The background is a vibrant, stylized space scene. It features large, flowing nebulae in shades of red, purple, and blue. Scattered throughout are various celestial bodies: a large red planet with orange and yellow bands in the top right, a yellow and orange striped planet in the bottom left, and several smaller, dark, cratered planets. White stars of varying sizes are also visible.

# Exoplanets: Modeling the Transit Light Curve

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# Motivations



## Understand Light Curves

Modeling demonstrates what the typical light curve looks like, helping us understand the different components that make each transit unique.



## Finding the Radius

Using the information we gained from the model, we can quite accurately predict the radius and of the planet.



## Gathering More Data

Knowing the radius, we can make further predictions that will help us understand what the planet would be like, and if it would be habitable.

# The Steps to the Final Result

1. Making the General Model

2. Adjusting to Fit Real Data

3. Finding the Reduced Chi Squared

4. Repeat 1-3 with a Better Model



# 1. Making the General Model



# Making the General Model

Understand what makes up the general Model

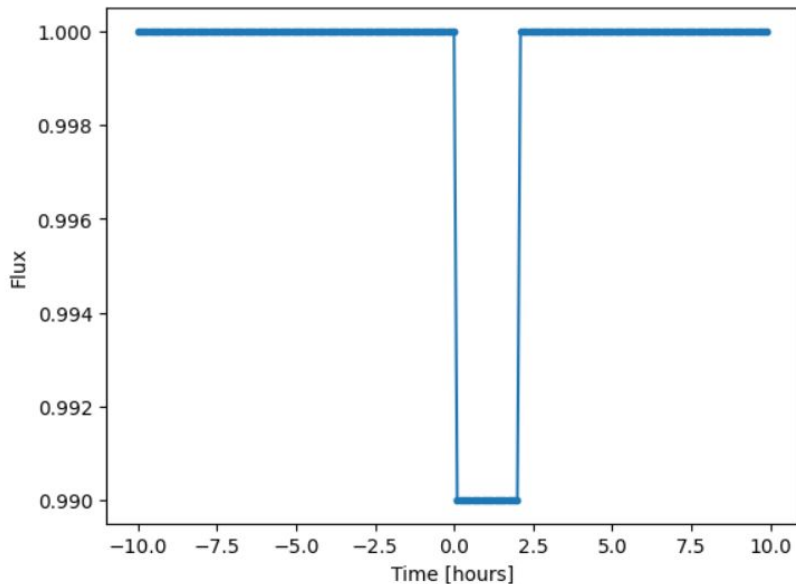
- $\tau$ : Changing this increases/decreases transit duration
- $T_0$ : Changing this shifts the transit left/right
- **Start Time**: Changing this changes when the transit will start
- **End Time**: Changing this changes when the transit will end
- **Delta**: Changing this changes the depth/flux of light reduction

# Making the General Model

Adding components to make the model function

- Assign index values to time array
- Using these values and the start/end time to determine where the flux/transit will start
- Graphing the components

```
delta = 0.01 #graphing what we just defined  
flux_arr[ind] = 1.0 - delta  
plt.plot(time_arr, flux_arr, linestyle = "--", marker = ".")  
plt.xlabel("Time [hours]")  
plt.ylabel("Flux")  
plt.show()
```



# Making the General Model

Turn the information into a single function to be used as a model

- ```
def generate_transit_lightcurve(time_arr, t_0, tau, delta):  
    flux_arr = np.zeros(np.shape(time_arr)) + 1.0  
    time_start = t_0 - tau / 2  
    time_end = t_0 + tau / 2  
    ind = np.where((time_arr >= time_start) & (time_arr <= time_end))  
    flux_arr[ind] = 1.0 - delta  
    return flux_arr
```
- Now we can use this model at any point as long as we define a specific **time array**,  $T_0$ ,  $\tau$ , and **Delta**



2.

