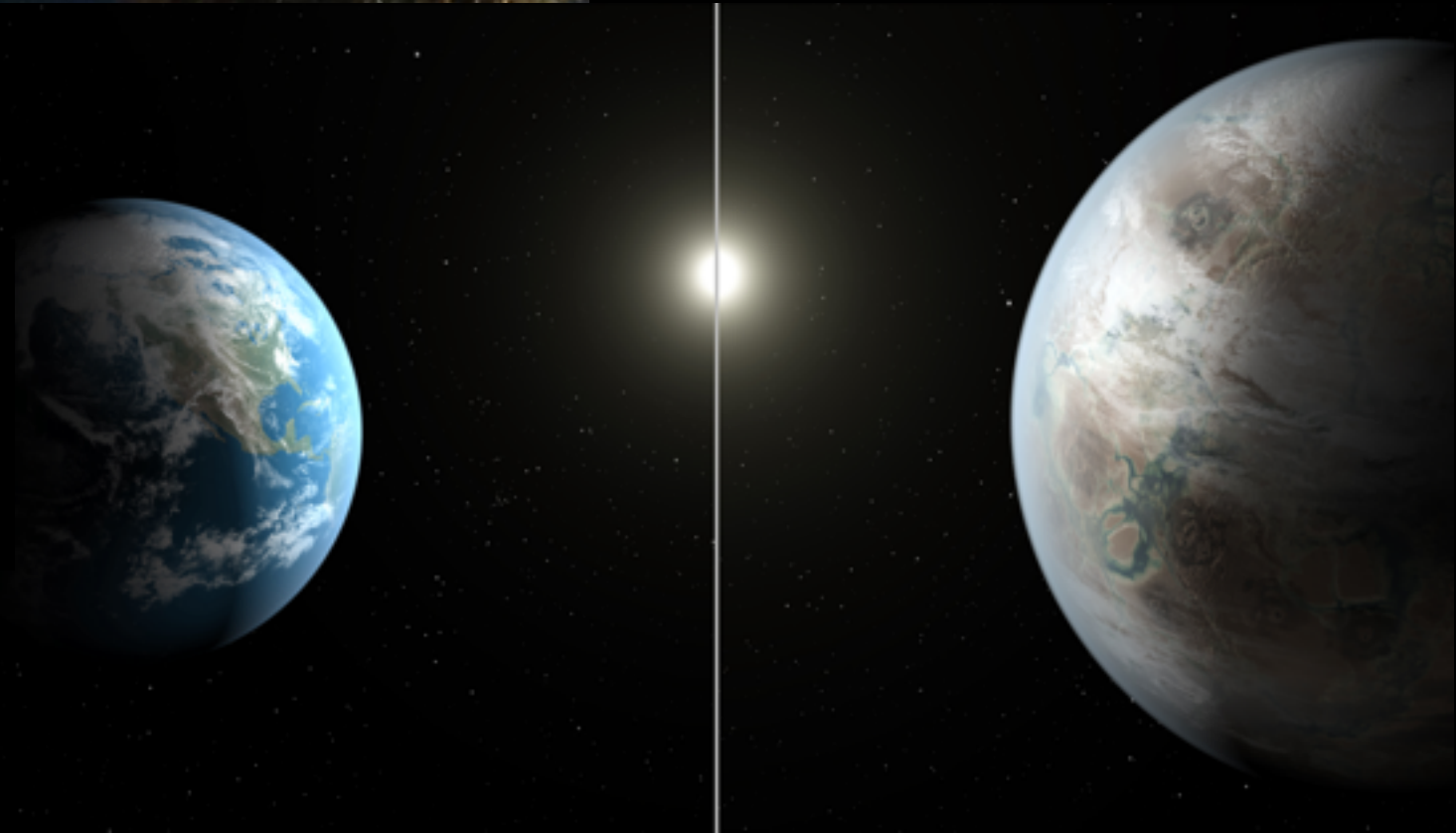
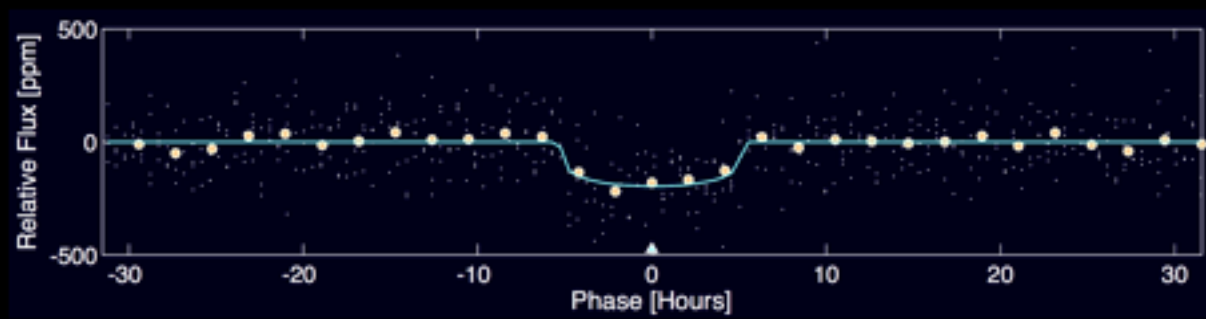


