yi_sphere = np.linspace(min(latitude), max(latitude), len(latitude))
  #xi_sphere, yi_sphere = np.meshgrid(xi_sphere, yi_sphere)
```

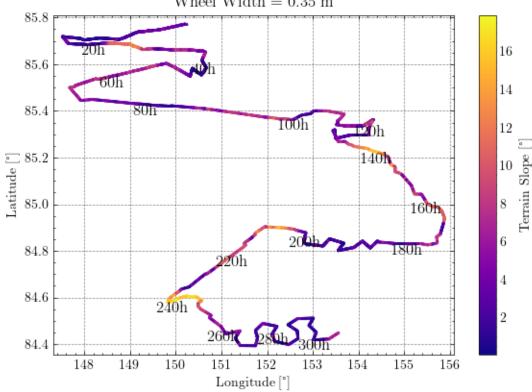
```
#zi sphere = griddata((longitude, latitude), slope, (xi sphere, yi sphere), u
 →method='nearest')
solution = func(slope, v rover constant, 1800, 0.8, 150*0.7, 0.35, 1.62, 0.03,
end_distance = distance.iloc[-1]
v_min = np.min(solution[0])
v_max = np.max(solution[0])
data['position_difference'] = data['position'].diff()
data['velocity'] = solution[0]
data['weighted_velocity'] = data['position_difference'] * data['velocity']
mean_velocity = data['weighted_velocity'].sum() / data['position_difference'].
 Silm()
slope mean = np.mean(data['TerrainSlope'])
end_distance = data['position'].iloc[-1]
total_time = end_distance / mean_velocity
fig, (ax1) = plt.subplots(1, 1, figsize=(6, 6))
scatter1 = ax1.scatter(longitude, latitude, c=slope, marker='.', s=5,_
 fig.colorbar(scatter1, ax=ax1, label='Terrain Slope [°]')
#scatter1 = ax1.scatter(longitude, latitude, c=solution[0], linewidths=0.1,
 →marker='o', s=5, cmap='viridis')
#fig.colorbar(scatter1, ax=ax1, label='Speed [km/h]')
ax1.axis('auto')
margin = 0.2
ax1.set_xlim(min(longitude) - margin, max(longitude) + margin)
ax1.set_ylim(min(latitude) - margin/5, max(latitude) + margin/5)
# Add markers for every k hours
k = 20 \# hours
tolerance = 0.1  # Adjust this value based on your desired tolerance
marker_added = False # Reset the marker flag at the start
for i in range(len(distance)):
    if i > 0:
        v_mean_up_to = np.cumsum(data['weighted_velocity'] /__

data['position_difference'])[i] / i

        time_passed = distance[i] / v_mean_up_to
        # Check if the current time_passed is within a new k-hour interval
        if abs((time_passed % k) / k - 1) < tolerance and not marker_added:
            label_text = '{}h'.format(round(time_passed / k) * k)
            #ax1.plot(longitude[i], latitude[i], 'o', color='white',
 →markersize=6, markeredgecolor='black')
```

```
ax1.text(longitude[i], latitude[i], label_text, ha='center', __
 ⇔va='top', color='black', fontsize=12, alpha=1)
            marker_added = True # Set the flag to True after adding the marker_
 →and label
        elif abs((time_passed % k) / k - 1) >= tolerance:
            marker_added = False # Reset the flag if outside the tolerance
ax1.set_xlabel('Longitude [°]')
ax1.set_ylabel('Latitude [°]')
ax1.set_title('Mission Path Map\n'
              'Mean Velocity = \{:.4f\} km/h\n'
              'Min Velocity = \{:.4f\} km/h\n'
              'Max Velocity = \{:.4f\} km/h\n'
              'Min Mission Length = \{:.2f\} h\n'
              'Total Gearbox and Motor efficiencies $\\eta$ = {:.2f}\n'
              'Constant Electrical Power = {:.0f} W\n'
              'RTS Mass = \{:.0f\} kg\n'
              'Wheel Diameter = {:.2f} m\n'
              'Wheel Width = {:.2f} m'
              .format(mean_velocity, v_min, v_max, total_time, eta, \_
 →power_constant/eta, mass, diameter, b_width))
plt.tight_layout()
plt.savefig("map slope.pdf")
plt.savefig("map_slope.png")
plt.show()
```

```
Mission Path Map
Mean Velocity = 0.4048 \text{ km/h}
Min Velocity = 0.0634 \text{ km/h}
Max Velocity = 0.5906 \text{ km/h}
Min Mission Length = 317.24 \text{ h}
Total Gearbox and Motor efficiencies \eta = 0.70
Constant Electrical Power = 150 \text{ W}
RTS Mass = 1800 \text{ kg}
Wheel Diameter = 0.80 \text{ m}
Wheel Width = 0.35 \text{ m}
```

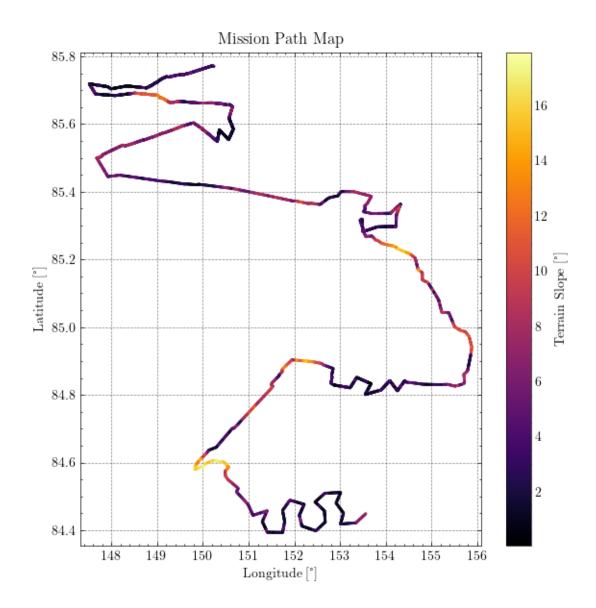


