







Solar Radiation Big Data Analysis for Strategic Positioning of Solar Panels

Solar Team: I-READ

Dean Haley & Cherrie Nelson

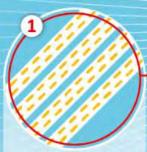
Faculty Mentors: Dr. Mohammad Hossain & Dr. Marsha Sowell

Student Mentor: Mr. Lovekesh Singh

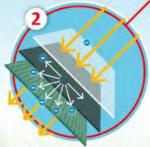
Industrial Advisor: Mr. Kevin Rees, P.E.

Background

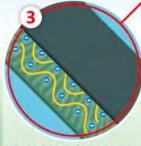
HOW DO SOLAR PANELS WORK?



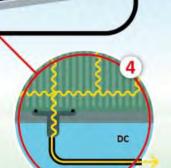
Light from the sun is made up of tiny particles called photons.



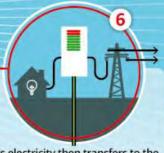
Photons hit the solar panel, knocking electrons free of the silicon atoms within the solar cells.



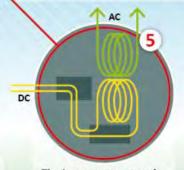
These loose electrons are now free to move, creating an electric current from the positive and negative charges now present. They pass through the layers of the solar cell.



The electric current flows to the edge of the panel into a conductive wire, which leads to the inverter.



This electricity then transfers to the building. If the solar panels are connected to the utility meter, excess electricity is transferred to the utility grid. This causes the meter to run backward, crediting the property for surplus generation.



The inverter converts the direct current (DC) to alternating current (AC), which is what is used to provide power.

Types of Solar Panels

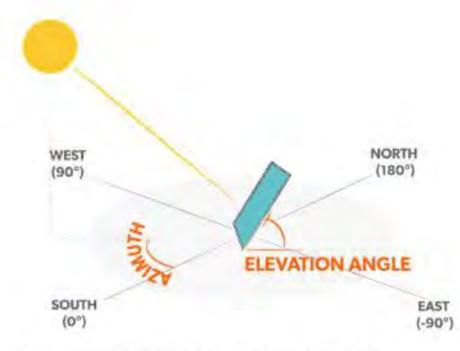
- 4 different types of solar panels:
 - ▶ 1) Monocrystalline: Also known as single-crystal panels, these are made from a single pure silicon crystal that is cut into several wafers. Since they are made from pure silicon, they can be readily identified by their dark black color. The use of pure silicon also makes monocrystalline panels the most space-efficient and longest-lasting among all solar panel types.
 - ▶ 2) <u>Polycrystalline</u>: As the name implies, these come from different silicon crystals instead of one. The silicon fragments are melted and poured into a square mold. This makes polycrystalline cells much more affordable since there is hardly any wastage and gives them that characteristic square shape. However, this also makes them less efficient in terms of energy conversion and space, since their silicon purity and construction are lower than monocrystalline panels.
 - 3) Passivated Emitter and Rear Cell (PERC) panels are an improvement of the traditional monocrystalline cell. This relatively new technology adds a passivation layer in the rear surface of the cell that enhances efficiency in several ways. It reflects light back into the cell, increasing the amount of solar radiation that gets absorbed. It reduces the natural tendency of electrons to recombine and inhibit the flow of electrons in the system. It allows greater wavelengths of light to be reflected. Light waves over 1,180nm can't be absorbed by silicon wafers and simply pass through, so they end up heating the cell's metal back sheet and reduce its efficiency. The passivation layer reflects these higher wavelengths and stops them from heating up the back sheet.
 - ▶ <u>4) Thin-film panels are characterized by very fine layers that are thin enough to be flexible. Each panel does not require a frame backing, making them lighter and easier to install. Unlike crystalline silicon panels that come in standardized sizes of 60, 72, and 96-cell counts, thin-film panels can come in different sizes to suit specific needs. However, they are less efficient than typical silicon solar panels.</u>

Types of Solar Panels (Cont.)

Information for each type of panel	Monocrystalline	Polycrystalline	PERC	Thin Film
Cost	1 to 1.50 per watt	.7 to 1 per watt	.32 to .65 per watt	.5 to 1 per watt
Life Expectancy	25+ years	15 to 20 years	Not Rated yet	10 to 20 years
Efficiency/Rating	20+ %	17 to 15%	Add up to 5%	15 to 8%
Advantage	Best efficiency without use of PERC. Longest life expectancy without PERC.	Lower cost.	Least space needed. Highest Power.	Use on any shape. Can be bent around surface. Not required to be sheet or rectangle size.
Disadvantage	Cost.	Low heat tolerance.	Most expensive. Quicker to lose effectiveness. Reduces results at high temperature.	Requires more space - for less energy generated. Not durable.

Solar Irradiance at Location: latitude vs. longitude

- ► Elevation angle in SOLPOS is the azimuth angle giving that the panel is facing North to South (180)
- Tilt Angle in SOLPOS is degrees tilt from horizontal panel



Tilt and azimuth angle in relation to the Equator.

Extraterrestrial Solar Irradiance (ETR) is the amount of solar radiation that would reach a location on Earth's surface measured in (Watts)/(Meter)²

Power Output

Global formula for estimating the electricity generated from a solar panel

E = ArHPR

- ► E Energy output in Wh
- ► A = Area of panel in m²
- r = Efficiency
- ► H = Solar Irradiance in W/m²
- ▶ PR = .75, Industry Standard

Thunderbolt

- E = (.75mX.75m)X(.237)X(1161)W/m²)X.75
- E = 116.08 W

Research Questions

- ▶ Using the Big Data available on SOLPOS (Solar Position and Intensity) calculator at NREL (National Renewable Energy Laboratory) website, obtain the solar energy yield throughout each day of the year at different locations and for a variety of tilt to find the best location and tilt.
- ► Compare the solar energy obtained from SOLPOS and experimental observation at various locations to understand the reliability of SOLPOS. Is there significant difference between simulation data from SOLPOS and experimental data?
- Measure the temperature on front and back of solar panels throughout the daytime hours to determine if there is a significant difference that can cause damage in the panel.
- Knowing that affordable solar panels are most commonly less than 25% efficient, what happens to the remaining energy gathered? Does the radiant energy transform to thermal?

Research Methodology

Methodology

- ► Conduct research on requirements to measure output from a solar panel.
- ▶ Determine the necessary properties of a solar panel in order to conduct the experiment.
- Download data from NREL SOLPOS Calculator, for different angles for different cities. Some cities had similar longitude and some cities had similar latitude.
- ▶ Use Microsoft Excel to analyze and graph data for different cities and angles.
- ▶ Determine if cities with similar longitude and latitude have variation in solar irradiance.
- ▶ Use actual solar panels to measure the output collected to observe the collected values compared to SOLPOS data for same location.

SOLPOS Interface

SOLPOS Calculator

Compute the solar position and intensity from time and location using NREL's SOLPOS.

Required input values: Enter start date: Year: 2005 | Month: January | Day: 1 | Part | Enter end date: Year: 2005 | Month: January | Day: 1 | Part | Enter output time interval: Interval: 10 | Units: Osecond @Minute

Enter site location information:

39.743	Latitude, degrees north (south negative)
-105.17	8 Longitude, degrees east (west negative)
-7.0	Time zone, east (west negative)
835.0	Surface pressure (mbar)
10	Ambient dry-bulb temperature (°C)

Optional input values:

180	Azimuth of panel surface
67	Degrees tilt from horizontal of panel
1361.1	Solar constant (W/m²)
7.6	Shadow-band width (cm)
31.7	Shadow-band radius (cm)
0.04	Shadow-band sky factor
0	Interval of a measurement period (sec)

Date	Time	Cos incidence	ETR tilt
1/1/2005	7:30:00	0.4827	680.0884
1/1/2005	7:40:00	0.5146	724.9449
1/1/2005	7:50:00	0.5462	769.4406
1/1/2005	8:00:00	0.5772	813.2216
1/1/2005	8:10:00	0.6077	856.0837
1/1/2005	8:20:00	0.6373	897.8829
1/1/2005	8:30:00	0.6662	938.5144
1/1/2005	8:40:00	0.6941	977.8844
1/1/2005	8:50:00	0.7211	1015.9075
1/1/2005	9:00:00	0.7471	1052.51
1/1/2005	9:10:00	0.772	1087.6155
1/1/2005	9:20:00	0.7958	1121.1552
1/1/2005	9:30:00	0.8185	1153.0638
1/1/2005	9:40:00	0.8399	1183.2784
1/1/2005	9:50:00	0.8601	1211.7417
1/1/2005	10:00:00	0.879	1238.3986
1/1/2005	10:10:00	0.8966	1263.1968
1/1/2005	10:20:00	0.9129	1286.089
1/1/2005	10:30:00	0.9278	1307.0327
1/1/2005	10:40:00	0.9412	1325.9868
1/1/2005	10:50:00	0.9532	1342.9152
1/1/2005	11:00:00	0.9638	1357.785
1/1/2005	11:10:00	0.9729	1370.5685
1/1/2005	11:20:00	0.9804	1381.2407
1/1/2005	11:30:00	0.9865	1389.7814
1/1/2005	11:40:00	0.991	1396.1742
1/1/2005	11:50:00	0.994	1400.4069
1/1/2005	12:00:00	0.9955	1402.4714
1/1/2005	12:10:00	0.9954	1402.3638
1/1/2005	12:20:00	0.9938	1400.0841
1/1/2005	12:30:00	0.9907	1395.6366
1/1/2005	12:40:00	0.986	1389.0303
1/1/2005	12:50:00	0.9797	1380.2769
1/1/2005	13:00:00	0.972	1369.3936
1/1/2005	13:10:00	0.9628	1356.4012
1/1/2005	13:20:00	0.9521	1341.3247
1/1/2005	13:30:00	0.9399	1324.1924

Research Data

- ▶ Using SOLPOS collect data at different location around the USA to determine if it is the longitude or latitude that makes a difference in the amount of solar ETR available
- ► Angle of the sun is different for different geographic location
- ▶ Determining the best locations what is the best angle at each location
 - ► Looking at differences in Latitudes: Corpus Christ (27° N) to Minneapolis (44°N)
 - ► Looking at different Longitudes: San Diego (117°W) to Charleston SC (79°W)

Comparison of Cities

Conditions/ City	Charleston, SC	Dallas, TX	Tucson, AZ	San Diego CA	CC, TX	Bishop, TX	KC, MO	OK, OK	MPLS, MN
Latitude	32.47	32.4645	32.1318	32.4254	27.4434	27.357	39.0559	35.287	44.5855
Longitude	79.555	96.4832	110.5535	117.0945	97.247	97.4758	94.3442	97.3117	93.1609
Time Zone	-5	-6	-7	-8	-6	-6	-6	-6	-6
STD	1atm/10	970/10	970/10	1atm/10	1 atm/12	1 atm/ 12	960/9	970/10	960/11

Experimental Work



Bishop, Texas

Back of Panels

Panels

- There are 48 panels for this set up.
- Each panel is rated at 250 watts.
- They are monocrystalline panels that were installed in 2021.
- The panels are facing North-South at a tilt angle of 20°
- The house runs mostly on these panels - the excess is sold back to the grid.

Bishop Solar Farm

Experimental Work (Cont.)

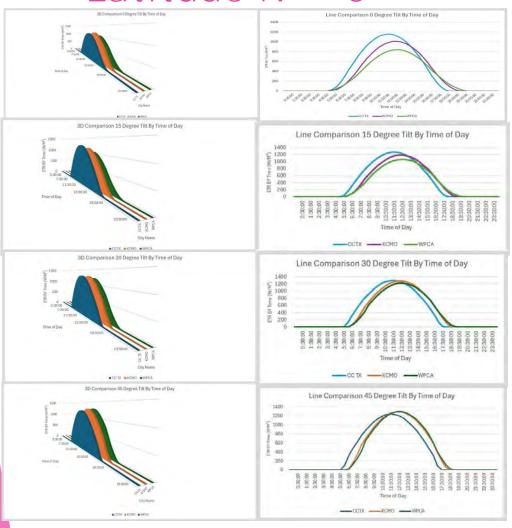
- ► Segway Panel
- ▶ 100 W panel
- ► Monocrystalline
- ► Efficiency rating at 23.7%



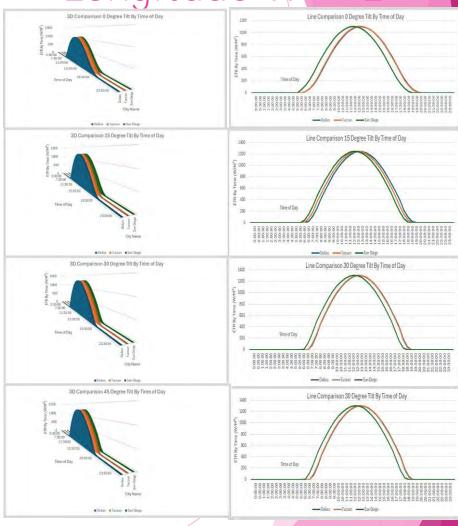
Results and Discussion

Big Data Analysis



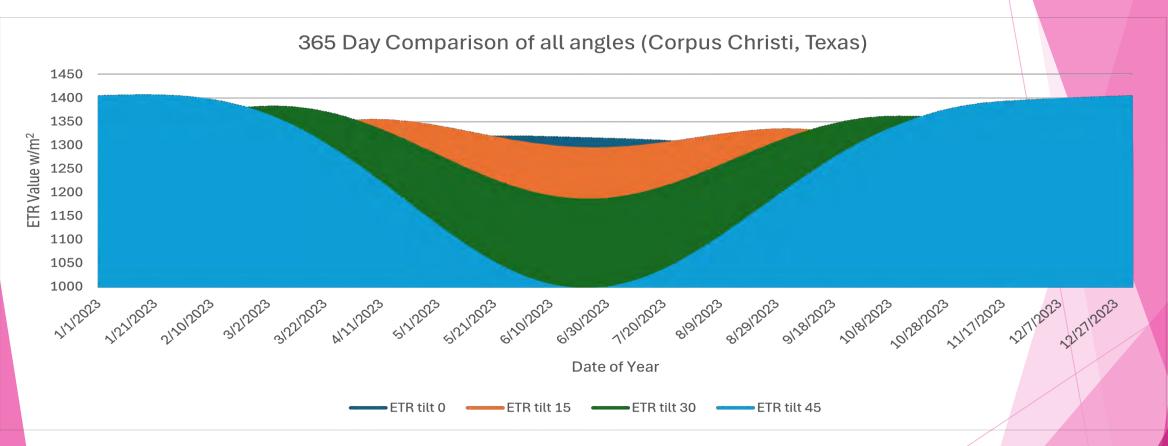


Longitude W<->E

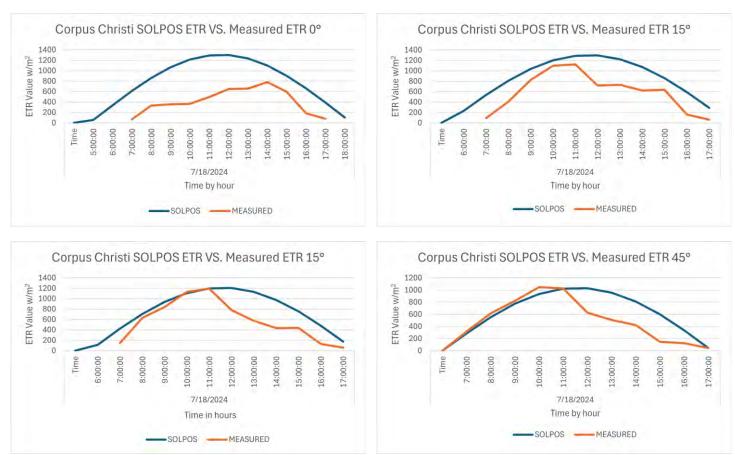


The data clearly demonstrated the variation of ETR differed greatly between cities of similar latitude while cities of similar longitude differed slightly.