Earth 2.0?

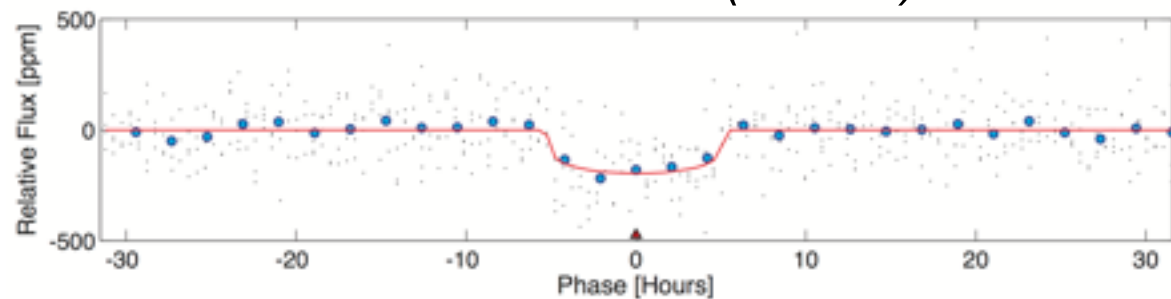


Quick bit of background...

In 2015, Kepler announced the discovery an Earth-like transiting planet

Discovery and Validation of Kepler-452b: A $1.6-R_{\oplus}$ Super Earth Exoplanet in the Habitable Zone of a G2 Star

Jenkins et al. (2015)

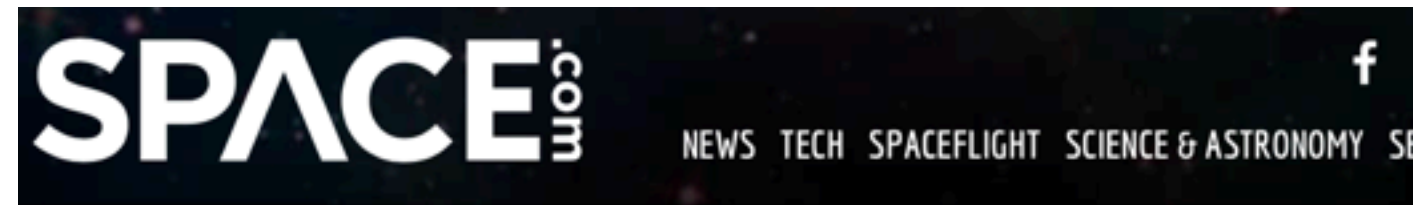


The New York Times

NASA Says Data Reveals an Earth-Like Planet, Kepler 452b

By DENNIS OVERBYE JULY 23, 2015

people got excited...



Space.com > Science & Astronomy

Kepler-452b: What It Would Be Like to Live On Earth's 'Cousin'

By Mike Wall, Space.com Senior Writer | July 24, 2015 06:00am ET

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MORE ▾



Artist's concept of the surface of the newfound exoplanet Kepler-452b, which is about 60 percent wider and five times more massive than Earth. This illustration imagines that a runaway greenhouse effect has begun to take hold on Kepler-452b, driving off much of the planet's surface water.

Credit: SETI Institute/Danielle Futselaar

Kepler-452b may be Earth's close cousin, but living on the newfound world would still be an alien [experience](#).