# Adjusting to Fit Real Data

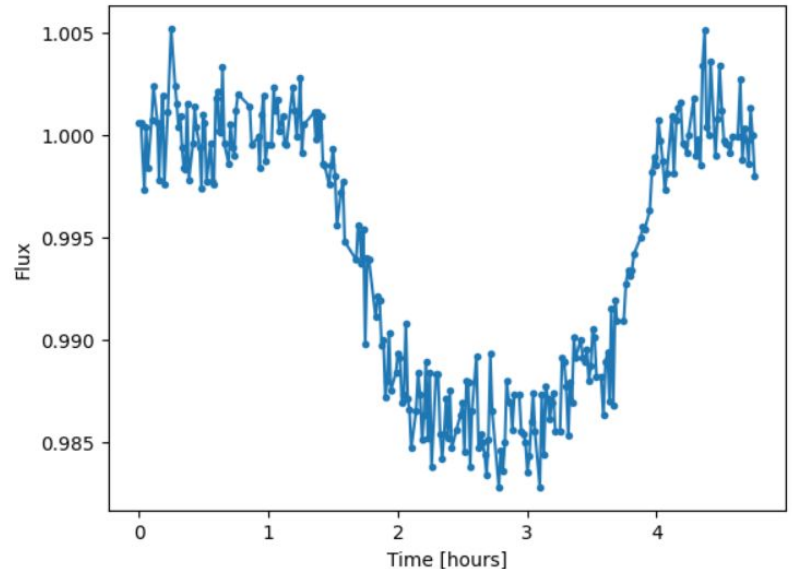




# Adjusting to Fit Real Data

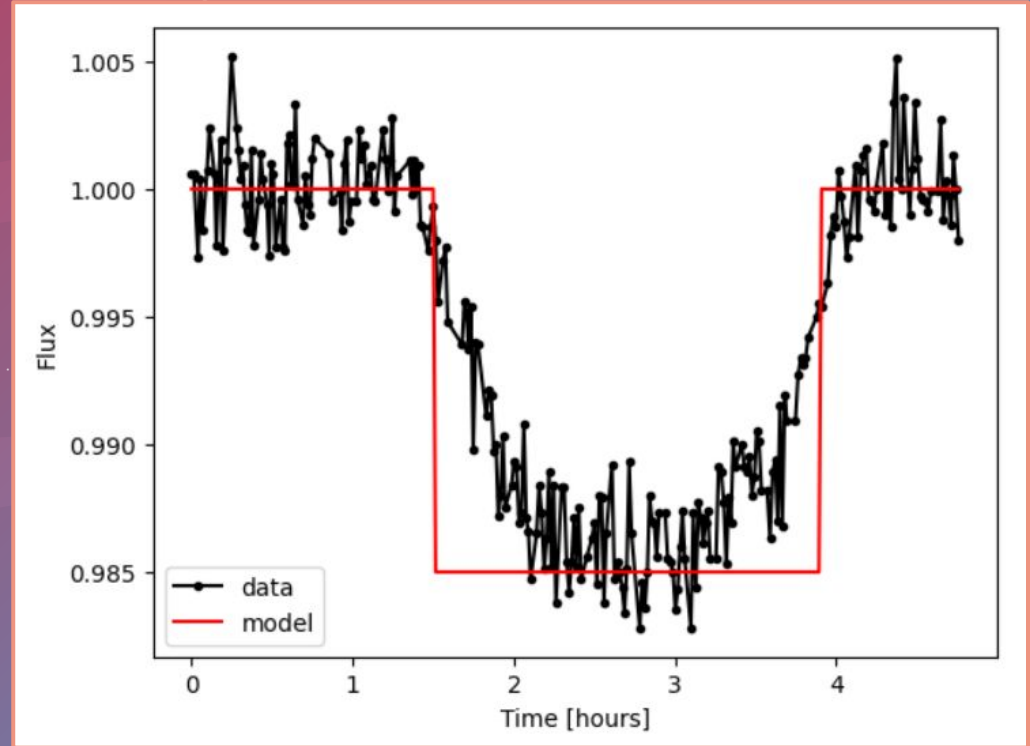
- Import data from the NASA Exoplanet Archive
  - HAT-P-1 b (Johnson et al. 2008 with 253 data points)
- Use `astropy.io.ascii` to read the data and make it a table
- Graph using values from the table

```
time_obs = dat["HJD"] #putting our time data in
time_obs = (time_obs - time_obs[0]) * 24.0 * u.hour #our time was in da
flux_obs = dat["Relative_Flux"] # putting our flux data in
plt.plot (time_obs, flux_obs, marker = ".")
plt.xlabel("Time [hours]")
plt.ylabel("Flux")
plt.show() #plotting the transit for HAT-P-1b
```



# Adjusting to Fit Real Data

- estimating parameters for the model
  - $\tau$ : 2.4 hours
  - $T_0$ : 2.7 hours
  - **Delta**: 0.015
- Overplot model





3.