Big Data Analysis 1 Year Analysis of ETR by Angle



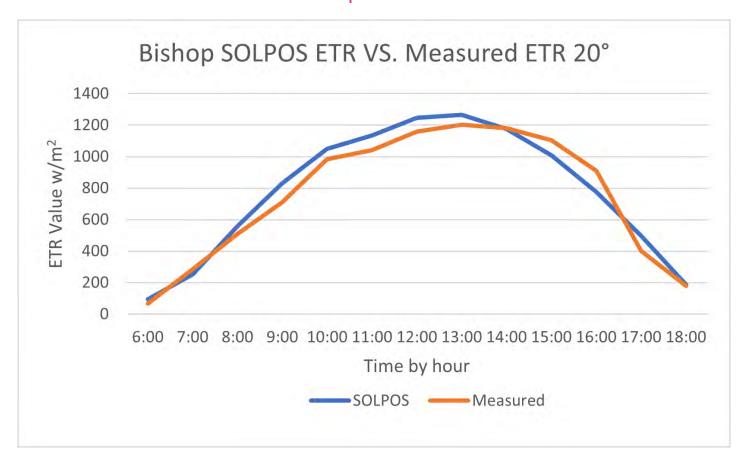
SOLPOS ETR VS MEASURED ETR Corpus Christi, Texas



The experimental data is consistently lower than the SOLPOS simulation data.

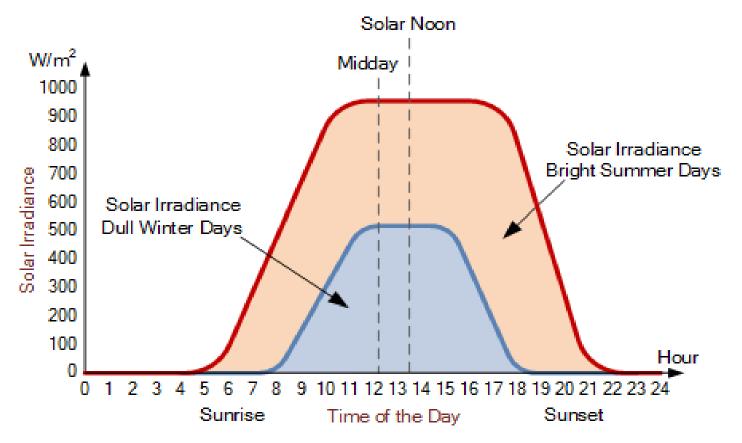
Cloud cover or obstructions – trees / buildings – possible explanations

SOLPOS ETR VS MEASURED ETR Bishop, Texas



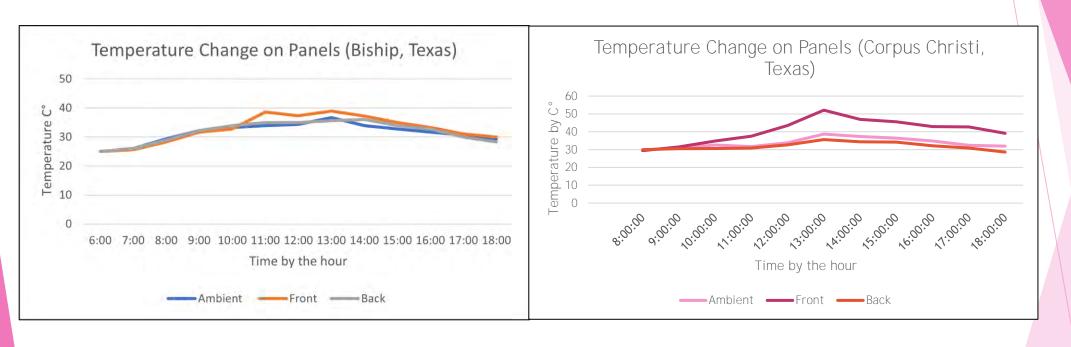
The Bishop data is tracking close between the SOLPOS simulation data and experimental data. Possible explanation: High quality solar panels / no obstruction (open field)

Solar Irradiation During the day, according to Alternative Energy



Partly cloudy days can create a result less than optimal ETR, such as the lower 2 areas shown in brown and blue.²¹

Panel Temperature Bishop, Texas and Corpus Christi, Texas



Bishop: Difference in temperatures between the front and the back of the solar panels is insignificant

Corpus Christi: Significant difference in temperature between the front and the back of the solar panels

Learning Modules

Module 1: Teaching students to use data to plot, interrupt and extrapolate information

- ▶ Each small group table will be given a city and the simulation data for Corpus Christi. The students will select their own city and locate the latitude and longitude for that city. Then using the SOLPOS students will collect the simulation data, create graphs with the data using Microsoft Excel or Google Sheets.
- After the students have made their graphs all cities' charts have been made, they will be displayed for the students to answer questions based on the information the student generated graphs and data information.
- Students will gather experimental data to compare with the simulation data then analyze using Microsoft Excel.
- ▶ Students will identify if the latitude or longitude changes the intensity for the city using Microsoft Excel to graph the class data.
- Students will provide report and presentation of their findings. <u>D:\solar module haley data analysis v3 pdf.pdf</u>
- https://www.dropbox.com/scl/fi/o6o7fbnkfs7hqrvgrdjed/solar-module-haley-data-analysis-v3-pdf.pdf?rlkey=s1w2yugge2u3ans7fcdorna71&dl=0

Module 1: Teaching students to use data to plot, interrupt and extrapolate information

- ▶ Each small group table will be given a city and the simulation data for Corpus Christi. The students will select their own city and locate the latitude and longitude for that city. Then using the SOLPOS students will collect the simulation data, create graphs with the data using Microsoft Excel or Google Sheets.
- After the students have made their graphs all cities' charts have been made, they will be displayed for the students to answer questions based on the information the student generated graphs and data information.
- Students will gather experimental data to compare with the simulation data then analyze using Microsoft Excel.
- Students will identify if the latitude or longitude changes the intensity for the city using Microsoft Excel to graph the class data.
- Students will provide report and presentation of their findings. <u>D:\solar module haley data analysis v3 pdf.pdf</u>
- https://www.dropbox.com/scl/fi/o6o7fbnkfs7hqrvgrdjed/solar-module-haley-data-analysis-v3-pdf.pdf?rlkey=s1w2yugge2u3ans7fcdorna71&dl=0

Module 2: Teaching the students to use data to plot, interrupt and extrapolate information

- ▶ Part 2 Extrapolate Information
- ► Each group will learn to find the maximum (highest solar intensity value) without being given a formula trail and error.
- ▶ Learn how to find highest solar intensity value by changing the tilt angle <u>D:\Cherie_Nelson_mod2_and_mod3_combined.pdf</u>

https://www.dropbox.com/scl/fi/yivgndl0yt2pybc63mbiq/Cherie_Nelson_mod2_and_mod3-combined.pdf?rlkey=zdnl4jo33ggq0tzzs1rbjrxvh&dl=0

Module 2 - Calculating SD, Mean and Range

- Range: The smallest to the largest number in a set of points
- ► Mean: The average value for a given set of data
- ► SD Standard Deviation: How much a number is away from the middle score.
- Here are the Steps to finding SD (standard Deviation)
- ▶ 1. Find the mean, (X) average, value for the data set
- **2.** For each point (μ), find the square of its distance from the mean
- 3. Sum (add) the values from step 2
- ▶ 4. Divide the sum of step 3 by the sample size minus 1 (this is called the variance)
- ▶ 5. Take the square root of the variance to get SD

Date			Change from mean	Change value Squared
		μ	Χ - μ ¯	(X - µ) ²
0/1/2025	4 am			
8/1/2025	6 am	231.8769		
8/1/2025	7 am	528.3141		
8/1/2025	8 am	793.3874		
8/1/2025	9 am	1008.422		
8/1/2025	10 am	1158.635		
8/1/2025	11 am	1233.758		
8/1/2025	12 pm	1228.632		
8/1/2025	1 pm	1143.6		
8/1/2025	2 pm	984.4446		
8/1/2025	3 pm	762.0406		
8/1/2025	4 pm	491.6748		
8/1/2025	5 pm	192.5114		
8/1/2025	6 pm	65.6425		
	Mean ETR value (X)		Sum of the (X-µ) ²	

Calculating the Efficiency of a Solar Panel

- From the first part 300 Watts maximum panel output for the highest solar panel (monocrystalline)
- Efficiency = Work out/Work In
- ► Efficiency = 300 W/ 1366 W = 0.2196 or 22%
- ► That means 78% of the energy is lost

Calculating the Efficiency of a Solar Panel

- To calculate error, you need to use the following formula for the Fluke meter:
- ► E = ArHPR where the E is what it should read (corrected value) A is the area of the panel r is the efficiency of the panel H is the solar constant (1161 kW/hr) PR is a default by the industry the accepted value is 0.75.
- E = (0.0002)(.25)(1161)(.75) = +/-3%
- The SOLPOS values are not including any of the following: solar flares or extra solar activities, any cloud or storms that will diminish the solar capacity.

Time	ETR From SOLPOS (W/m²)	Reading ETR Fluke (W/m²)	Change in Values	% error
6:00	44.8457	62	17	38 +/- 3 = 35%
7:00	326.1114	277	49	15 +/- 3 =12%
8:00	603.6946	499	105	17 +/- 3 = 14%
9:00	855.0256	722	133	16 +/- 3 = 13%
10:00	1062.6797	987	76	7 +/- 3 = 4%
11:00	1212.4233	1050	162	13 +/- 3 = 10%
12:00	1294.0376	1135	159	12 +/- 3 = 9%
13:00	1301.9274	1201	101	8 +/- 3 = 5%
14:00	1235.5479	1180	55	4 +/- 3 = 1%
15:00	1099.4076	980	119	11 +/- 3 = 8%
16:00	902.1545	805	97	11 +/- 3 =8%
17:00	385.2481	353	32	8 +/- 3 = 5%
18:00	101.4754	111	10	10 +/- 3 = 7%

Module 3: Interference with Solar Panel

- ► This is a week long experiment:
- ► TEKS Science 5(A) 8(D)
- Equipment needed:

Solar panel, Fluke measuring device with temperature gauges, level, polarized panels, colored panels to lay over the solar panel (Red, Blue, Yellow, Green, Purple/Violet, Orange), mirrors

D:\Cherie_Nelson_mod2_and_mod3 combined.pdf

https://www.dropbox.com/scl/fi/yivgndl0yt2pybc63mbig/Cherie_Nelson_mod2_and_mod3-combined.pdf?rlkey=zdnl4jo33ggq0tzzs1rbjrxvh&dl=0

Teacher Instruction:

- Students will take a solar panel and place it outside and collect the following information each hour:
- Students will be asked for the leading question
- What is solar power? energy from the sun that comes to the earth on electromagnetic waves
- ► How is it measured? W/m^s
- What about temperature? Is this transferred by the sun? (yes) How is it measured?
- What effects would happen when you ... (questions for them to think about and answer individually) - this is what we will be studying this week.
- use a mirror?
- Does color make a difference in the reading?
- Does polarization make a difference?
- What things can interfere with your readings?

30

Conclusions

Concluding Remarks

- ▶ The irradiance (ETR) was found to differ significantly between North and South locations, however little difference can be found between cities from West to East.
- The measured ETR for Corpus Christi, Texas was found to be consistently lower than the SOLPOS ETR data. Bishop, Texas experimental data matches well with the SOLPOS.
- ► Temperature difference between the front and back of the solar panels showed dependence on the location and positioning of the panel.
- ▶ Using the simulated data collected from SOLPOS suggests that changing angles during the day can increase the amount of energy collected.
- Learning modules were developed based on the research experiences and findings, to be implemented in the upcoming school year.

References

- 1) https://midcdmz.nrel.gov/solpos/solpos.html
- 2) https://www.pveducation.org/pvcdrom/properties-of-sunlight/solar-radiation-on-a-tilted-surface
- 3) https://www.sciencedirect.com/topics/engineering/solar-constant
- 4) https://www.nrel.gov/grid/solar-resource/solpos.html
- 5) https://www.nrel.gov/docs/fy24osti/87524.pdf
- 6) https://www.cleanenergyreviews.info/blog/most-powerful-solar-panels
- 7) https://www.sunbasedata.com/blog/how-to-calculate-solar-panel-output
- 8) https://solarpower.guide/solar-energy-insights/how-do-solar-panels-work

Acknowledgement

This material is based upon work supported by the National Science Foundation under Award No. 2206864. Any opinions, findings, and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the National Science Foundation.

Why is there no plant life here?



Analysis of Wind Speed Pattern Changes Before and After Wind Farm Operations

Debra Carpentier & Marisa Hamilton









Faculty Mentor:

Dr. Hua Li

Educational Mentor:

Dr. Marsha Sowell

Graduate Mentor:

Yahya Al Bustanji

Industry Mentor:

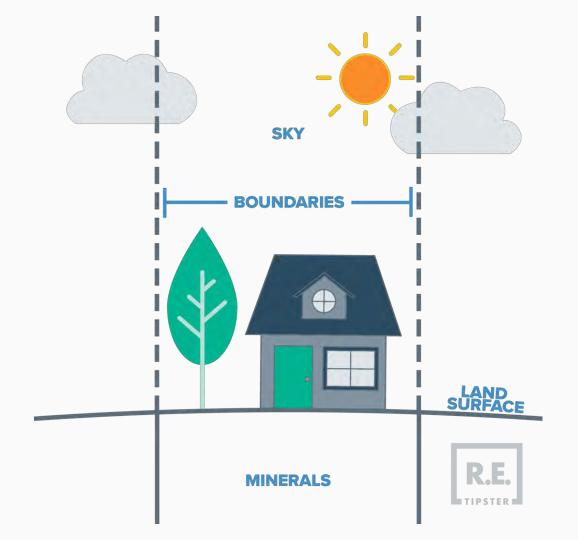
Rene Ramirez



This material is based upon work supported by the National Science Foundation under Award No. 2206864. Any opinions, findings, and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the National Science Foundation.