A group of pioneers magically transported to the surface of [Kepler-452b](#) — which is the closest thing to an "Earth twin" yet discovered, researchers announced yesterday (July 23) — would instantly realize they weren't on their home planet anymore. (And magic, or some sort of warp drive, must be invoked for such a journey, since

Kepler-452b lies 1,400 light-years away.)

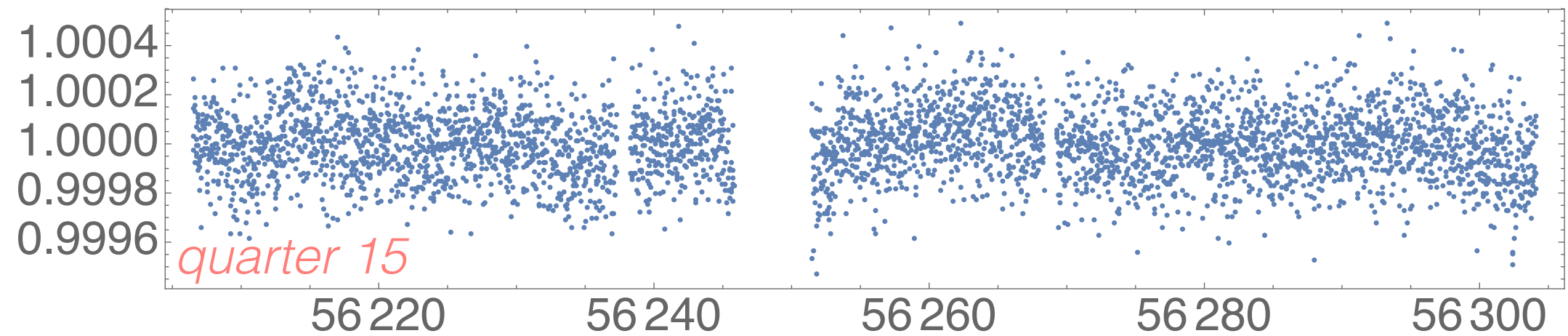
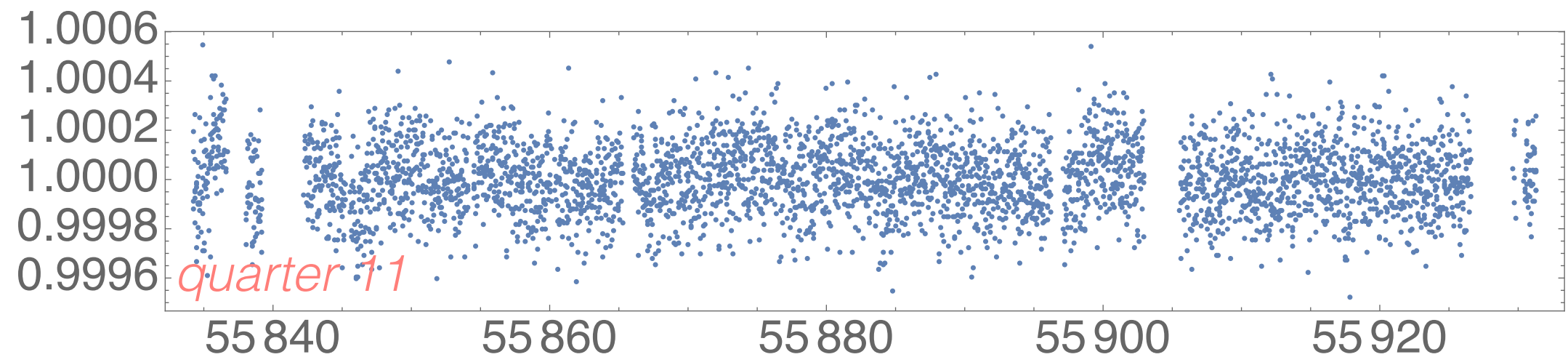
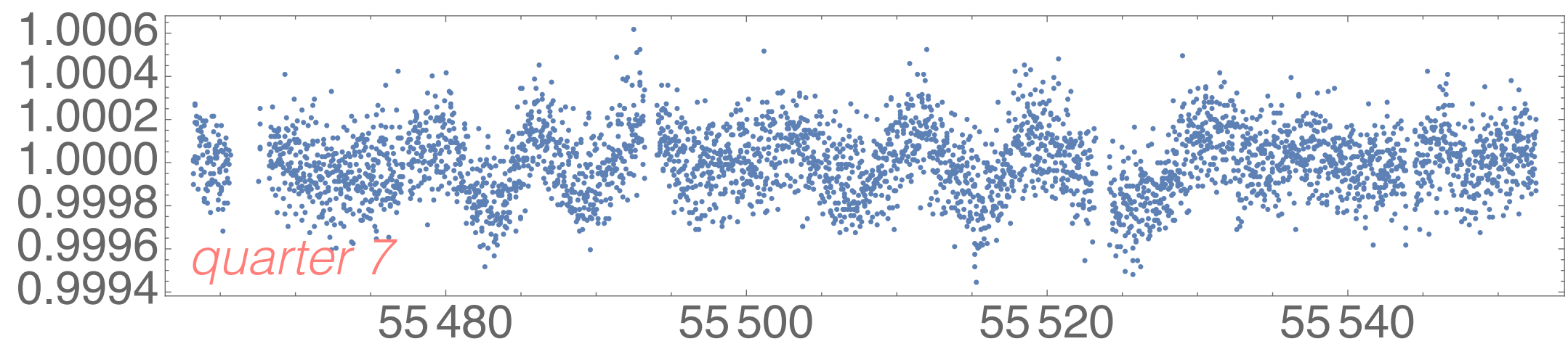
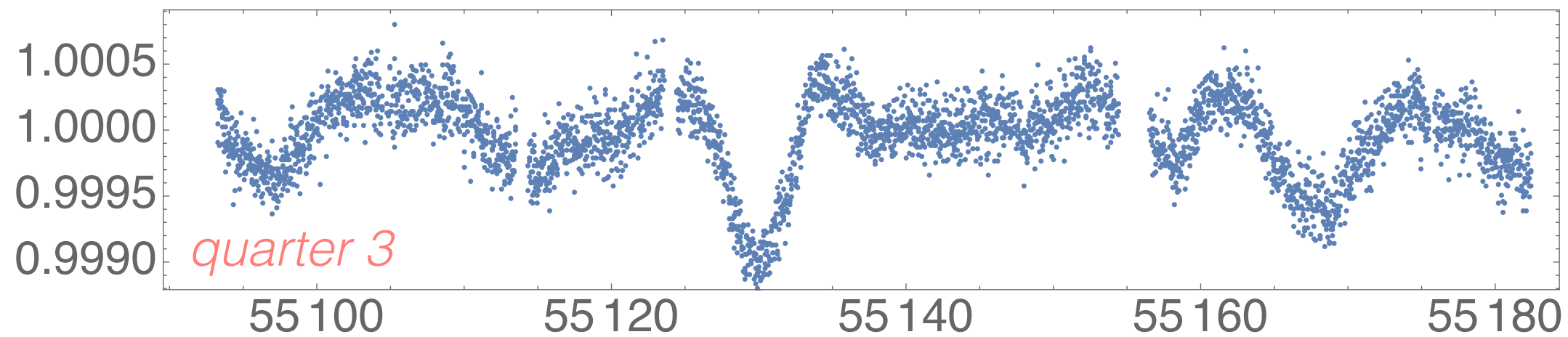
Let's test that for ourselves!