```
[]: from scipy.interpolate import griddata
  data = pd.read_csv('Pfad_komplett_Steigung_V2.csv')

distance = data['position']
  longitude = -data['lon']
  latitude = -data['lat']
  slope = data['TerrainSlope']

# Create a grid spherical
  #xi_sphere = np.linspace(min(longitude), max(longitude), len(longitude))
  #yi_sphere = np.linspace(min(latitude), max(latitude), len(latitude))
  #xi_sphere, yi_sphere = np.meshgrid(xi_sphere, yi_sphere)
```

```
#zi sphere = griddata((longitude, latitude), slope, (xi sphere, yi sphere), u
 ⇔method='nearest')
solution = func(slope, v rover constant, 1800, 0.8, 150*0.7, 0.35, 1.62, 0.03,
end_distance = distance.iloc[-1]
v_min = np.min(solution[0])
v_max = np.max(solution[0])
data['position_difference'] = data['position'].diff()
data['velocity'] = solution[0]
data['weighted velocity'] = data['position difference'] * data['velocity']
mean_velocity = data['weighted_velocity'].sum() / data['position_difference'].
 Silm()
slope mean = np.mean(data['TerrainSlope'])
end_distance = data['position'].iloc[-1]
total_time = end_distance / mean_velocity
fig, (ax1) = plt.subplots(1, 1, figsize=(6, 6))
scatter1 = ax1.scatter(longitude, latitude, c=slope, marker='.', s=5,_
 ⇔cmap='inferno')
fig.colorbar(scatter1, ax=ax1, label='Terrain Slope [°]')
#scatter1 = ax1.scatter(longitude, latitude, c=solution[0], linewidths=0.1,
 →marker='o', s=5, cmap='viridis')
#fig.colorbar(scatter1, ax=ax1, label='Speed [km/h]')
ax1.axis('auto')
margin = 0.2
ax1.set_xlim(min(longitude) - margin, max(longitude) + margin)
ax1.set_ylim(min(latitude) - margin/5, max(latitude) + margin/5)
ax1.set_xlabel('Longitude [°]')
ax1.set ylabel('Latitude [°]')
ax1.set_title('Mission Path Map')
plt.tight_layout()
plt.savefig("map_slope_sys_management.pdf")
plt.savefig("map_slope_sys_management.png")
plt.show()
```



```
[]: from scipy.interpolate import griddata
  data = pd.read_csv('Pfad_komplett_Steigung_V2.csv')

distance = data['position']
  longitude = -data['lon']
  latitude = -data['lat']
  slope = data['TerrainSlope']

# Create a grid spherical
  #xi_sphere = np.linspace(min(longitude), max(longitude), len(longitude))
  #yi_sphere = np.linspace(min(latitude), max(latitude), len(latitude))
  #xi_sphere, yi_sphere = np.meshgrid(xi_sphere, yi_sphere)
```

```
#zi sphere = griddata((longitude, latitude), slope, (xi sphere, yi sphere), u
 →method='nearest')
solution = func(slope, v rover constant, 1800, 0.8, 150*0.7, 0.35, 1.62, 0.03,
end_distance = distance.iloc[-1]
v_min = np.min(solution[0])
v_max = np.max(solution[0])
data['position_difference'] = data['position'].diff()
data['velocity'] = solution[0]
data['weighted_velocity'] = data['position_difference'] * data['velocity']
mean_velocity = data['weighted_velocity'].sum() / data['position_difference'].
 Silm()
slope mean = np.mean(data['TerrainSlope'])
end_distance = data['position'].iloc[-1]
total_time = end_distance / mean_velocity
fig, (ax1) = plt.subplots(1, 1, figsize=(6, 6))
scatter1 = ax1.scatter(longitude, latitude, c=slope, marker='.', s=5,_
 fig.colorbar(scatter1, ax=ax1, label='Terrain Slope [°]')
#scatter1 = ax1.scatter(longitude, latitude, c=solution[0], linewidths=0.1,
 →marker='o', s=5, cmap='viridis')
#fig.colorbar(scatter1, ax=ax1, label='Speed [km/h]')
ax1.axis('auto')
margin = 0.2
ax1.set_xlim(min(longitude) - margin, max(longitude) + margin)
ax1.set_ylim(min(latitude) - margin/5, max(latitude) + margin/5)
# Add markers for every k hours
k = 50 \# hours
tolerance = 0.1  # Adjust this value based on your desired tolerance
marker_added = False # Reset the marker flag at the start
for i in range(len(distance)):
    if i > 0:
        v_mean_up_to = np.cumsum(data['weighted_velocity'] /__

data['position_difference'])[i] / i

        time_passed = distance[i] / v_mean_up_to
        # Check if the current time_passed is within a new k-hour interval
        if abs((time_passed % k) / k - 1) < tolerance and not marker_added:
            label_text = '{}h'.format(round(time_passed / k) * k)
            #ax1.plot(longitude[i], latitude[i], 'o', color='white',
 →markersize=6, markeredgecolor='black')
```

