



# Determining the Hubble Constant from Observations of Distance Modulus and Redshift for Type Ia Supernovae

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# Background & Purpose

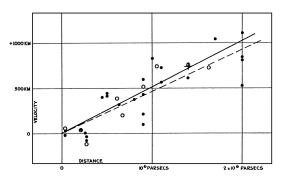
• Universe is expanding, modeled by equation

$$\mathbf{v} = \mathbf{H}_0 \mathbf{d}$$

where v = recessional velocity of galaxy

 $H_0$  = Hubble's constant

d = distance of galaxy



Plot of recessional velocity v vs distance d for several galaxies. The slope of the best-fit line is the Hubble constant  $H_0$ .

- Disagreement on  $H_0$ : current accepted value is  $H_0 = 72 \pm 2 \text{ km s}^{-1} \text{ Mpc}^{-1}$ .
- Prior research: Scolnic 2018<sup>1</sup> Pantheon Dataset
  - Observed distance modulus and redshift for 1048 Type Ia supernovae
  - $\circ$  distance modulus  $\sim$  distance d, redshift  $\sim$  velocity v.

<u>Purpose of experiment:</u> To determine a precise value of  $H_0$  by fitting a model to the Pantheon dataset and using the model parameters to calculate  $H_0$ .

### Procedure

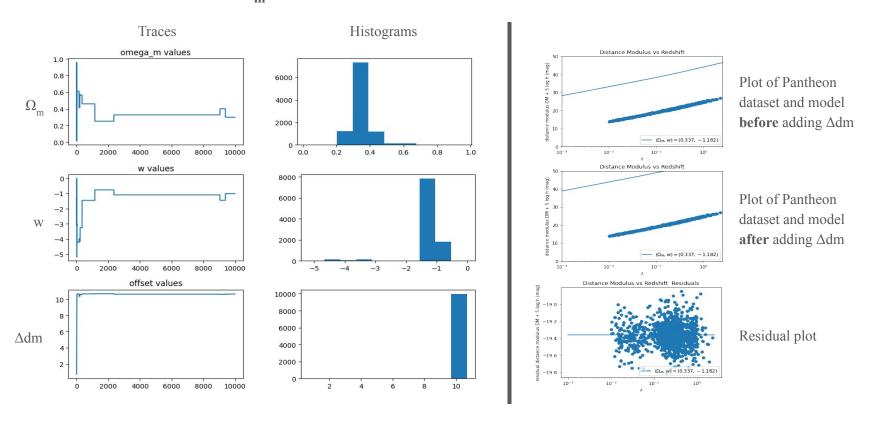
- Definitions of parameters:
  - $\circ$   $\Omega_{\rm m}$  = mass density of universe
  - $\circ$  w = p/ $\rho$  = pressure / energy density = equation of state of universe
  - $\circ$   $\Delta$ dm = distance modulus offset between the model and data
- 1. Plot distance modulus vs redshift values for Pantheon dataset of 1048 points.
- 2. Initialize Markov Chain with  $(\Omega_{m0}, w_0, \Delta dm_0) = (0.0, -1.0, 0.0)$ .
- 3. Iterate through Markov Chain. Each Markov-Chain Monte Carlo (MCMC) iteration:
  - a. Draw the latest value of chain  $\theta$ , as original parameters. Draw proposal  $\theta'$  from normal distribution centered on  $\theta$ .
  - b. Calculate log likelihood function of original parameters  $lnf(\Omega_m, w, \Delta dm)$  and proposed parameters  $lnf(\Omega_m', w', \Delta dm')$ .

$$\ln f(\Omega_m, w, \Delta_{dm}) = \sum_{i=1}^{N} \left[ \ln \left( \frac{1}{\sigma_{dm_i} \sqrt{2\pi}} \right) - \frac{1}{2} \frac{(dm_{obs} - dm_{predicted}(\Omega_m, w, \Delta_{dm}))^2}{(\sigma_{dm_i})^2} \right]$$

- c. Calculate ln(r) where 0 < r < 1.
- d. If  $\frac{\ln f(\Omega_m', w', \Delta dm') \ln f(\Omega_m, w, \Delta dm)}{\ln (r)} > \ln (r),$  then add proposed parameters  $\theta' = (\Omega_m', w', \Delta dm')$  to chain. Otherwise add current parameters  $\theta = (\Omega_m, w, \Delta dm)$  again.
- 4. After desired number of iterations, for each parameter:
  - $\circ$  mean  $\rightarrow$  experimental value
  - $\circ$  standard deviation  $\rightarrow$  uncertainty in experimental value

### Results

• Experimental values:  $\Omega_m = 0.337 \pm 0.066$ ,  $w = -1.182 \pm 0.521$ ,  $\Delta dm = 10.634 \pm 0.261$ 



## Discussion & Conclusion

	Scolnic (2018)	Shi and Farr	z-score	p-value	Significant? (α=0.05)
$\Omega_{\mathrm{m}}$	$0.307 \pm 0.012$	$0.337 \pm 0.066$	0.447	0.655	no
W	$-1.026 \pm 0.041$	$-1.182 \pm 0.521$	-0.299	0.765	no

- Our  $\Omega_m$  and w values are not significantly different from Scolnic (2018)
- Adding  $\Delta$ dm decreased quality of fit  $\rightarrow$  sampling process may be faulty!
- Unable to calculate  $H_0$  from parameters  $\rightarrow$  results are inconclusive

### Future goals:

- Minimize systematic error in model fit  $\rightarrow$  improve precision and accuracy for  $\Omega_{\rm m}$ , w,  $\Delta_{\rm dm}$
- Determine the value of the Hubble constant  $H_0$  based on  $\Omega_m$ , w,  $\Delta_{dm}$

# References & Acknowledgments

- Papers consulted
  - Scolnic, D.M. The Astrophysical Journal, 859:101 (2018).
  - Kirshner, Robert. PNAS, 101:8-13 (2004).
  - Hogg, David W. "Data Analysis Recipes: Using Markov Chain Monte Carlo," (2017).
  - Hogg, David W. "Distance Measures in Cosmology," (2000).
  - Hogg, David W. "Fitting a Model to Data," (2010).
- MCMC program written in Python and its libraries SciPy, NumPy, Matplotlib, and Jupyter Notebook, courtesy of Python Software Foundation <a href="https://www.python.org">www.python.org</a>
- Log likelihood function uses cosmological distance functions from Hogg (2000)
- Supported with a grant from the PSEG Explorations in STEM 2020 program