*science question: What is the probability
this is a rocky and HZ planet?*

Let's test that for ourselves!

1] import the raw data...

*(I have removed bad data flags, normalized by quarter median,
parsed the data for you and trimmed irrelevant quarters)*



2] remove outliers and apply detrending (as necessary)

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
import batman
```

```
In [2]: # Define some things we know
P_literature = 384.843
tau_literature = 54833 + 314.98
T_literature = 10.63/24.0      # literature reported transit duration
```

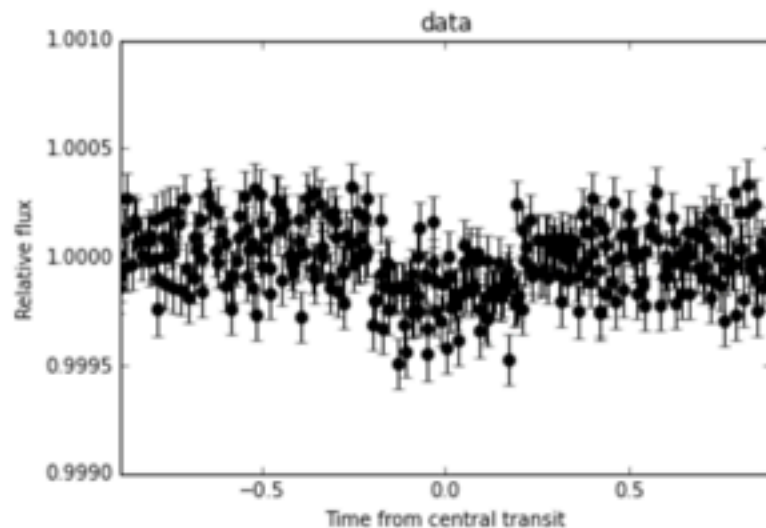
```
In [3]: # Define a set of test parameters
theta=np.empty([5])
theta[0] = 0.0128      # Rp/Rstar
theta[1] = 3.306       # log(rho* [kg/m3])
theta[2] = 0.69        # b impact parameter
theta[3] = P_literature # orbital period
theta[4] = tau_literature # transit mid-point
```

```
In [20]: # Define a set of test times
#t = np.linspace(theta[0]-2.0*T_literature, theta[0]+2.0*T_literature, 1000)

data = np.genfromtxt("lightcurve.dat", dtype=None)
tobs=data[:,0]      # times of observations
fobs=data[:,1]      # relative fluxes
sigfobs=data[:,2]   # errors on fluxes
tobsfolded=np.empty([len(tobs)])
for i in range(len(tobs)):
    tobsfolded[i]=tobs[i]-tau_literature-P_literature*np.round((tobs[i]-tau_literature)/P_literature)

%matplotlib inline

plt.figure()
plt.errorbar(tobsfolded, fobs, yerr=sigfobs,fmt='ko')
plt.title("data")
plt.xlabel('Time from central transit')
plt.ylabel('Relative flux')
plt.xlim(-2*T_literature,+2*T_literature)
plt.ylim(0.999,1.001)
plt.show()
```



3] I'll give you a starter notebook to plot the data...

(no data cleaning shown here, that's raw data!)

```

In [5]: # Set a definition for the model
def model(theta, times):
    Grv = 6.67408e-11
    impact = theta[2]
    logrho = theta[1]
    rhostar = np.power(10.0, logrho)
    aR = np.power( ( Grv*(theta[3]*86400.0)**2*rhostar )/( 3.0*np.pi ), 0.33333333 )
    incdeg = (180.0/np.pi)*np.arccos(impact/aR)
    # Initialize the Batman code
    params = batman.TransitParams()
    # Set the model parameters
    params.rp = theta[0]          #R_planet/R_star
    params.a = aR                 #a/R_star
    params.inc = incdeg           #orbital inclination (degrees)
    params.per = theta[3]         #orbital period (days)
    params.t0 = theta[4]          #transit midpoint time (days)
    # These parameters are just kept fixed
    params.ecc = 0.0              #eccentricity
    params.w = 90.                #longitude of periastron (degrees)
    params.limb_dark = "nonlinear" #limb darkening model--other choices include quadratic, etc.
    # Coefficients for a nonlinear limb darkening law.
    params.u = [0.1570656, 0.9361135, -0.4313317, 0.0572520]
    # Call the Batman code
    m = batman.TransitModel(params, times, supersample_factor=30, exp_time=0.020434)
    flux = m.light_curve(params) #calculate the flux
    return flux

```

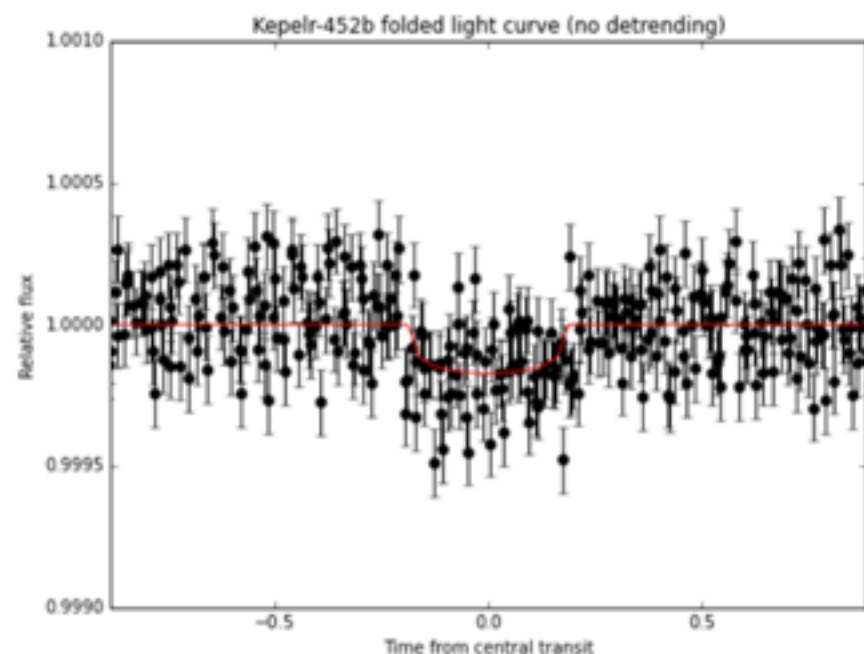
```