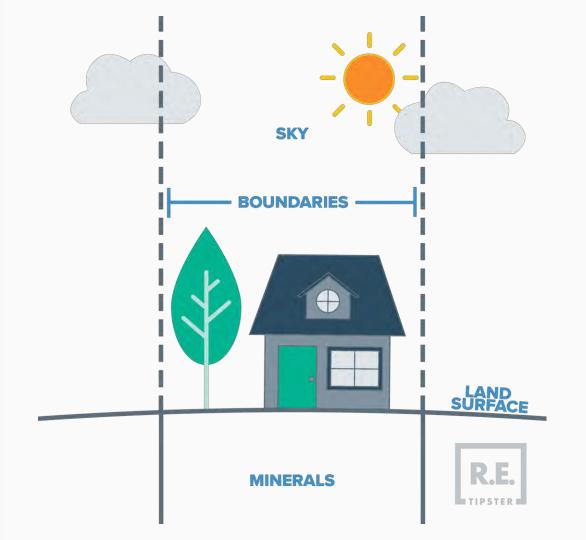
Land Rights

Surface Rights: include building structures, farming, trees for timber, accessing water below ground



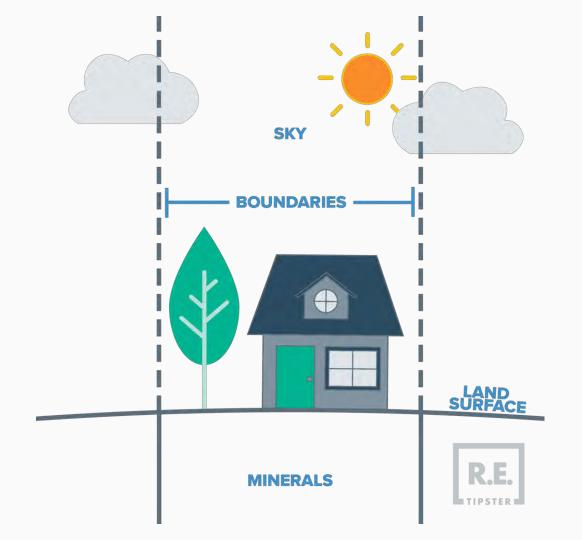
Land Rights

Mineral Rights: include oil, gas, coal, and other minerals below the ground

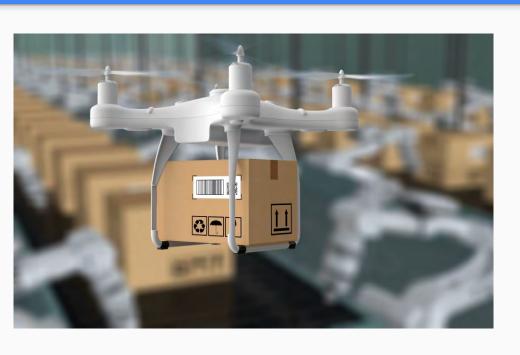


Land Rights

But what about the air above the land?



What can you do with the air above your property?



What can you do with the air above your property?





More landowners are now adding wind farms.



Wind Energy Facts¹ In Texas

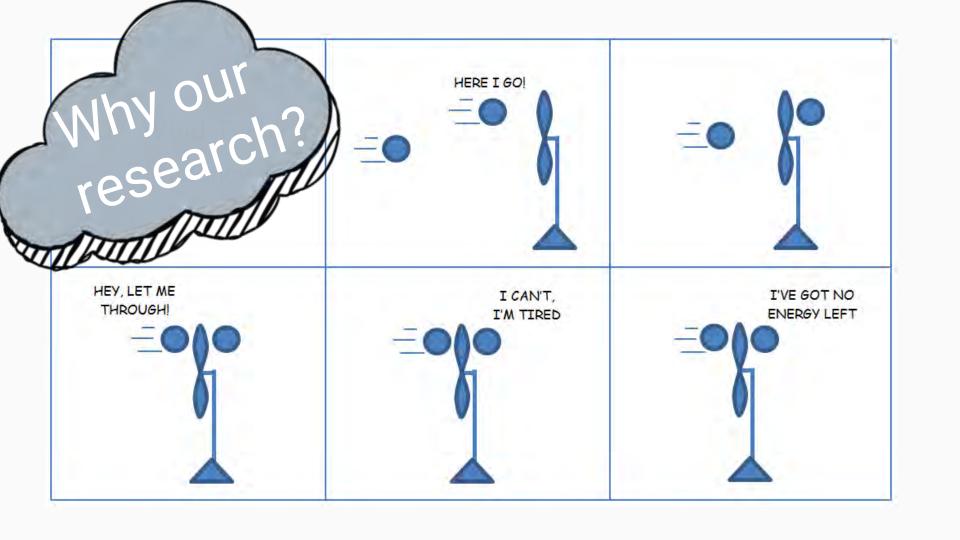
- In 1999, mandated 2,000 megawatts of renewable energy by 2009
- Goal achieved by 2005
- New challenge:10,000 megawatts by 2025
- By April 2016, 19,000 megawatts of renewables



¹ George W. Bush Helped Make Texas a Clean-Energy Powerhouse

Wind Energy Facts² in Texas

- In 2023, wind represented 28.6 percent of Texas energy generation, second to natural gas
- There are 239 wind-related projects in Texas and more than 15,300 wind turbines, the most of any state.
- In 2021, gross domestic product for wind electric power generation was \$1.7 billion



Why Our Research?

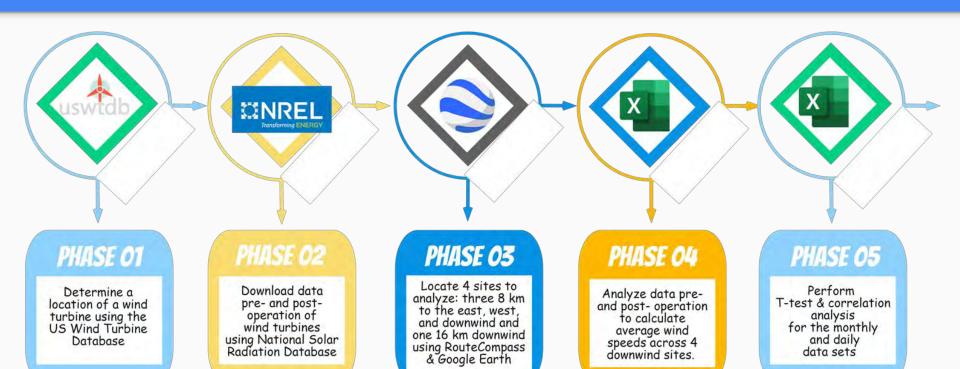
RESEARCH QUESTION

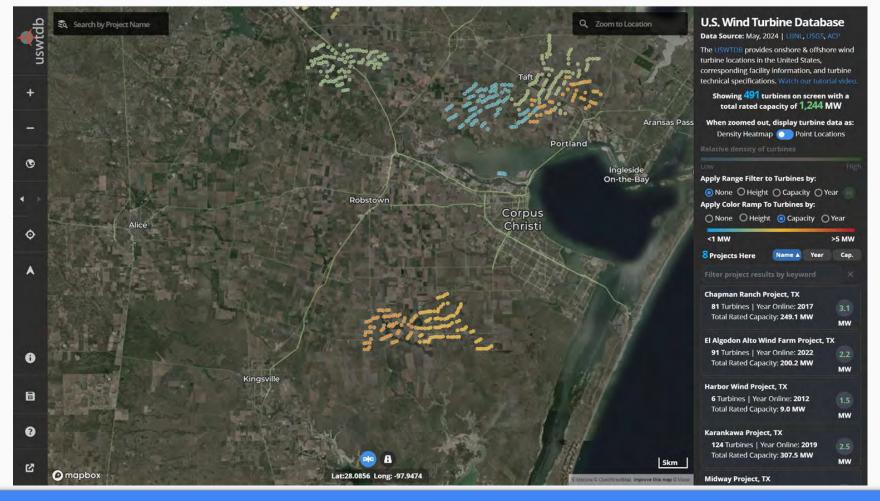
Does a wind farm influence the wind speed of surrounding areas?

RATIONALE

The operation of a wind farm may diminish the potential energy production of nearby sites.

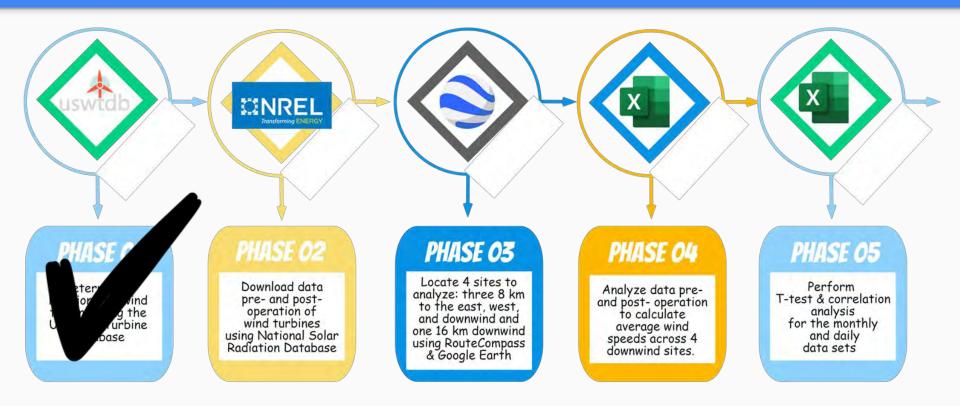
Research Plan



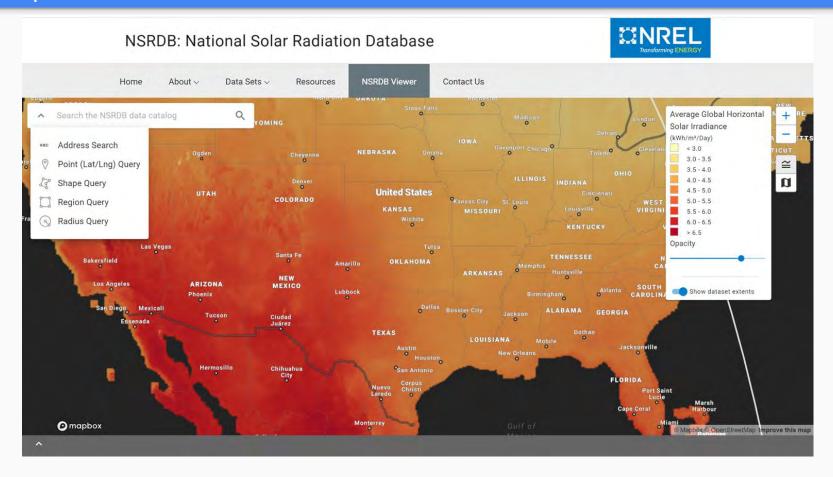


Phase 1: U.S. Wind Turbine Database was used to locate turbines to analyze

Research Plan



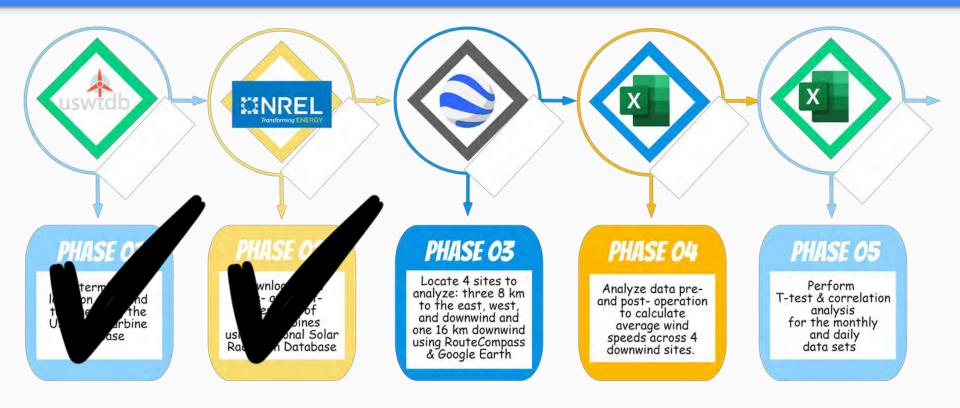
Phase 2: Data Acquisition from the National Solar Radiation Database to analyze wind speed & direction

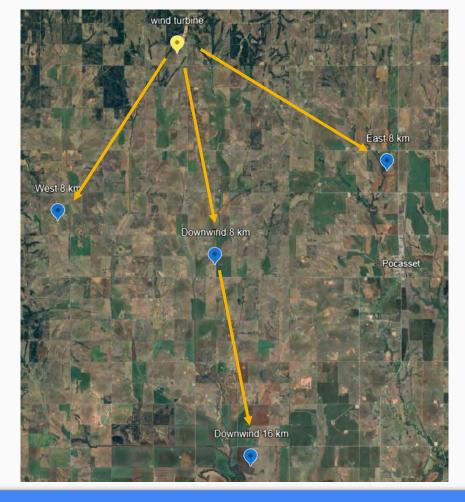


Phase 2 Results: wind speed & direction pre- and post- construction of wind turbine



Research Plan

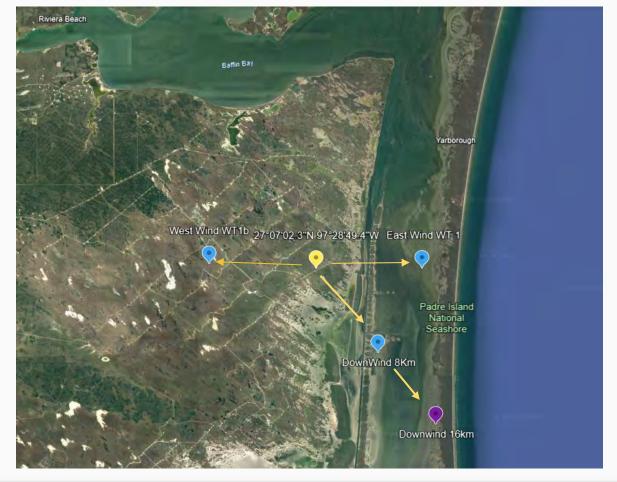


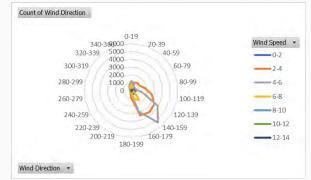




Routecompass.mapog.com was used to locate 4 sites. Those locations were translated using Google Earth

Phase 3: Determine 3 sites downwind 8 km and 1 16 km from turbine

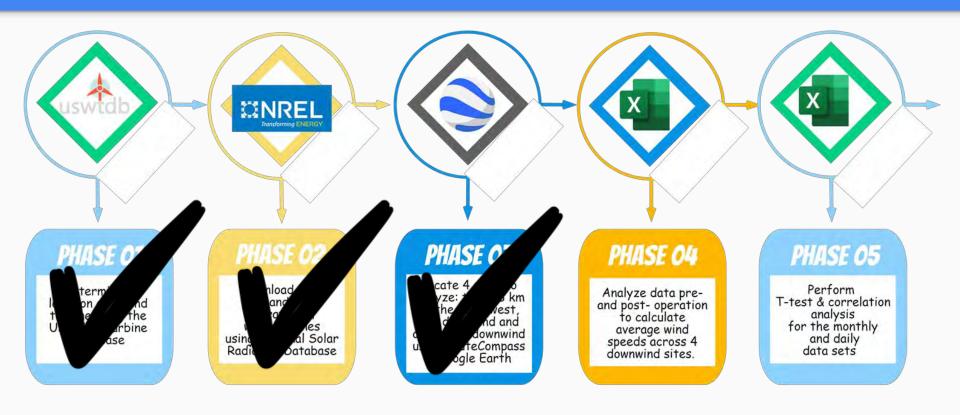




Downwind determined by a Radar Chart using excel and data received online at National Solar Radiation Data Base (NRDB)