# Finding the Reduced Chi Squared



# Finding the Reduced Chi Squared

## Calculating the Reduced Chi Squared

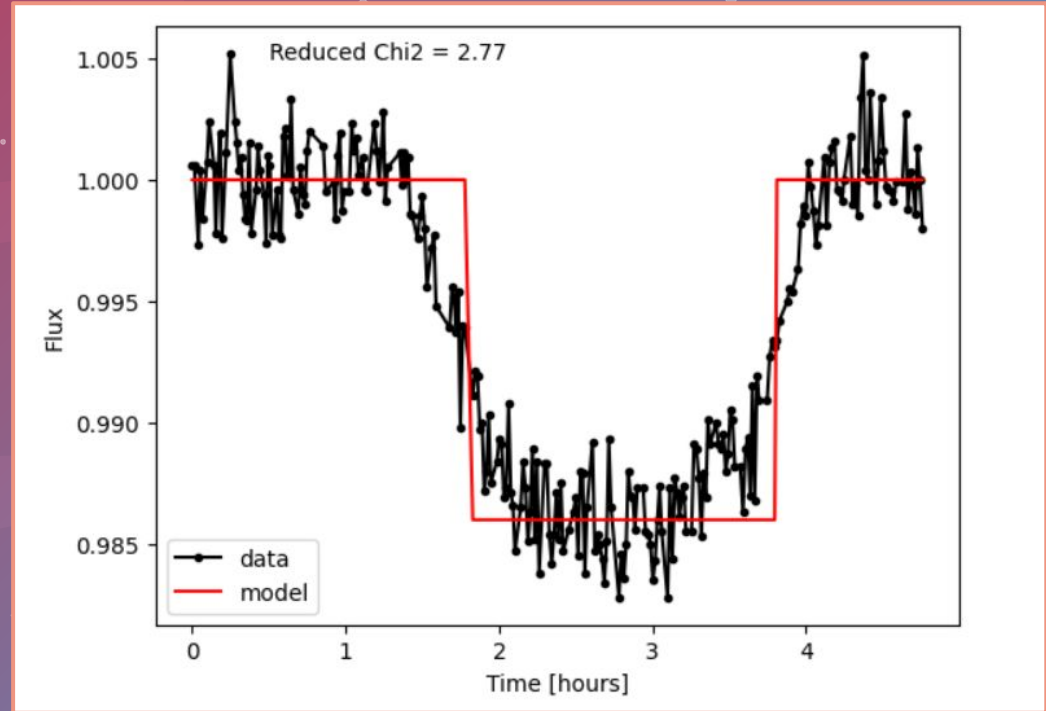
```
ind = np.where(time_obs < 1.5 * u.hour)
error = np.std(flux_obs[ind])

degree_of_freedom = len(time_obs) - 3
reduced_chi2 = np.sum(((flux_obs - flux_model) / error)**2) / degree_of_freedom
print(reduced_chi2)
```

```
5.4712288561723925
```

# Finding the Reduced Chi Squared

- Re-adjusting Values until we get as low a Reduced Chi Squared as possible
  - $\tau$ : 2.8 hours
  - $T_0$ : 2.0 hours
  - **Delta**: 0.014
- Re-graphing the overplotted model with new Reduced Chi Squared listed





4.

Repeat 1–3 with a  
Better Model



# Repeat Step 1

To more accurately measure the shape of the light curve, we will define a trapezoidal model that slopes in and out of the transit

- $\tau$ : Changing this increases/decreases transit duration
- $T_0$ : Changing this shifts the transit left/right
- **Start Time**: Changing this changes when the transit will start
- **End Time**: Changing this changes when the transit will end
- **Delta**: Changing this changes the depth/flux of light reduction
- **Ingress egress time**: Changing this changes the slope in/out of the transit

NEW!