In [19]: %matplotlib inline

theta[4]=0.0 # since I have folded the lightcurve, purely for the sake of visualization,
            # i am going to set tau=0 to give a nice plot
# since i'm going to do a continuous line for my model, i need to sort the times first
fmod = model(theta, np.sort(tobsfolded))

fig = plt.figure(figsize=(8,6))
plt.errorbar(tobsfolded, fobs, yerr=sigfobs, fmt='ko')
plt.plot(np.sort(tobsfolded), fmod, c='r')
plt.title('Kepler-452b folded light curve (no detrending)')
plt.xlabel('Time from central transit')
plt.ylabel('Relative flux')
plt.xlim(-2*T_literature, +2*T_literature)
plt.ylim(0.999, 1.001)
plt.show()

```



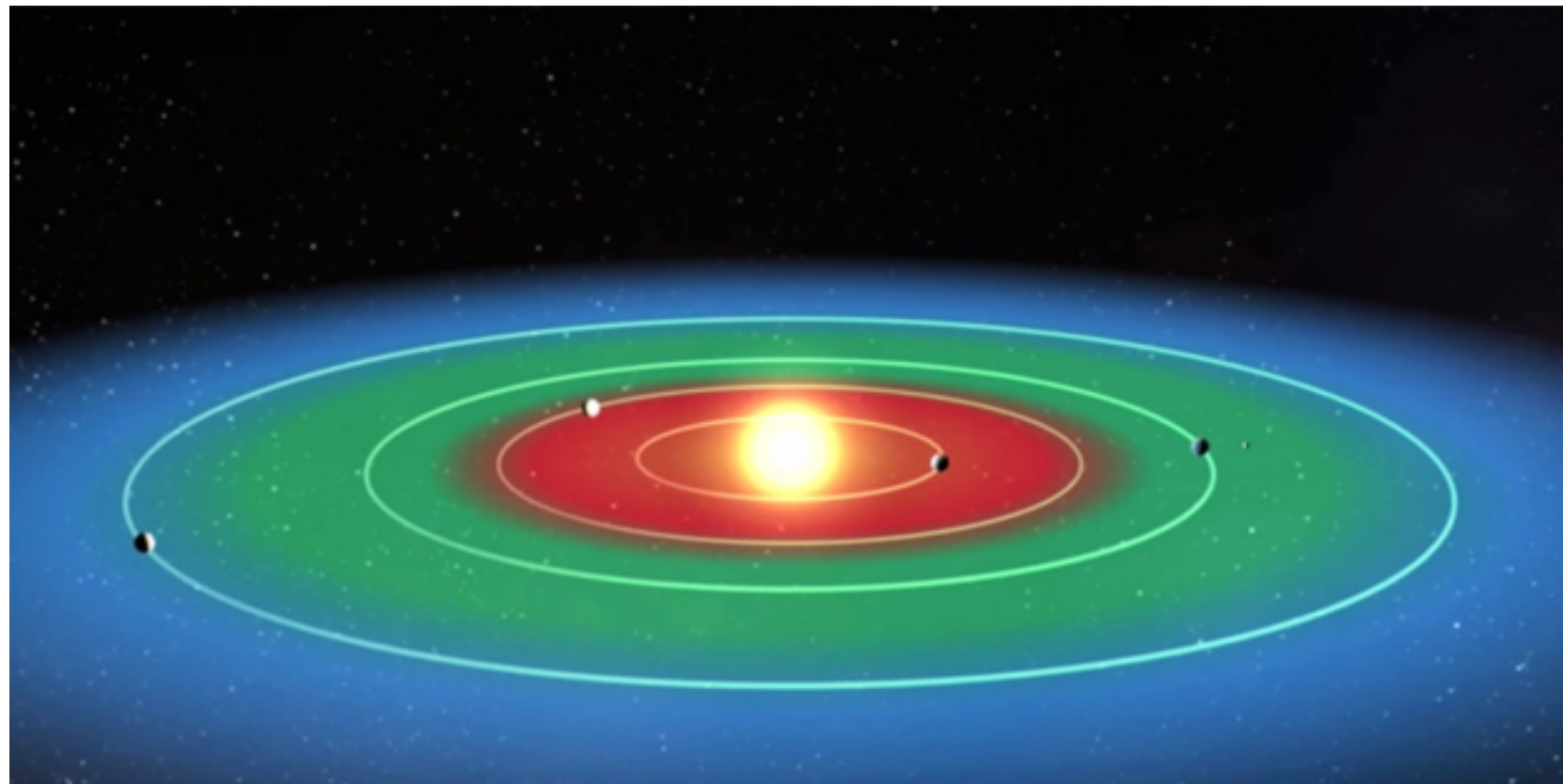
3] ...and how to generate a transit model for some choice of R_p/R^* , b (and $\log\rho^*$)

4] Fit the parameters R_p/R^ , b (and ρ^*) with an MCMC*

5] I'll give you a set of posterior sample for T_{eff} , R^* and $\log \rho^*$

...use it to calculate the temperature of the planet (see Wikipedia page on Stefan-Boltzmann law for eqn) and the radius of the planet

6] what fraction of your samples (i.e. what is the probability) fall within the HZ ($207.5K < T_p < 320.4K$) and fall below rocky-gas divide (1.23 Earth radii)?



Suggested path forward...

- 1] Import the data! Get familiar with it, plot it, make sense of it
- 2] Remove outliers and apply detrending
- 3] Use starter notebook to plot a guess model through the cleaned data