Phase 3: Determine 3 sites downwind 8 km and 1 downwind 16 km from turbine

Research Plan





File Edit View Favorites Tools Help	
→ → × i	
Add Extract Test Copy Move Delete Info	
C:\Users\mhamilton\Desktop\c766abd3aa57a4e702c0a1be52co	ee60b.zip\c766abd3aa57a4e702c0a1be52cee60b\
Name	Size
3 674893_27.4997.86_1998.csv	495 324
674893_27.4997.86_1999.csv	496 579
674893_27.4997.86_2000.csv	497 174
674893_27.4997.86_2001.csv	497 856
674893_27.4997.86_2002.csv	495 860
674893_27.4997.86_2003.csv	497 316
3 674893_27.4997.86_2004.csv	496 788
674893_27.4997.86_2005.csv	498 583
674893_27.4997.86_2006.csv	496 831
3 674893_27.4997.86_2007.csv	495 191
3 674893_27.4997.86_2008.csv	496 872
3 674893_27.4997.86_2009.csv	496 190
3 674893_27.4997.86_2010.csv	497 312
3 674893_27.4997.86_2011.csv	497 817
3 674893_27.4997.86_2012.csv	497 276
3 674893_27.4997.86_2013.csv	495 944
3 674893_27.4997.86_2014.csv	496 497
674893_27.4997.86_2015.csv	495 432
3 674893_27.4997.86_2016.csv	495 528
3 674893_27.4997.86_2017.csv	494 136
3 674893_27.4997.86_2018.csv	496 280
3 674893_27.4997.86_2019.csv	494 360
3 674893_27.4997.86_2020.csv	496 232
674893_27.4997.86_2021.csv	495 755
3 674893 27 49 -97 86 2022 csv	495 784



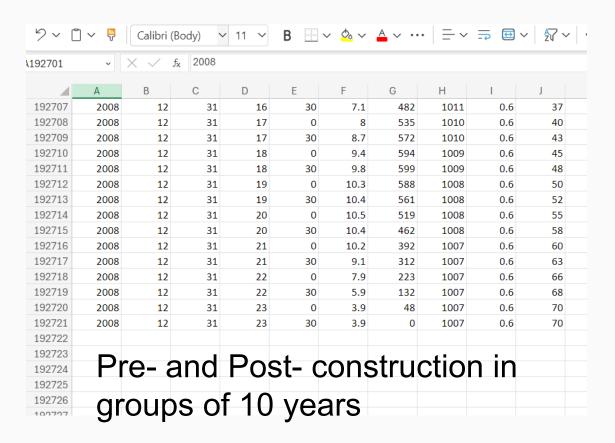
d	Α	В	C	D	E		F G		H	11	J	K	L
1	Year	Month	Day	Hour	Minut	e	Temperati GHI		Pressure	Precipitab W	/ind Dire W	Vind Speed	
2	1998		1	1	0	0	3.2	0	1013	0.9	152	0.9	
3	1998		1	1	0	30	3	0	1013	0.9	153	0.9	
4	1998		1	1	1	0	2.8	0	1013	0.9	154	1	
5	1998	-	1	1	1	30	2.6	0	1013	0.9	155	1.1	
6	1998		1	1	2	0	2.5	0	1013	0.9	156	1.1	
7	1998		1	1	2	30	2.3	0	1013	0.9	157	1.2	
8	1998		1	1	3	0	2.1	0	1013	0.9	158	1.2	
9	1998		1	1	3	30	1.8	0	1012	0.9	160	1.2	
10	1998		1	1	4	0	1.5	0	1012	0.8	162	1.2	
11	1998		1	1	4	30	1.2	0	1012	0.8	165	1.2	
12	1998		1	1	5	0	0.9	0	1012	0.8	167	1.2	
13	1998		1	1	5	30	0.6	0	1012	0.8	170	1.2	
14	1998		1	1	6	0	0.2	0	1012	0.8	172	1.2	
15	1998		1	1	6	30	0	0	1012	0.7	175	1.2	



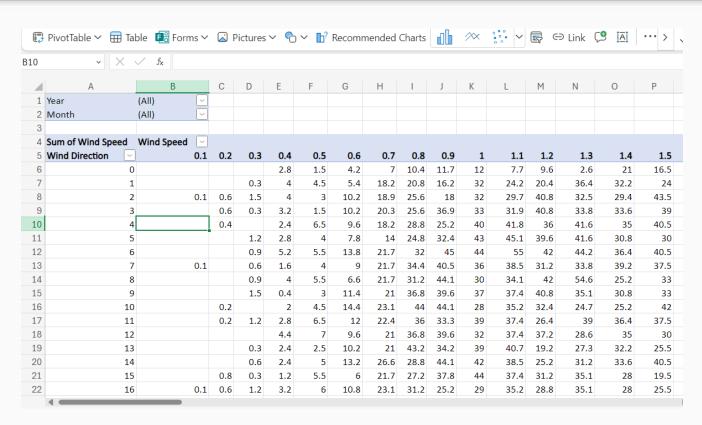
1 file

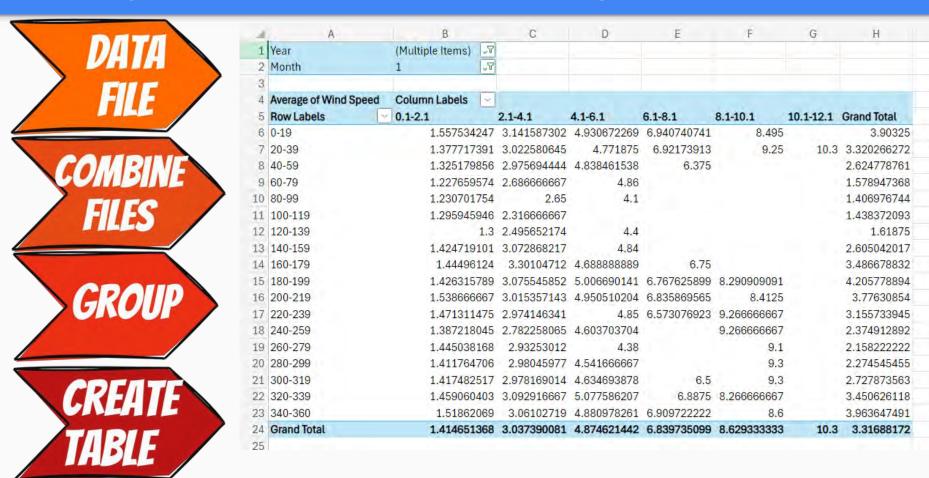
	A	В	C	D	E	F	G	Н
17511	1998	12	31	17	30	2.8	113	
17512	1998	12	31	18	0	2.8	117	
17513	1998	12	31	18	30	2.7	120	
17514	1998	12	31	19	0	2.7	124	
17515	1998	12	31	19	30	2.6	125	
17516	1998	12	31	20	0	2.6	126	
17517	1998	12	31	20	30	2.4	125	
17518	1998	12	31	21	0	2.3	123	
17519	1998	12	31	21	30	1.9	119	
17520	1998	12	31	22	0	1.5	115	
17521	1998	12	31	22	30	1.3	109	
17522	1998	12	31	23	0	1.1	103	
17523	1998	12	31	23	30	1.1	103	
17524								
17525								
17526								











17,520,000 data points



17,520,000 data points



approx. 20 miles tall

Phase 4: Examples of Pivot tables of inland average wind speed data monthly and daily

SITE 2 Average Wind Speed 8 km downwind from turbine

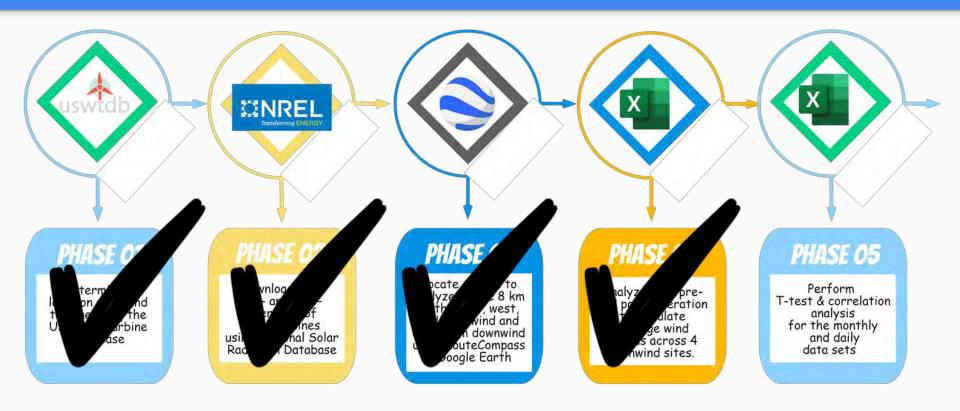
	\Box	
\vdash	К	□-

	<u> </u>
	10 years prior
January	3.251631232
February	3.489779491
March	3.677871457
April	3.915984848
May	3.458620479
June	3.28292298
July	3.004282747
August	2.8153348
September	2.940593434
October	3.304062805
November	3.213787879
December	3.259133675

POST- Wind Turbine Operation

	1 year	2 years	3 years	5 years	10 years
January	2.860618	3.234577	3.179256	3.295336	3.317036
February	3.643527	3.515662	3.526935	3.501086	3.5416
March	3.765457	3.78498	3.725605	3.601909	3.652702
April	4.381389	3.949861	3.930324	3.9275	3.822139
Мау	4.36297	4.048051	4.011246	3.824341	3.582137
June	4.625972	3.940764	3.780486	3.642347	3.395896
July	2.940457	3.014348	2.983938	3.001774	2.860168
August	2.943884	2.989483	2.850851	2.80207	2.75496
September	3.145347	2.990104	2.929722	2.990986	2.924903
October	3.223589	3.363743	3.433356	3.248038	3.362036
November	3.876944	3.563056	3.60125	3.631972	3.424493
December	2.862298	3.132964	3.078651	3.10457	3.09328

Research Plan



Phase 5: Analysis of Data

- T-Test Analysis of average wind speed data
- Null hypothesis: There is no significant difference of average wind speed between pre- and postoperation of a wind farm in surrounding areas.
- $\alpha = 0.05$

Phase 5: Monthly Avg Wind Speed T-Test Analysis Results

Data: Minco, Oklahoma Wind Turbine Location

	West 8 km	Downwind 8 km	East 8 km	Downwind 16 km
1 year	0.1027	0.1194	0.1043	0.078
2 year	0.0330	0.0459	0.0335	0.037
3 year	0.0733	0.1064	0.1139	0.008
5 year	0.1169	0.1745	0.1252	0.061
10 year	0.5798	0.7721	0.6004	0.495

Average Wind Speed 8 km				
	Pre- constr	Post- constr		
1 year		3.369		
2 year		3.353		
3 year	3.288	3.404		
5 year		3.314		
10 year		3.284		

Phase 5: Monthly Avg Wind Speed T-Test Analysis Results

Data: Baffin Bay Wind Turbine Location

	West 8 km	East 8 km	Downwind 8 km	Downwind 16 km
1 year	0.0284	0.0292	0.0292	0.029
2 year	0.044	0.0458	0.0456	0.046
3 year	0.0159	0.0164	0.0163	0.017
5 year	0.107	0.1067	0.1067	0.108
10 year	0.93	0.9285	0.9323	0.952

Average Wind Speed 8 km				
	Post- constr			
1 year		4.283		
2 year		4.142		
3 year	4.012	4.133		
5 year		4.092		
10 year		4.032		

Correlation Analysis

Correlation Coefficients for various weather variables at 8 km downwind

Wind Speed and	Coastal	Inland
Temperature	-0.045	0.090
Pressure	-0.040	-0.241

Conclusion

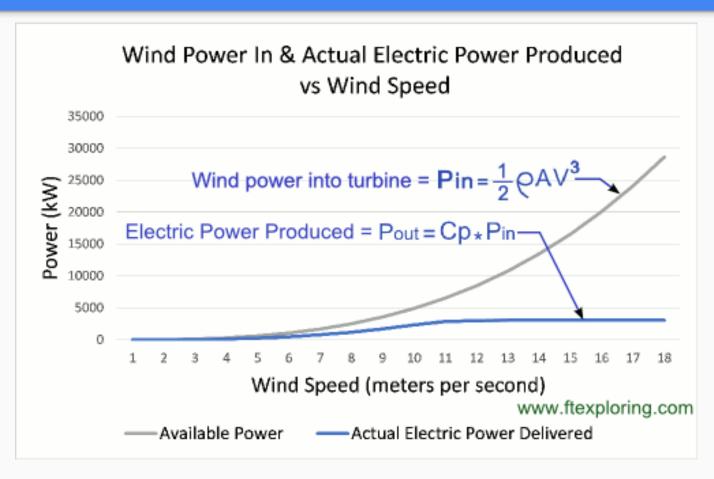
A significant increase in the average wind speed for both the inland and coastal wind farms and the surrounding areas.

Conclusion

A significant increase in the average wind speed for both the inland and coastal wind farms and the surrounding areas.

