

Calculating Hubble Constant

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Motivations

- Understanding the origin and eventual fate of the entire universe
- Understanding the age of the universe and seeing how the expansion model we've calculated fits the data



Methods

- Utilizing python packages to plot and fit models to data
- Using the equation for the Hubble Constant and data from supernovae with known brightnesses

```

import numpy as np
import matplotlib.pyplot as plt
import astropy.io.ascii
import astropy.units as u
import astropy.constants as ac


from google.colab import files
uploaded = files.upload()
#importing necessary packages and uploading data file

```

```

data = astropy.io.ascii.read("Tonry_2003.vot") # formatting imported data into python
print(data) # printing data

```

 /usr/local/lib/python3.10/dist-packages/astropy/io/ascii/html.py:91: XMLParsedAsHTMLWarning

```

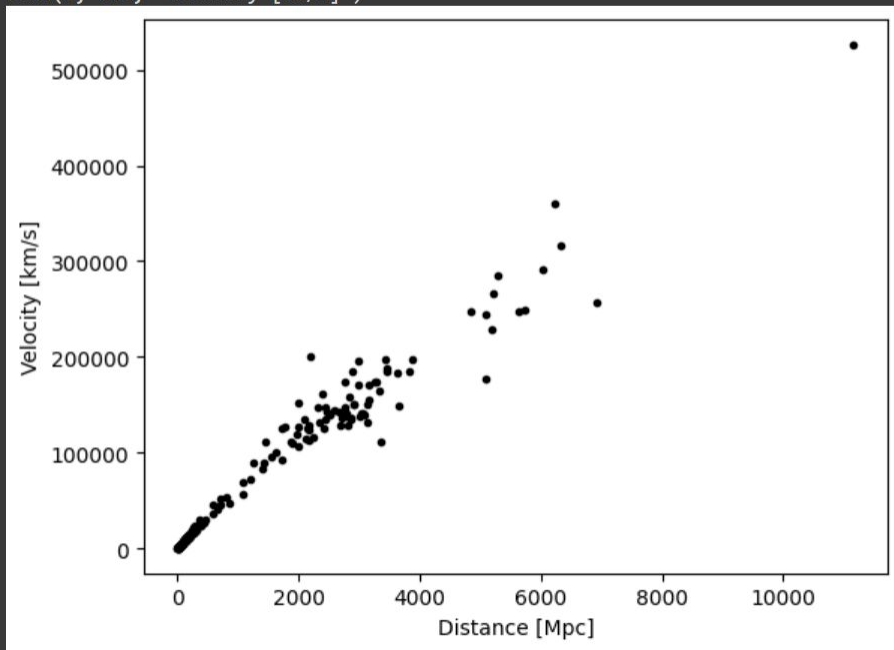
soup = BeautifulSoup("\n".join(lines))
col1    col2    col3    col4    col5    col6    col7    col8    col9    col10   col11
-----
SN1972E 314.84   30.08  0.0023  N5253   16  2.839  2.399  0.033  204.9697 -31.6692
SN1980N 240.161 -56.689 0.0056  N1316    9  3.225  3.14  0.043  50.6753 -37.2074
SN1981B 292.97   64.743 0.0072  N4536    2  3.334  3.077  0.041  188.6233  2.1995
SN1981D 240.161 -56.689 0.0056  N1316    9  3.225  3.044  0.055  50.6753 -37.2074
SN1986G 309.543  19.401 0.0027  N5128   26  2.908  2.44  0.035  201.4028 -43.0316
SN1988U   8.737 -81.227   0.31  Anon    24  4.968  5.096  0.072   3.5756 -30.4164
SN1989B 241.991  64.403 0.0036  N3627   37  3.033  2.844   0.03  170.0578  12.9718
SN1990N 294.369  75.987 0.0044  N4639   21   3.12  3.204  0.035  190.7367  13.2566
SN1990O 37.654   28.36 0.0307  M+034403 10  3.964  3.977  0.025  258.8999  16.3241
SN1990T 341.503 -31.526   0.04  P63925  10  4.079  4.101  0.042  299.7601 -56.2583
...      ...      ...      ...      ...      ...      ...      ...      ...      ...      ...
SN2000cx 136.506 -52.482   0.007  N0524 11:19 3.322  3.32  0.023  21.1926   9.509
SN2000dk 126.834 -30.344 0.0164  N0382  11  3.692  3.677  0.023  16.8483  32.4068
SN2000dz  84.367  -56.39   0.5  Anon   13  5.176  5.352  0.053  352.6728  0.3119
SN2000ea 167.211 -61.402   0.42  Anon   13   5.1  5.091  0.053  32.4751 -5.4719
SN2000ec 166.295 -60.176   0.47  Anon   13  5.149  5.34  0.043  32.8836 -4.2321
SN2000ee 166.045 -53.429   0.47  Anon   13  5.149  5.343  0.041  36.8945  1.1971
SN2000eg 167.097 -53.108   0.54  Anon   13  5.209  5.236  0.049  37.5883  1.0639
SN2000eh 188.297 -31.651   0.49  Anon   13  5.167  5.224  0.042  63.7604  4.3888
SN2000fa 194.167  15.479 0.0218  U03770 11  3.815  3.837  0.026  108.8752  23.4287
SN2001V 218.929 77.733 0.0162  N3987  22  3.686  3.662  0.023  179.3542  25.2024
Length = 230 rows