- 4] Run an MCMC fitting for b and R_p/R^* (+ $\log \rho^*$ for extra credit)
- 5] Use posterior samples of star's T_{eff} and R^* to calculate joint posterior for R_p and $T_p \Rightarrow$ what fraction of trials are HZ and rocky?

6] Can you go further and take your $\log \rho^*$ posterior and compare to independent measurement to e_{min} via...

$$e_{\text{min}} = \frac{|1 - (\rho_{\text{obs}}/\rho_{\text{tru}})^{2/3}|}{1 + (\rho_{\text{obs}}/\rho_{\text{tru}})^{2/3}}$$

- 7] Can you go further and use [github/forecaster](#) to predict probability planet is rocky using mass-radius models?
- 8] Prepare your summary slides and talk!

Suggested path forward...

1] Import the data! Get familiar with it, plot it, make sense of it

2] Remove outliers and apply detrending

3] Use starter notebook to plot a guess model through the cleaned data

4] Run an MCMC fitting for b and R_p/R^* (+ $\log\rho^*$ for extra credit)

make sure you quote your final R_p and T_p !

5] Use posterior samples of star's T_{eff} and R^* to **calculate joint posterior for R_p and T_p** \Rightarrow **what fraction of trials are HZ and rocky?** covariance \rightarrow go through samples

6] Can you go further and take your $\log\rho^*$ posterior and compare to independent measurement to e_{min} via...

$$e_{\text{min}} = \frac{|1 - (\rho_{\text{obs}}/\rho_{\text{tru}})^{2/3}|}{1 + (\rho_{\text{obs}}/\rho_{\text{tru}})^{2/3}}$$

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