Inland and coastal wind farms are showing different numbers of significant differences

Conclusion



Future Research

Investigate other factors could influence wind speed

Future Research

Investigate other factors could influence wind speed

Examine terrain for possible impediments

Future Research

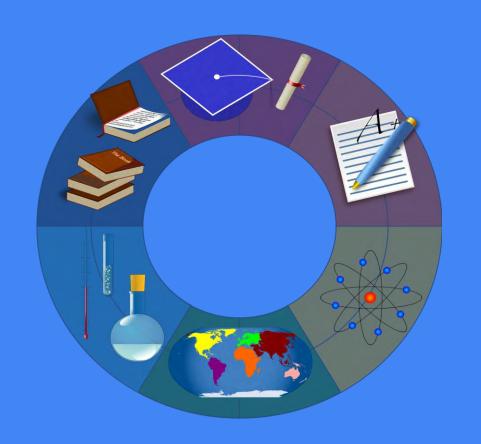
Investigate other factors could influence wind speed

Examine terrain for possible impediments

Explore how location differences affect wind speed

Curriculum Modules:

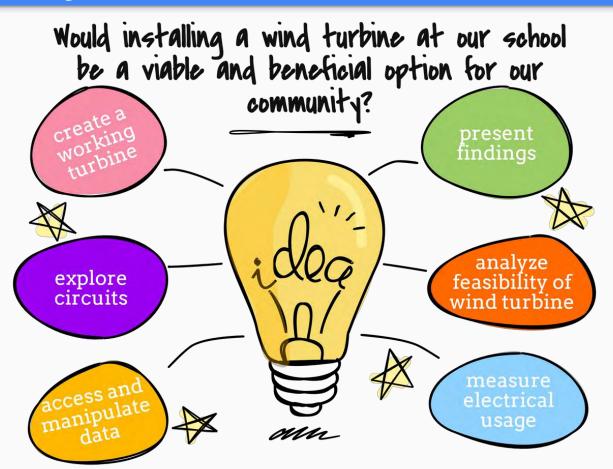
Physics and Chemistry





Both modules will be utilizing the Tempes t Weather System with Built-in Wind Meter, Rain Gauge, and Accurate Weather Forecasts

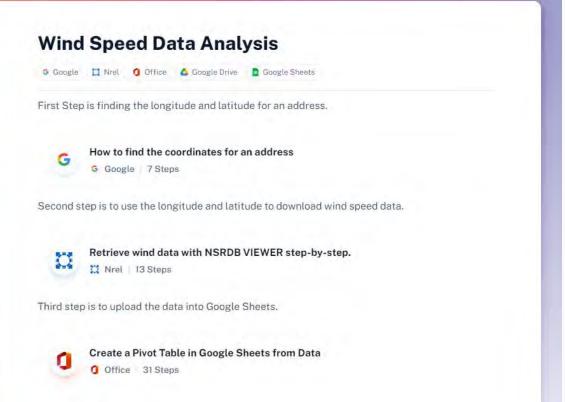
Physics Modules:

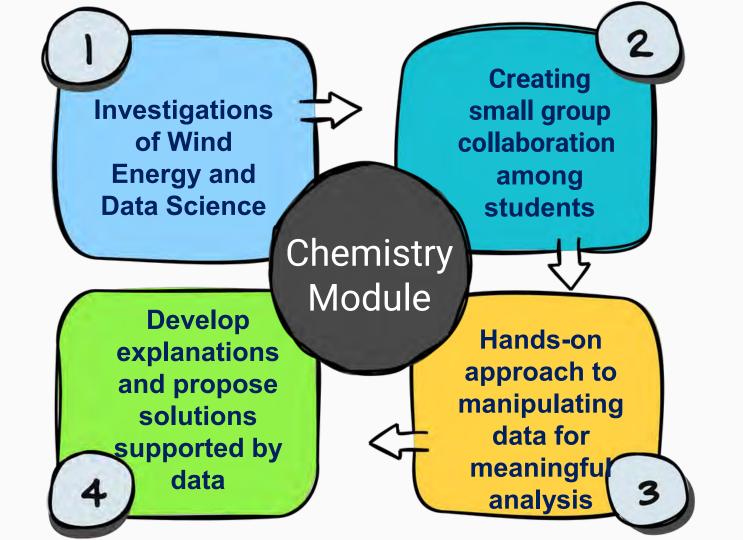


- 3-to-4-week project
- Incorporating electrical, energy, and power
- Science content
 AND science &
 engineering practices



Phy Cur





Chemistry Module

Day 1: Question student's about wind turbines

Day 2: Data Science focusing on Temp, Pressure & Solar Radiation

Day 3: Data Analysis

Day 4: Communication

Acknowledgements





Principle Investigator: Dr. Mohammad Hossain

Faculty Mentor: Dr. Hua Li

Educational Mentor: Dr. Marsha Sowell

Graduate Mentor: Yahya Al Bustanji



Superintendent: Dr. Veronica Alfaro

SGA Principal: Charles Odom



Industry Mentor: Rene Ramirez



Superintendent: Dr. Cissy Reynolds-Perez Asst. Superintendent: Dr. Ada Besinaiz Director of C & I: Dr. Olivia Ballesteros

HMK Principal: Dana Moore

Questions or Comments

Effect of Daylighting on Students' Learning and Classroom Electricity Consumption

Texas A&M University – Kingsville

Teacher Participants:

Kelsey Correa-Sinton ISD Teresa Cherry-CCISD



Faculty

Mentors: Dr. Hui Shen

Dr. Marsha

Sowell



Student

Mentor:

Curtis

Davenport



Industrial Advisor:

Ralph Pitzer, P.E



Illuminating Learning: The Impact of Daylight on Student Performance and Energy Efficiency

Our project:

Little scientific data exists on students' learning efficiency under different daylighting levels to enhance learning experience.

Research question:

How and to what extent daylighting affect students' learning efficiency and how much electricity can be saved via daylighting in the classrooms?

Background:

What is daylighting?

Daylighting refers to the practice of using natural light from the sun to illuminate the interior spaces of buildings.

Why daylighting?

This design strategy not only reduces the need for electric lighting (energy saving), but also has been linked to various benefits, including improved student performance in educational settings.

Background: Key Terms

- Daylighting-the practice of using natural light from the sun to illuminate the interior spaces of buildings
- Pyranometer-device that measures global solar radiation
- Photometric Sensor-device that measures light as seen by the human eye in lux or lumens

building. The UDI range is usually defined between 100 and 2,000 lux.

- Lux-the SI unit of illuminance, equal to one lumen per square meter. (500 lux is the ideal level
- for workspaces)
- Daylight Autonomy (DA)- represents the percentage of the occupied time during which a space receives enough daylight to meet the required lighting level (500 lux), without the need
- for artificial lighting.
- Useful daylight illuminance (UDI)-a metric used to assess the quality of natural light in a

Background: Case Studies

Case 1:	Case 2:	Case 3:	Case 4: Strategic Daylighting in Schools: More is not always better [4]	
Daylighting Impacts on Human Performance in School _[1]	The Impact of Classroom Design on Pupils' Learning: Final Results of a Holistic, Multi-Level Analysis. [2]	Analysis of The Performance of Students in Daylit Schools [3]		
Improved Student Performance	Impact on Learning	Health and Attendance	Glare vs. Window Position	
 Faster progress in math and reading tests 	Light levels affect student learning	Full spectrum lighting: healthier students, better attendance	 Glare impacts tech use more than window position 	
Higher Attendance Rates	Window Orientation		-	
More daylight leads to better attendance	Orientation matters more than size; glare risk	Daylit libraries: lower noise levels	Sun shades can reduce glare	
Energy Savings	Artificial Light Quality	noise levels	giaio	
Less reliance on artificial lighting	 Positive correlation with student performance Temperature Control Controlled temperature 	Full spectrum lighting improves student moods Performance Correlation	Window Size ■ Size ≠ amount of useful daylight	
	improves performance	Studies show positive impost on student		
		impact on student performance with full spectrum lighting	5	

Research Materials: Human and Non-Human Data Collection



LI-COR LI-210R Photometric Sensor



LI-COR LI-200R Pyranometer



EME Systems UCLC
Amplifier



HOBO 4-Channel Data Logger

Measures Light

 Measures amount of light in an environment

Human Perception

• Similar to how the eye perceives brightness (photometric measurements)

Error %

• Has an absolute error of 3%.

Measures Solar Radiation

measures the amount of solar radiation (sunlight) reaching a surface.

Error %

• Has an absolute error of 3%.

Boosts Signals

signals

Enhances Sensitivity

• Improves sensor sensitivity

Strengthens weak electrical

Data Collection

 Collects and stores data from up to four sensors

Simultaneous Measurements

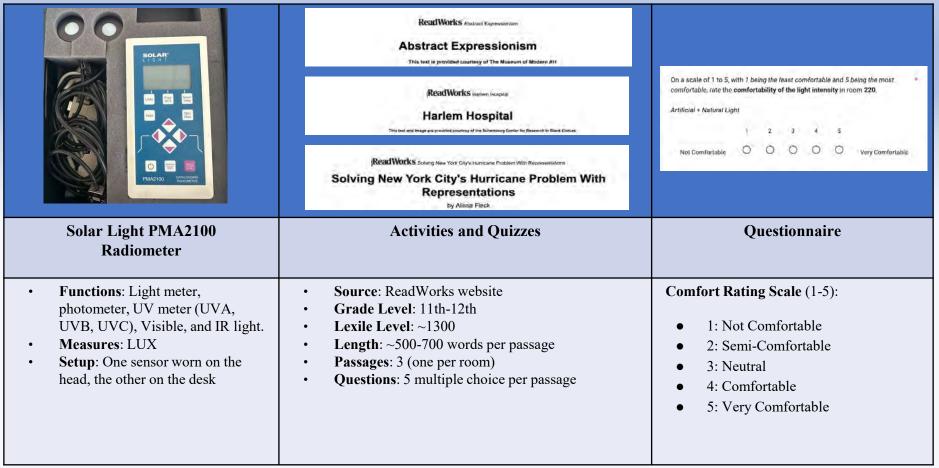
Supports four different measurements at once

Ease of Use

 Simple setup and data retrieval with software for analysis

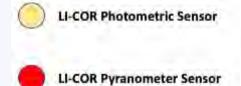
6

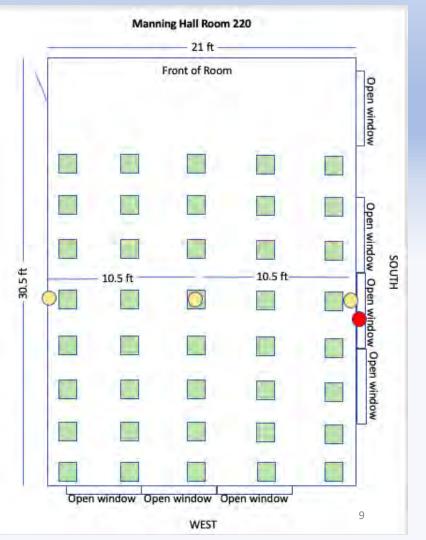
Research Materials: Human Data Collection



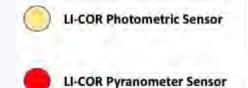
- Conditions: Windows open, electric lights off
- Data Collection Period: 72 hours (Friday 7/5 12:00 PM to Monday 7/8 12:00 PM)
- Locations: Manning Hall Rooms 220 and 224
- Data Collection: Every 5 minutes, undisturbed
- Sensor Placement: Refer to the figure on the next slides

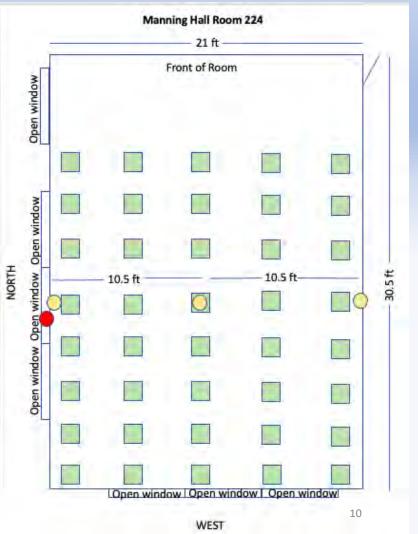
Sensor	Distance from Window (feet)	
1.Pyranometer	0 (on window)	
2.Photometric	o (on windowsill)	
3.Photometric	10.5	
4.Photometric	21	





Sensor	Distance from Window (feet)	
1.Pyranometer	o (on window)	
2.Photometric	o (on windowsill)	
3.Photometric	10.5	
4.Photometric	21	



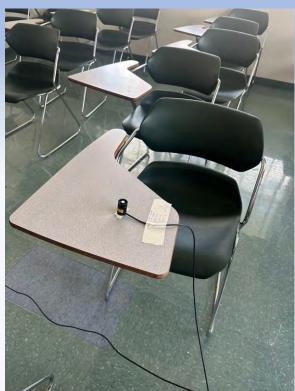














Data Analysis: Non-Human Data Collection

Analysis:

- Calculated DA (Daylight Autonomy) and UDI (Useful Daylight Illuminance) for each sensor in each room
- Assessed energy savings for each room over the collection period
- Data used: 8:00 AM 5:00 PM (typical working/school hours)

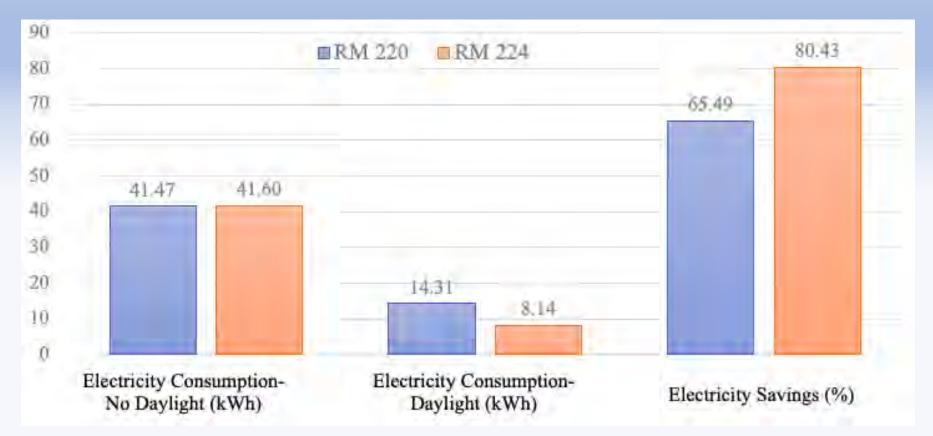
Data Analysis: DA and UDI Room 220

Manning RM 220 South and West Windows 8am-5pm				
LUX	Windowsill	10.5 ft from Window	21ft from Window	
DA >500	98.77%	39.51%	0.00%	
UDI (>=100,<500)	1.23%	60.49%	2.16%	
UDI (>=500,<1000)	15.74%	39.51%	0.00%	
UDI (>=1000,<2000)	39.51%	0.00%	0.00%	
UDI (>3000)	15.74%	0.00%	0.00%	

Data Analysis: DA and UDI Room 224

Manning RM 224 North and West Windows 8am-5pm				
LUX	Windowsill	10.5 ft from Window	21ft from Window	
DA >500	100.00%	39.69%	7.69%	
UDI (>=100,<500)	0.00%	60.31%	89.54%	
UDI (>=500,<1000)	6.77%	39.69%	7.69%	
UDI (>=1000,<2000)	61.54%	0.00%	0.00%	
UDI (>3000) 5.85%		0.00%	0.00%	

Data Analysis: Energy Savings



Setup:

• Participants placed one Solar Light PMA2100 sensor on their head and one on the desk.

Task:

• Read a passage (~1300 Lexile Level, 500-700 words) and complete a 5-question quiz.

Comfort Rating:

• After each quiz, participants rated light comfort on a scale of 1-5 (1 = Not Comfortable, 5 = Very Comfortable).

LUX Measurement:

• Recorded LUX level after each passage before moving to the next room.

Lighting Scenarios:

- Mixed Lighting: Manning Hall Room 220
- Only Electric Lighting: Manning Hall Room 222
- Only Natural Lighting: Manning Hall Room 224







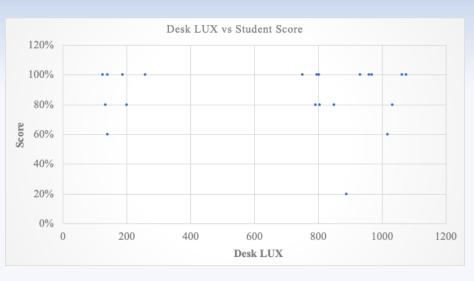
Manning Hall Room 220 Mixed Lighting

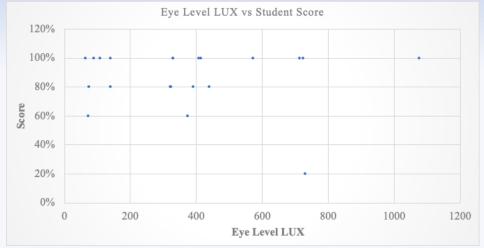
Manning Hall Room 222 Electric Lighting

Manning Hall Room 224 Natural Lighting

Data Analysis: Human Data Collection

• Our plan was to investigate whether there is a correlation between LUX levels and student scores. However, due to the small sample size, the data is inconclusive.