```

```
[ ] dist = ((10**data["col8"] * u.km / u.s) / (72 * ((u.km * u.s**-1)/ u.mpc)))
"""
"col8" has the values of Final median distance times H_0_ in log(km/s).
Raising 10 to the power of the value in column 8 gives the receding speed of the supernova in km/s.
Then dividing that by the assumed hubble constant gives the distance of the supernova in mpc
"""
vel = 10**data["col7"] * u.km / u.s # "col7" has the values of the receding velocity in units of log(km/s). Raise 10 to the value for the velocity
```

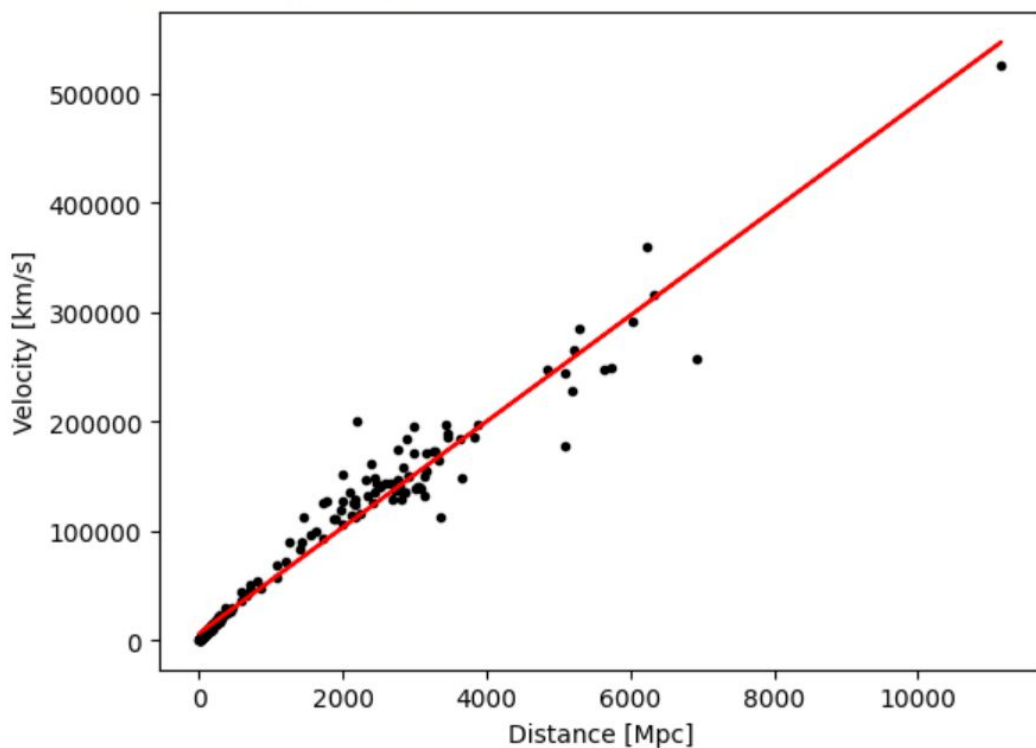
```
plt.plot(dist, vel, marker=".", color="black", linestyle="none") # plotting and formatting graph
plt.xlabel("Distance [Mpc]")
plt.ylabel("Velocity [km/s]")
```

```
Text(0, 0.5, 'Velocity [km/s]')
```





```
lin_fit = np.polyfit(dist.to(u.mpc).value, vel.to(u.km / u.s).value, 1) # calculating linear fit coefficients  
vel_lin = lin_fit[0] * dist.to(u.mpc).value + lin_fit[1] # creating calculated velocity curve  
plt.plot(dist, vel, marker=".", color="black", linestyle="none")  
plt.plot(dist, vel_lin, color="red")  
plt.xlabel("Distance [Mpc]")  
plt.ylabel("Velocity [km/s]")
```





```
h = lin_fit[0] * u.km / u.s / u.Mpc # slope of graph = hubble constant  
print(h)
```

```
[ ] age = 1/h # age of universe = inverse of hubble constant  
    print(age.to(u.Gyr))
```

20.174816604618965 Gyr



Conclusions

- Our model used a linear fit and included ALL data points
- This led to multiple inaccuracies in our model
 - The linear fit did not cross the origin
 - The large amounts of measurement error on farther away data points