# Physics 307L 

Spring 2021

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## Reminders

- Schedule of assignment due dates through the end of the semester is now posted on wiki, and in Teams
- https://ghz.unm.edu/juniorlab/index.php?title=Schedule_Spring_2021\#Cour se_Schedule
- Schedule for Talk 2 is posted, first talks April 19, 26
- To get full credit, you need to submit your slides at least 1 hour ahead of time (submit through assignment in Teams)
- Try to wear lab-safety-appropriate long pants and closed-toed shoes, especially if you are working on an experiment with clear hazards (high voltage, lead bricks)
- Send an email and stay home if you aren't feeling well


## Error Analysis

- Some steps in error analysis so far
- Estimating uncertainties from equipment
- Repeating measurements to estimate uncertainty
- Propagating uncertainties
- Plotting and correlating data
- Choosing how to combine separate measurements, and possibly rejecting/cleaning data
- Least-squares fitting with errors in both dimensions
- There are many techniques for fitting and analyzing your experimental data to understand its statistics and sources of random/statistical error


## Chauvenet's Criterion

- If you make $N$ measurements of a single quantity $x$, Chauvenet's criterion gives a simple test for deciding whether to reject a 'suspect value'
- $t_{\text {sus }}=\frac{\left|x_{\text {sus }}-\bar{x}\right|}{\sigma_{x}}$
- $n=N \times \operatorname{Prob}\left(\right.$ outside $\left.t_{\text {sus }} \sigma\right)$
- Use Appendix A to look up values
- If $\mathrm{n}<0.5$, then it is reasonable to reject $\mathrm{x}_{\text {sus }}$


The slope $=0.9974273274123038$, with uncertainty 0.0012558490265354887 The intercept $=0.9282554690792497$, with uncertainty 0.6452914621011211

## Speed of Light Data - Our Class



## Speed of Light Data - Our Class Scrubbed of Inconsistent Data



# Speed of Light Data - Our Class Best fit to data from 5 students 



```
The slope = 3.4470692242258648e-09, with uncertainty 1.741025266191759e-11
The intercept = 3.111703798698238e-09, with uncertainty 1.298000077865901e-11
The speed of light = 290101513.76480633
```

Can we reject the hypothesis that the speed of light is $299,700,000 \mathrm{~m} / \mathrm{s}$ in our lab?

## How do we estimate systematic errors?

- We have identified some straightforward examples, which are closely related to how we estimate uncertainty from equipment
- Systematic error in approach to measure distance
- Front vs back vs middle of a line or edge
- Especially relevant for us when combining measurements from different students
- Systematic error from backlash in a knob
- Systematic error from calibration of oscilloscope
- Other examples?

How do we estimate systematic errors, in experiments with unknown results?

- Experimental design can sometimes lend itself to measuring a 'null' result (or at least a consistent result) which helps quantify systematic errors
- Example (Balmer Series): We could carefully compare line position with knob going one direction, to line position going the other direction. Since we expect it to be the same, we can compare to understand the magnitude of the systematic error coming from backlash.
- You could either take data in a consistent direction to avoid backlash, or quantify its effect on your data (for example, if it is negligible compared to another unavoidable error)

How do we estimate systematic errors, in experiments with unknown results?

- Experimental design can sometimes lend itself to measuring a 'null' result (or at least a consistent result) which helps quantify systematic errors
- Example (Speed of Light): We cannot claim that we are measuring the speed of light with our setup, since there are systematic errors beyond our capability of measuring.
- Need an absolute reference for time to calibrate
- We could still use the setup to accurately measure the index of refraction of a material. Why?

How do we estimate systematic errors, in experiments with unknown results?

- Experimental design can sometimes lend itself to measuring a 'null' result (or at least a consistent result) which helps quantify systematic errors
- Example (Poisson Statistics): We measured the background rate without a radioactive source to quantify the expected level of signal from other sources (cosmic rays, other radioactive elements in surroundings, background noise in detector)
- This background follows nice Poisson statistics, but can be considered a systematic effect
- Can't tell exactly where each 'hit' came from, can only quantify expected level of background hits

How do we estimate systematic errors, in experiments with unknown results?

- What if we can't do direct measurements?


## Simulating experiments

- Often in real experiments, analytical calculations, estimates, and experimental data are not sufficient to fully understand the magnitude of sources of error, and how they can interact
- Systematic errors also typically have random fluctuations, and can fluctuate up and add together all at once making a big outlier in your data
- Very important to understand sources of uncertainty for reporting new results


## Simulating experiments

- Many decisions to be made and justified on how to set up a full simulation
- What 'shortcuts' are ok and why?
- What effects are important to model accurately?
- What framework to use to allow for future capacity?
- What resources are available?
- What are the limitations of the simulation?
- How does simulation compare to actual data?


## Poisson Statistics Modeling

- We set up a 'toy model' for understanding the experiment and what would be needed to simulate it accurately.
- This can be useful for designing an experiment and understanding its limitations



## Poisson Statistics Modeling

- https://trinket.io/glowscript/8060842cfb


## Poisson Statistics Modeling



## Poisson Statistics Modeling



## Poisson Statistics Modeling



## Poisson Statistics Modeling



## Poisson Statistics Modeling



## Poisson Statistics Modeling



## Accounting for Experimental Uncertainties (from Friday's colloquium by Holger Mueller)

### 0.16 ppb systematic errors



## Accounting for Experimental Uncertainties (from Friday's colloquium by Holger Mueller)



Morel, Yao, Cladé, and Guellati-Khélifa 2020, Nature 588, 61-65