Data Analysis: Human Data Collection

• Participant feedback on the comfortability of the light in each room is shown below.

Participant	Mixed Light RM 220	Electric Light RM 222	Natural Light RM 224
1	2	4	5
2	2	4	5
3	2	2	5
4	3	5	1
5	3	5	4
6	3	4	5
7	4	3	5
Average Rating:	2.71	3.86	4.29

KEY				
1	2	3	4	5
Not Comfortable	Semi-Comfortable	Neutral	Comfortable	Very Comfortable

Conclusion:

UDI Range: Ideal UDI (500-1000 lux) achieved ~40% of the time in the center of each classroom.

Energy Savings:

- Room 220: 65.49% savings with daylighting
- Room 224: 80.43% savings with daylighting

Participant Preference: Room 224 (natural lighting) was favored.

Lux vs. Student Score: No significant correlation found.

Curriculum Module: Algebra 1

Analyzing the Effects of Daylight on Student Performance

Objective:

Students will investigate the relationship between daylighting and student performance.

TEKS:

- •calculate, using technology, the correlation coefficient between two quantitative variables and interpret this quantity as a measure of the strength of the linear association **A.4(A)**
- •compare and contrast association and causation in real-world problems A.4(B)
- •write, with and without technology, linear functions that provide a reasonable fit to data to estimate solutions and make predictions for real-world problems A.4(C)

Materials:

- Photometric sensors
- Cumulative worksheets
- Graph paper and graphing software
- Calculators

Curriculum Module: Algebra 1



Data Collection

Students will complete a cumulative worksheet in a room using natural light, and record LUX level using photometric sensor.

Data Collection

Students will complete a similar cumulative worksheet in a room using electric light and record LUX level using photometric sensor.

Linear Regression

Students will create a scatter plot of the data collected (LUX and Student Score), then determine the linear regression equation that best fits the data.

Correlation Coefficient

Students will calculate the correlation coefficient using technology and determine the strength of the relationship.

Correlation vs. Causation

Students will reflect on their findings and discuss whether the correlation they found suggests a causal relationship, and consider other variables and potential confounding factors.

Assessment

Students will complete a reflective activity.

Extension: Challenge students to explore non-linear relationships between daylighting and student performance using different types of regression models.

Curriculum Module: Biology

Types of Lighting Influence The Growth of Plants

Objective:

Students will investigate how different types of lighting influence plant growth. They will identify and analyze factors affecting growth and energy conservation in plants, and understand the role of light in photosynthesis and plant development.

TEKS Alignment:

- §112.34. Biology, Grade 9-12:
 - (1) Scientific and engineering practices: Plan and carry out investigations, analyze data, and communicate findings.
 - (9)(A): Compare structures and functions of different cell types.
 - (10)(B): Investigate interactions in an ecosystem and effects on the environment.
 - (11)(B): Investigate and analyze conditions necessary for plant growth, including light.

Curriculum Module: Biology

Types of Lighting Influence The Growth of Plants

Materials:

- Potted plants (same species, same size)
- Different types of light sources (e.g., natural sunlight, incandescent bulbs, fluorescent bulbs, LED lights, solar lamps)
- Light meters (possible phone app)
- Rulers or measuring tapes
- Data recording sheets or notebooks
- Graph paper or software for data visualization
- Watering can and water
- Soil
- Timer or clock

Curriculum Module: Biology

Day 1

Set up Plants

Determine watering schedule,

Identify variables

Initial measurements

2-3 Weeks

Measure plant growth for different lighting types

Record data in a notebook

2 Days

Analyze Data

Graph the data and interpret

Compare with other groups with different lighting

Final day

Determine the conclusion and assessment

Form conclusion based on the data of their group and other groups.

Take Quiz

Acknowledgements:

This material is based upon work supported by the National Science Foundation under Award No. 2206864. Any opinions, findings, and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the National Science Foundation.

References:

- 1. Heschong, L., Wright, R. L., & Okura, S. (2002). Daylighting Impacts on Human Performance in School.
- 2. Barrett, P., Davies, F., Zhang, Y., & Barrett, L. (2015). The impact of classroom design on pupils' learning: Final results of a holistic, multi-level analysis. Building and Environment, 89, 118–133.
- 3. Nicklas, M H, & Bailey, G B. Analysis of the performance of students in daylit schools. United States.
- 4. Bernard, M. Strategic Daylighting in Schools: More Is Not Always Better (2021). Architectmagazine.com. https://www.architectmagazine.com/technology/lighting/strategic-daylighting-in-schools-more-is-not-always-better
- 5. *Hobo 4-channel analog data logger*. HOBO 4-Channel Analog Data Logger | Onset's HOBO and InTemp Data Loggers. (n.d.). https://www.onsetcomp.com/products/data-loggers/ux120-006m#specifications
- 6. Li-Cor Environmental. Contact LI-COR Biosciences. (n.d.). https://www.licor.com/env/
- 7. Reinhart, C. F., Mardaljevic, J., & Rogers, Z. (2006). Dynamic Daylight Performance Metrics for Sustainable Building Design. *LEUKOS*, *3*(1), https://doi.org/10.1582/leukos.2006.03.01.001
- 8. Nabil, A., & Mardaljevic, J. (2005). Useful daylight illuminance: A new paradigm for assessing daylight in buildings. *Lighting Research & Camp; Technology*, 37(1).

Questions?

Assessing Heat Generation in Anaerobic Composting of Yard and Food Waste for Backyard Renewable Energy Potential

<u>Teacher</u>

William Johnson - Flour Bluff High School (FBISD)

Faculty Advisors

Dr. Xiaoyu Liu - Research Project Faculty (TAMUK)
Dr. Marsha Sowell - Curriculum Development Faculty Mentor (TAMUK)

Student Mentor
Ravi Chandra Madasani (TAMUK)

Industrial Advisor Enrique Molina





research in <u>Renewable Energy Across</u>
<u>Disciplines (I-READ)</u>



OVERVIEW

- Composting and Facts on Yard and Food Waste
- Research and Results
- CurriculumModules andClassroomImplementation



Composting

Types

Aerobic Composting

- •Relies on oxygen and moisture
- Promotes heat generation
- •Generate energy from heat
- •Faster 6 weeks

Anaerobic Composting

- •Fueled by bacteria and moisture
- Produces end products like biogas
- •Generate energy from combustion
- •Slower 6 months







Composting

Types

Aerobic Composting

- Open Pile
- Windrow
- Static Pile
- In-Vessel
- Vermicomposting

Anaerobic Composting

- •Covered static compost heap or bin
- •Submersion or underwater composting
- Pits or trenches (Landfill)







YARD WASTE



The 3 Classifications Yard waste include:

- 1. Grass clippings
- 2. Small Yard Waste
 - Brush
 - Leaves
- 3. Heavy Brush
 - Tree trimmings greater than 3"in
 - Tree trunks
 - Root balls

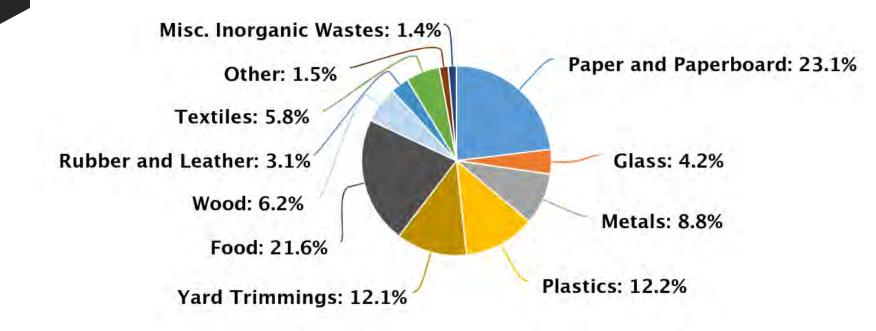






Total MSW Generated by Material, 2018

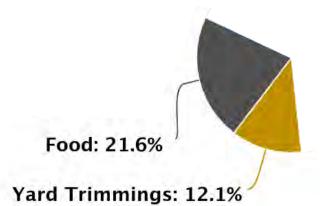
292.4 million tons



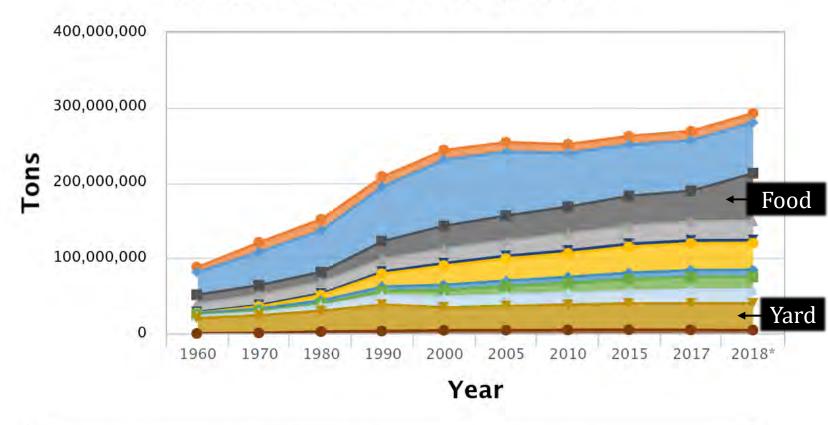
Total MSW Generated by Material, 2018

292.4 million tons

33.7%

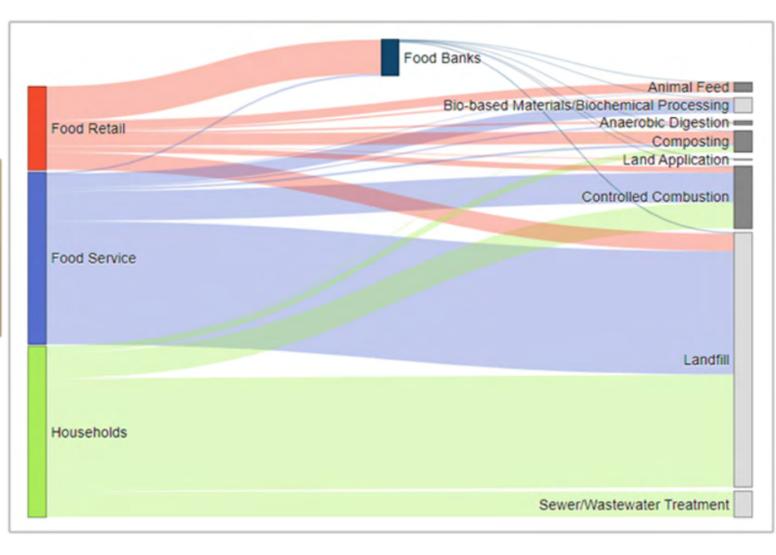


Generation Tonnages, 1960-2018









Assessing Heat Generation in Anaerobic Composting of Yard and Food Waste for Backyard Renewable Energy Potential

Present Focus

Future Focus

Generation

Capture

Utilization

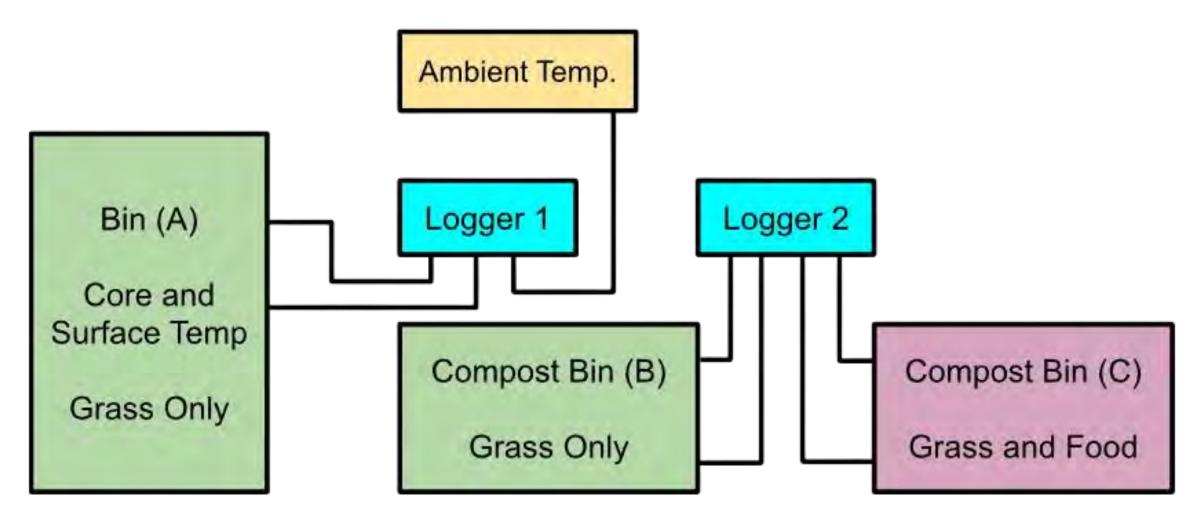






Research Objectives

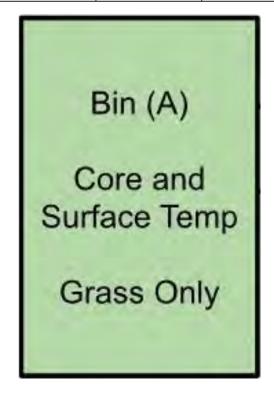
- 1. Investigate heat generation from anaerobic composting of yard and food waste.
- 2. Compare large bin core and surface level temperatures.
- 3. Compare core temperatures between large and small bins.
- 4. Compare core temperatures of small bins with grass-only versus grass/food.



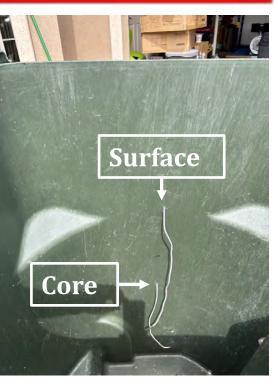
Trash Can (A) (mass in lbs)			
Load 1	Grass	14	
Load 2	Grass	21.4	
Load 3	Grass	25	
Load 4	Grass	24.2	
Load 5	Grass	26.2	
Total		110.8	

Approx. 115 gallons
Depth 108 cm

Surface Depth 13 cm Core Depth 54 cm Sealed shut to keep heat inside









Compost Bin B (mass in lbs)			
Load 1	Grass	25.2	
Load 2	Grass	17	
Total		42.2	

Approx. 43 gallons



Tested / Sealed



Compost Bin (B)
Grass Only

Sensors Location (2)



Compost Bin C (mass in lbs)			
Load 1	Grass	4.2	
Load 2	Food	13.8	
Load 3	Grass	5	
Load 4	Food	12.6	
Load 5	Grass	6.8	
Total		42.4	

Approx. 43 gallons



Sensors Location (2)



Tested / Sealed



Compost Bin (C)
Grass and Food



Ambient Temp.