# Repeat Step 1

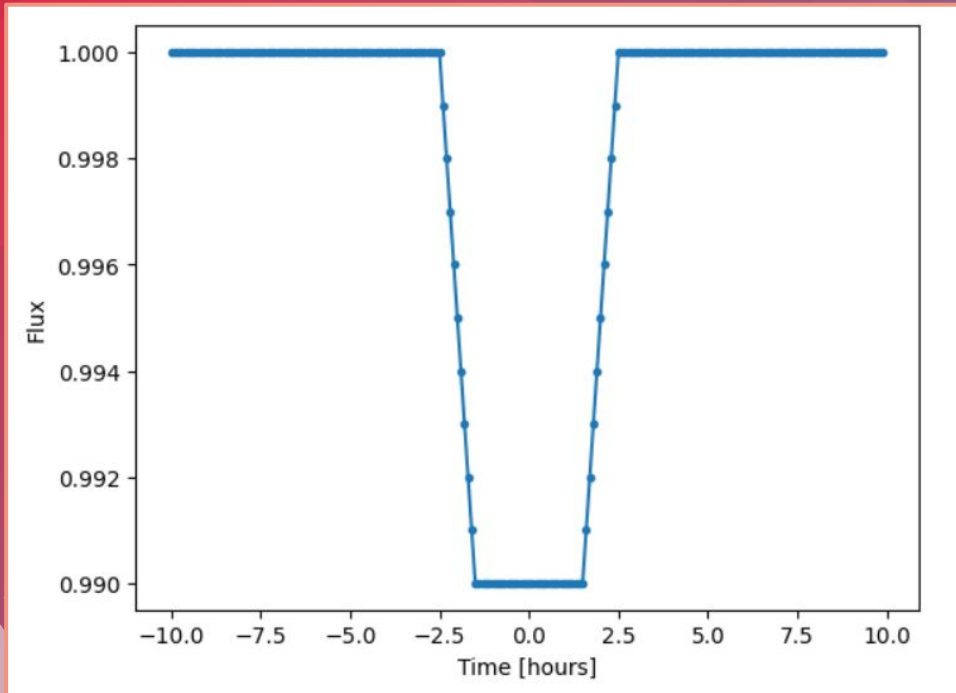
- Defining our new Model Function

```
def trapezoid_model(time_arr, t_0, tau, delta, ingress_egress_time):  
    flux_arr = np.zeros(np.shape(time_arr)) + 1.0  
    time_start = t_0 - tau / 2  
    time_end = t_0 + tau / 2  
  
    #Find the indices for the transit  
    transit_start = np.where(time_arr >= time_start)[0][0]  
    transit_end = np.where(time_arr <= time_end)[0][-1]  
  
    #changing the flux array for a trapezoidal shape (#making the lines down/up an adjustable V slope)  
    for i in range(transit_start, transit_end + 1):  
        if time_arr[i] < time_start + ingress_egress_time:  
            flux_arr[i] = 1.0 - delta * (time_arr[i] - (time_start)) / ingress_egress_time  
        elif time_arr[i] > time_end - ingress_egress_time:  
            flux_arr[i] = 1.0 - delta * (time_end - time_arr[i]) / ingress_egress_time  
        else:  
            flux_arr[i] = 1.0 - delta  
  
    return flux_arr
```