Suspended in Air

Logger 1



- Bin A Core
- Bin A Surface
- **Ambient Temp.**

Logger Settings

Start Date: July 14, 2024 8:00pm

End Date: July 21, 2024 8:00pm

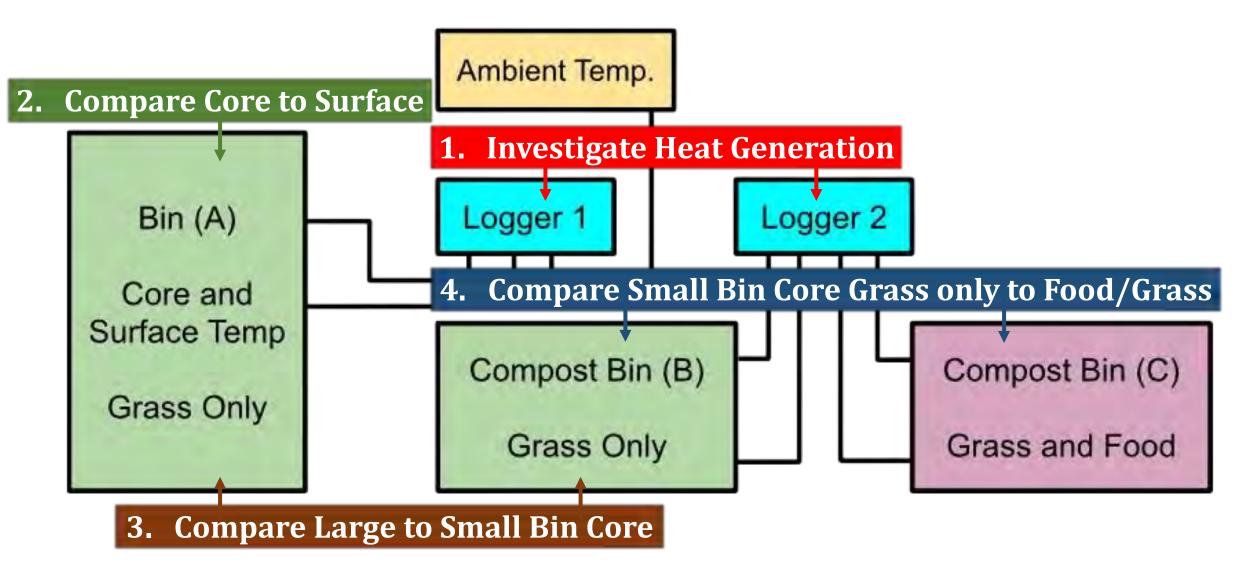
Data Point Intervals: Every 30 minutes

Logger 2

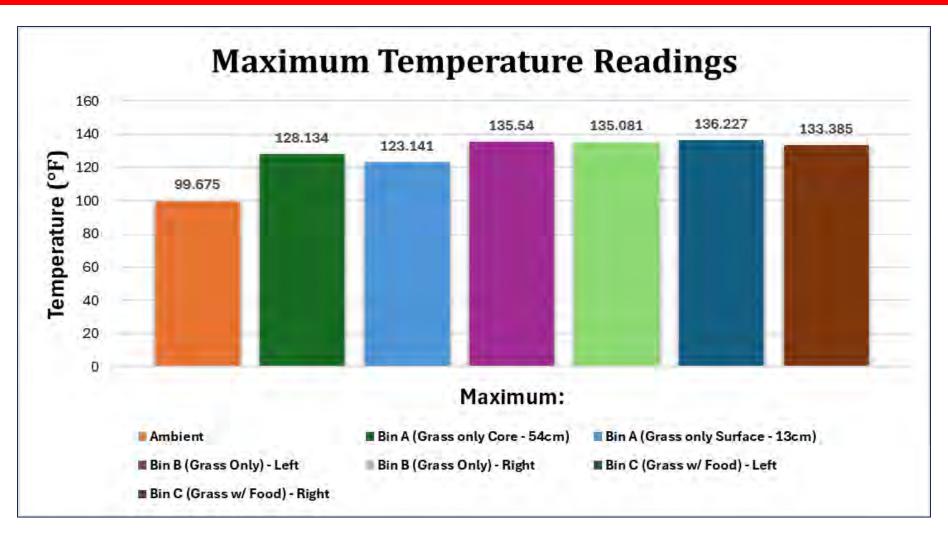
4 Sensor Ports use:

- Bin B Left
- Bin B Right
- Bin C Left
- Bin C Right

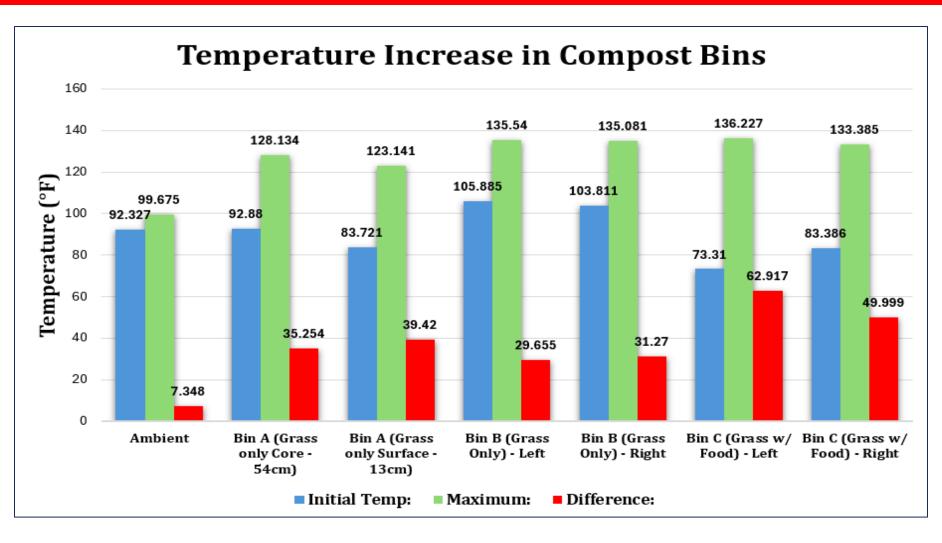




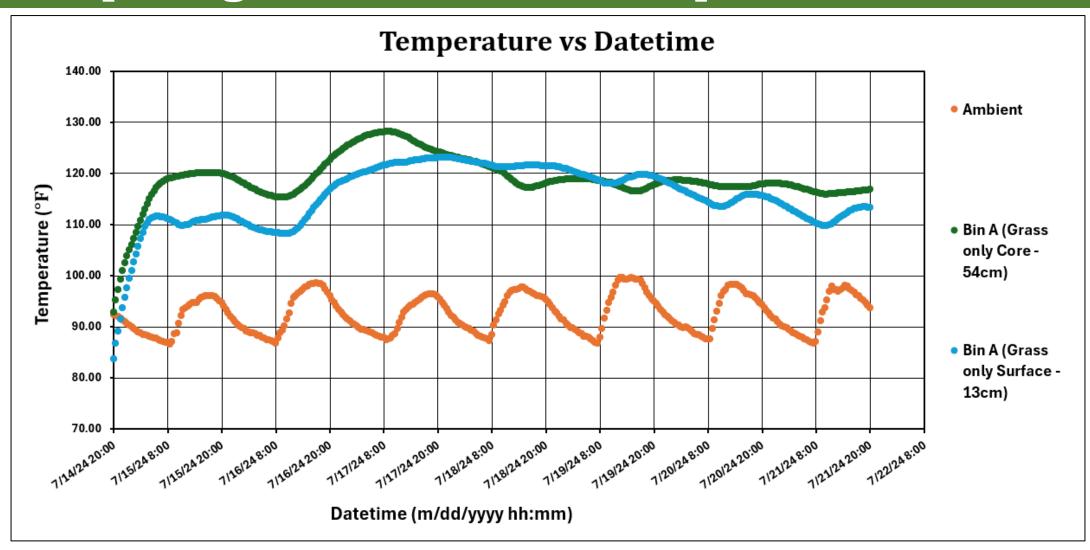
Investigate Heat Generation



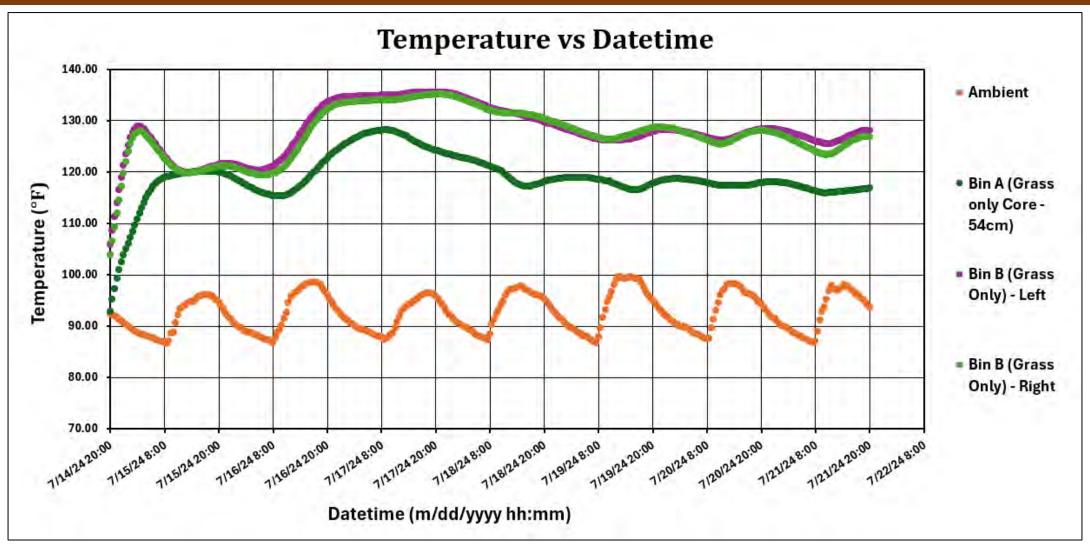
Investigate Heat Generation



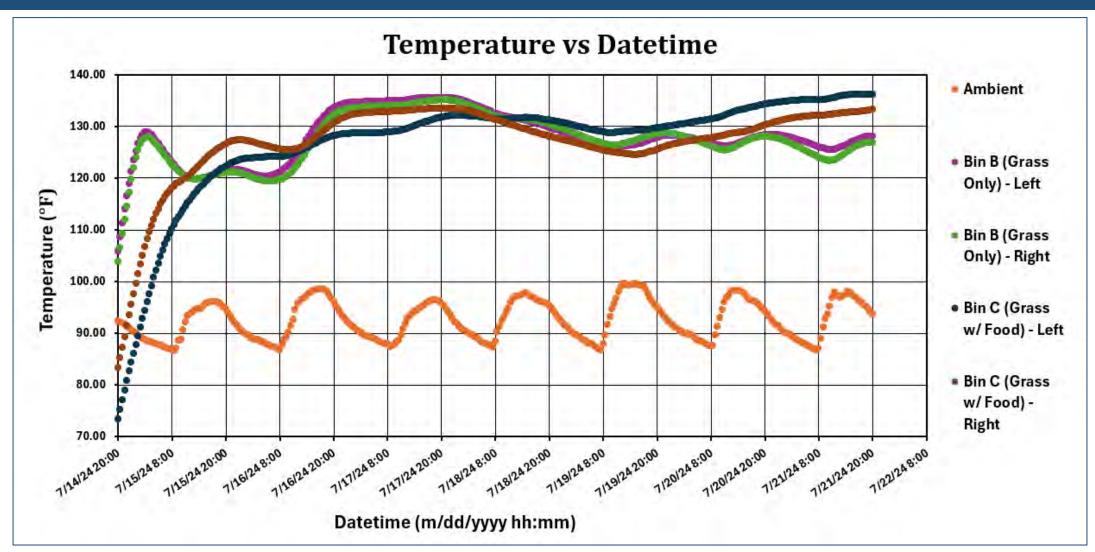
Comparing Core to Surface Temperatures in Bin A



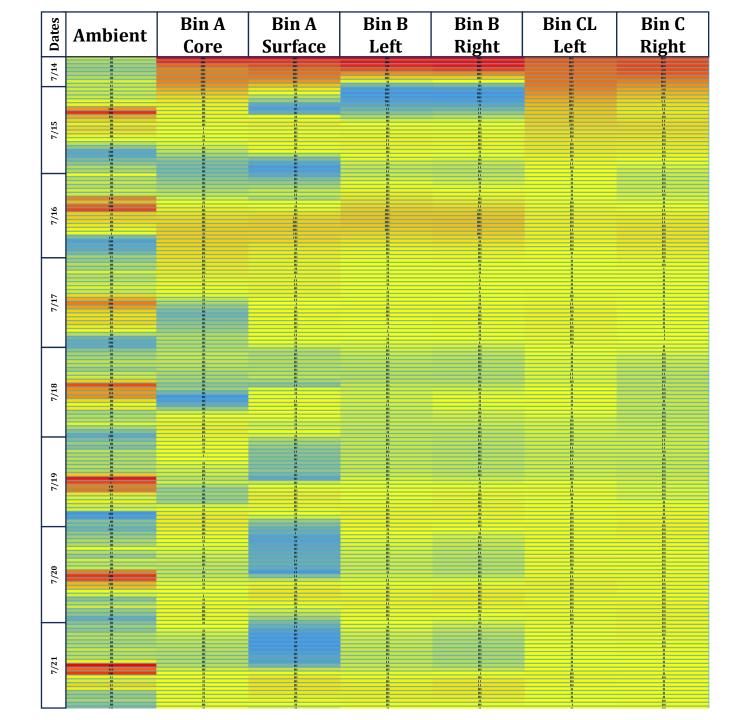
Comparing Large (A) to Small (B) Bin Core Temperatures



Comparing Small Bin (Grass Only to Food/Grass) Temperatures



Rate of **Change**of Temperature



Conclusions

- **❖** Investigate Heat Generation
 - ➤ Heat can be produced through the anaerobic composting
 - ➤ Evidenced by maximum temperatures in each bin exceeding the ambient temperature by 23°F to 36°F
- Comparing Large (A) to Small (B) Bin Core Temperatures
 - Bin B had higher initial rate of change
 - > Bin B reached a higher max temperature
 - Bin B maintained a higher average temperature

- **Comparing Core to Surface Temperatures in Bin A**
 - > Similar rates of change
 - > Core reached a higher max temperature
 - Core maintained a higher average temperature
- Comparing Small Bin (Grass Only to Food/Grass) Temperatures
 - Bin C had higher / longer sustained initial rate of change
 - ➤ Bin C reached a higher max temperature by the week's end
 - ➤ Bin B maintained a slightly higher average temperature (0.679 °F)