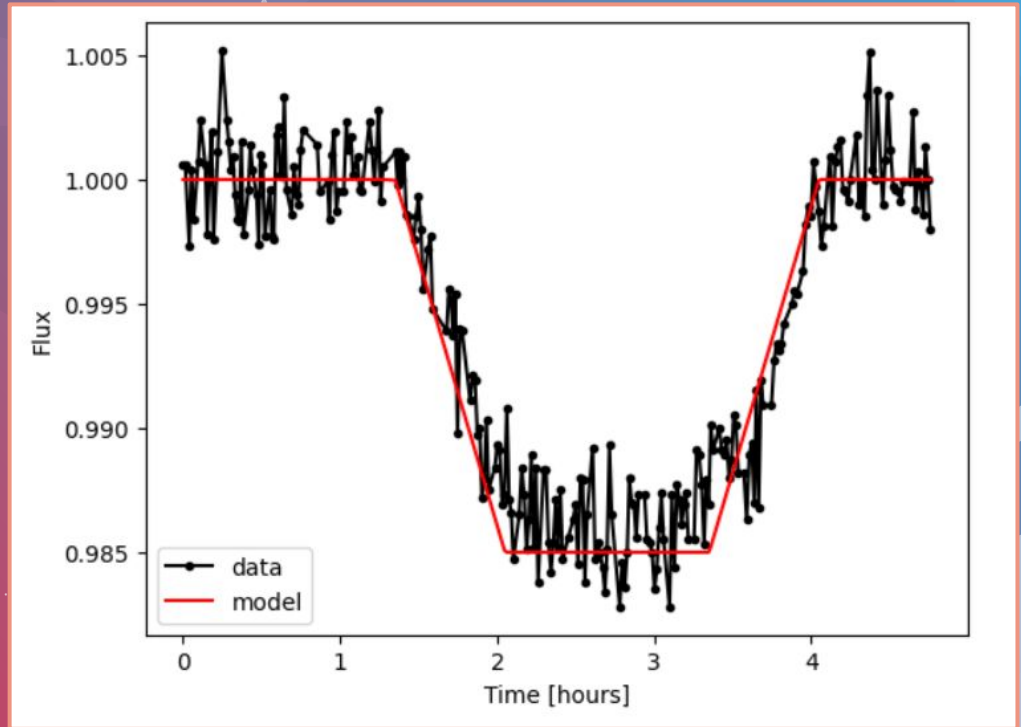
# Repeat Step 1

- Graphing New Model



# Repeat Step 2

- estimating parameters for the model
  - $\tau$ : 2.7 hours
  - $T_0$ : 2.7 hours
  - **Delta**: 0.015
  - **I/E time**: 0.7 hour
- Overplot model



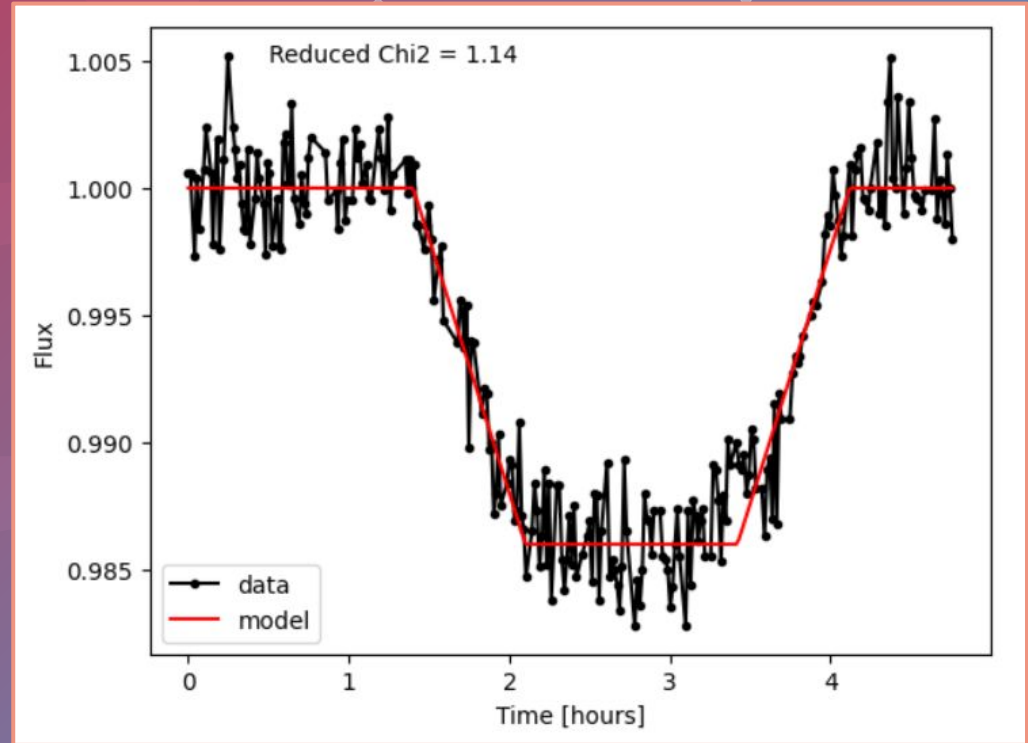
# Repeat Step 3

## Calculating the Reduced Chi Squared

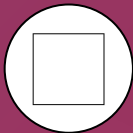
```
ind = np.where(time_obs < 1.5 * u.hour  
error = np.std(flux_obs[ind])  
  
degree_of_freedom = len(time_obs) - 3  
reduced_chi2 = np.sum(((flux_obs - flux_model2) / error)**2) / degree_of_freedom  
print(reduced_chi2)  
  
1.5226942614463883
```

# Repeat Step 3

- Re-adjusting Values until we get as low a Reduced Chi Squared as possible
  - $\tau$ : 2.76 hours
  - $T_0$ : 2.72 hours
  - **Delta**: 0.014
  - **I/E time**: 0.7 hour
- Re-graphing the overplotted model with new Reduced Chi Squared listed



# Conclusion



## Box Model

This model works, and gives a fairly accurate reduced chi squared of **2.77**



## Trapezoidal Model

This model works far better, giving an even more accurate reduced chi squared of **1.14**



Thank You!