Conclusions

- **❖** Investigate Heat Generation
 - > Heat can be produced through the

Comparing Core to Surface Temperatures in Bin A

Evice each tem

Based on results from this experiment:

rature

- Compar Temper
 - > Bin
 - > Bin

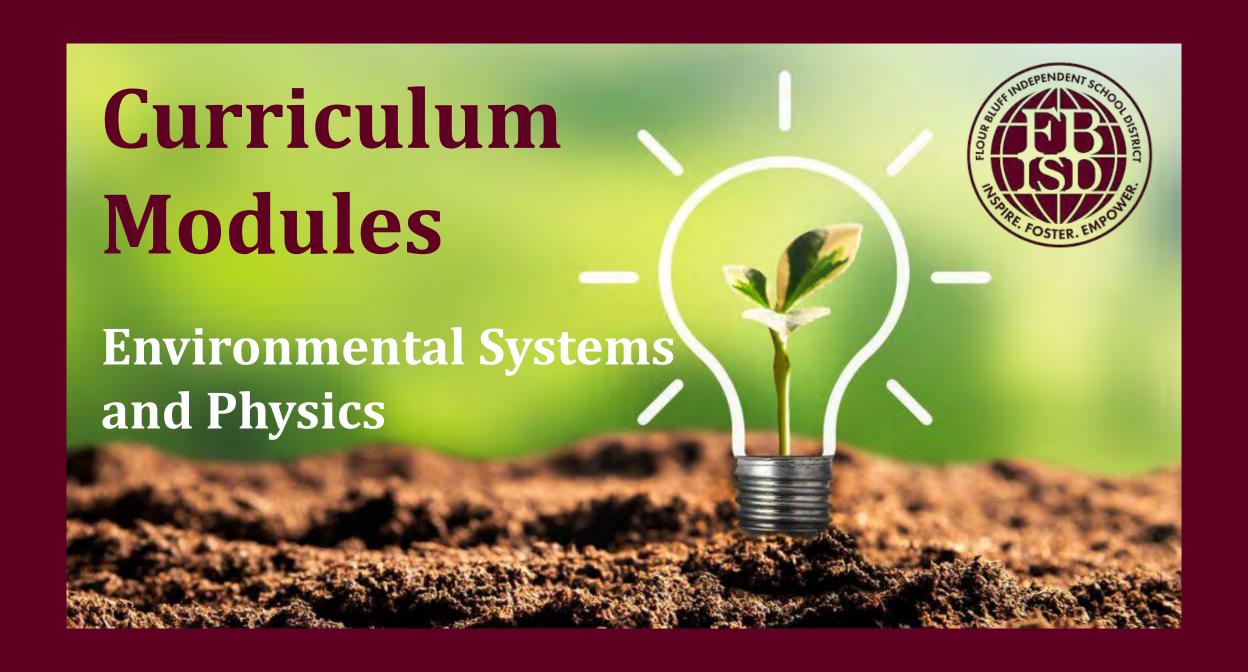
A smaller bin containing food and grass with temperature sensors located within the core will produce the most heat within a week's timeframe.

ed

erature

➤ Bin B maintained a higher average temperature

- by the week's end ► Bin B maintained a sli
- ➤ Bin B maintained a slightly higher average temperature (0.679 °F)



Environmental Systems

Composting in the Classroom

❖ Background Information and Vocabulary Exploration

> Students will be given the opportunity to research and explore the science behind composting.

Composting Project (Weeks 1 through 6)

- > Activities over several weeks will include:
 - Setting Up Composting Bins
 - Daily Monitoring and Rotation
 - Data Collection
 - Week-by-Week assessment the decomposition progress of each bin
 - Final Report and Presentation





CONDUCTOR

P-TYPE SEMICONDUCTOR

PELLETS

SEMICONDUCTOR

Physics

Thermoelectric Generator Investigation Activity

Four Day Activity

❖ Day 1: In-depth Background

Students will gain a comprehensive understanding of thermoelectric generators, the Seebeck effect, and related principles.

❖ Day 2: LED Module

> Students will use a thermoelectric generator to power an LED and understand the conversion of heat energy to electrical energy.

❖ Day 3: EMF Module

> Students will measure the EMF produced by the TEG and calculate the current using a known resistance.

❖ Day 4: Heat vs EMF Module

> Students will measure and graph the voltage output of the TEG as the heat input is varied.

Thank You!

Acknowledgement:

This material is based upon work supported by the National Science Foundation under Award No. 2206864. Any opinions, findings, and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the National Science Foundation.



References





https://www.epa.gov/facts-and-figures-about-materials-waste-and-recycling/food-material-specific-data



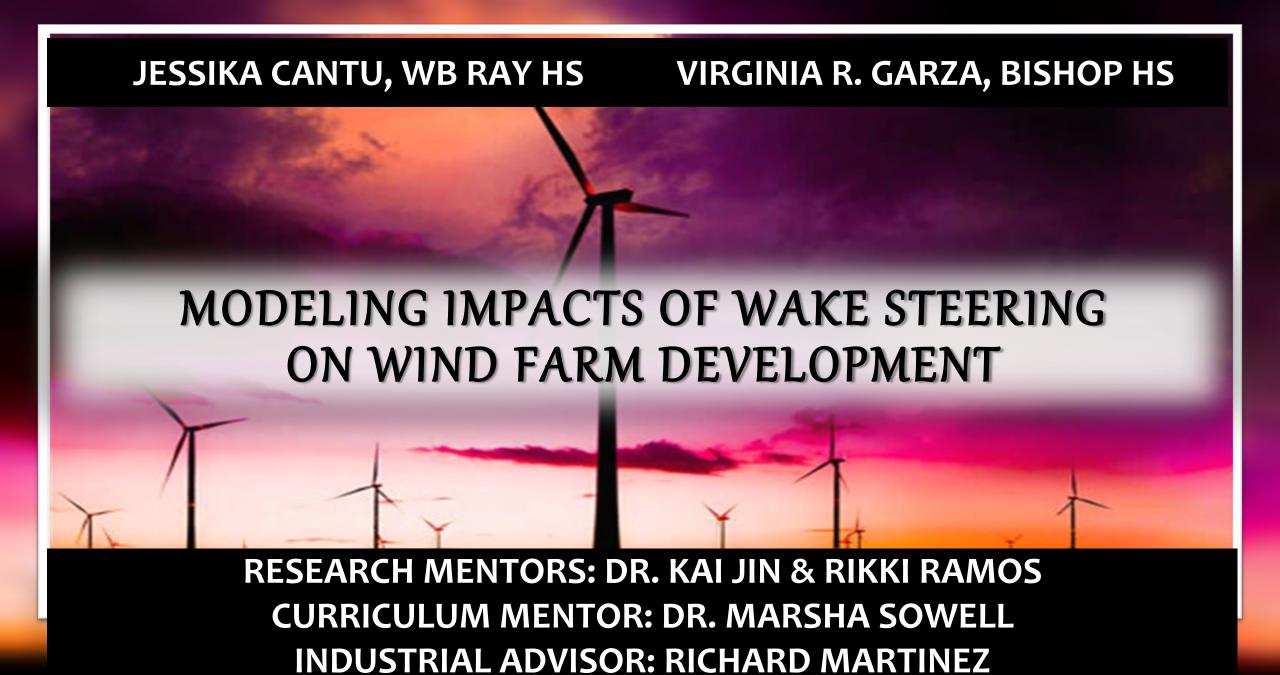
https://www.epa.gov/facts-and-figures-about-materials-waste-and-recycling/yard-trimmings-material-specific-data

Yard waste facts

https://www.cctexas.com/services/utilities/storm-water-pollution-prevention/disposing-yard-waste https://www.nctcog.org/envir/watershed-management/stormwater/yard-waste



https://www.epa.gov/land-research/farm-kitchenenvironmental-impacts-us-food-waste



INNOVATIVE SOLUTIONS:

- Wake Steering Techniques:
- Potential to increase energy output by up to 13%.
- Builds on previous research to optimize performance.

emciency.

WIND FARM

efficiency. BINE

position of formance of

ed to 2035.

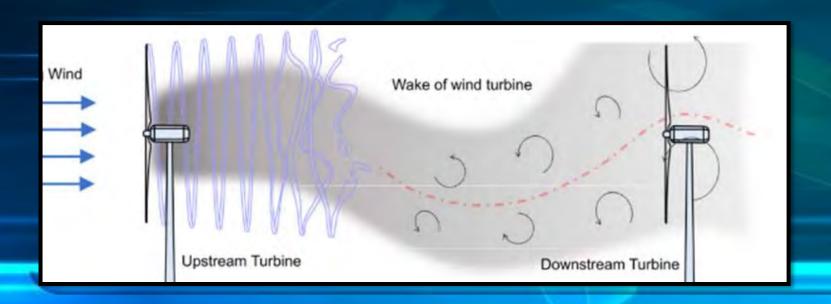
WIND TURBINE WAKE EFFECT

Definition:

- Disturbance in airflow caused by a wind turbine.
- Creates a downstream area with reduced wind speed and increased turbulence.

Impacts:

- Reduces efficiency and power output of downstream turbines.
- Similar to how a boat leaves a wake in the water behind it.

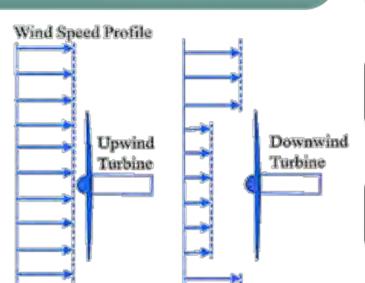




WIND TURBINE WAKE LOSSES

Definition:

 Reduction in energy production caused by the wake effect.



Mechanism:

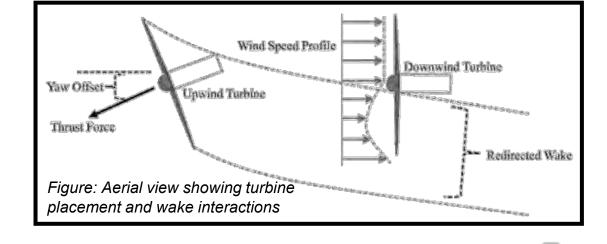
Impact:

Mitigation Strategies:

Importance:

- Wind turbines create a wake of slower, turbulent air behind them.
- This wake affects downstream turbines, reducing their efficiency and power output.
- Downstream turbines generate less power in disturbed wind conditions.
- Significant reduction in overall energy output of the wind farm.
- Strategic turbine spacing and alignment with prevailing winds.
- Proper turbine placement to ensure optimal wind conditions for each turbine.
- Maximizes power production.
- Enhances economic viability of the wind farm.

Wake Steering



Objective:

Optimize wind farm layout and increase power production

Key Strategy:

 Redirect wakes of upstream turbines away from downstream turbines using yaw angle adjustment.

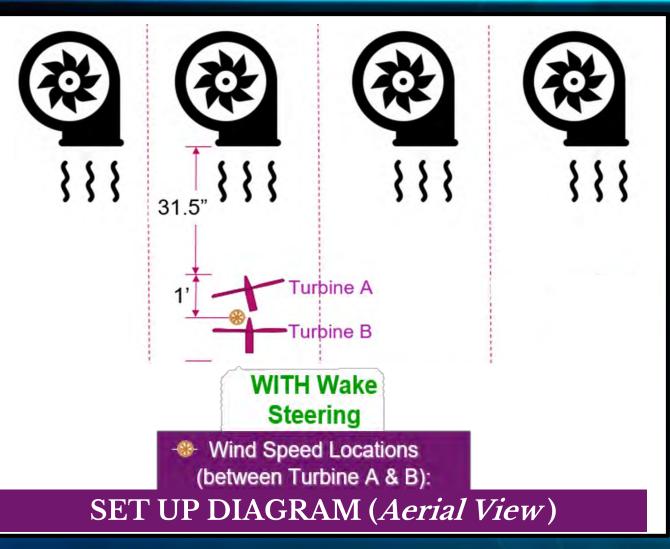
Benefits:

- Increases power production for downstream turbines.
- Enhances overall wind farm efficiency and economic potential.
- Provides design flexibility, especially in land-constrained locations.

Evidence:

Proven success through field trials and simulation studies.

Experiment Set Up



Baseline Measurements:

- Set up four industrial blowers.
- Model is in a controlled environment.
- Measure initial wind speeds at various distances using

Single Turbine Measurements:

- Place the first turbine 31.5 inches from a blower.
- Position a second turbine at incremental distances of 1 ft. behind the first, up to 10 feet.

Wake Steering Tests:

- Set first turbine to wake steering.
- Repeat measurements for the second turbine.



WAKE STEERING Model In Action

Mitigation Strategies:

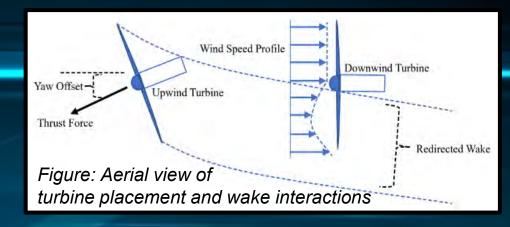
Importance:

Strategic turbine positioning.

Reduces costs.

Optimal wind farm design to reduce wake interference.

Advances sustainable wind energy solutions.



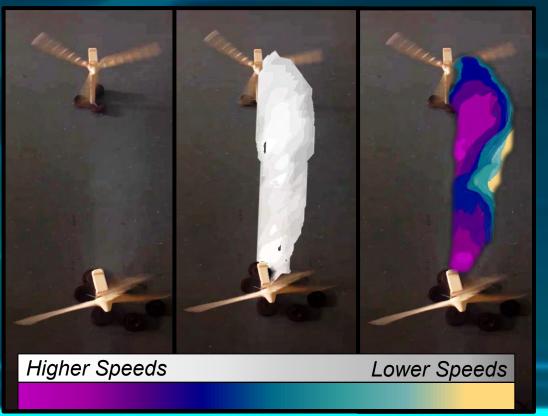
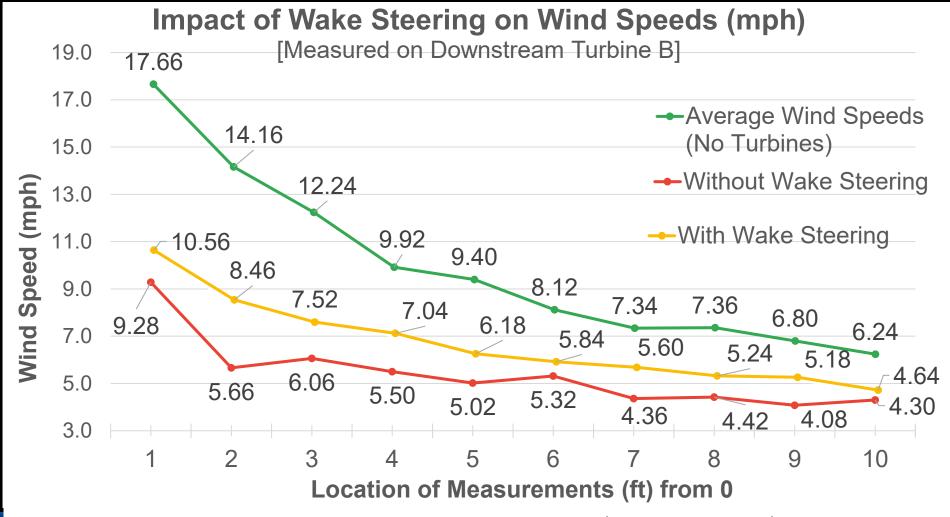


Figure: Aerial view of wake steering on Turbine A and Turbine B downwind with no wake steering.

WIND SPEEDS

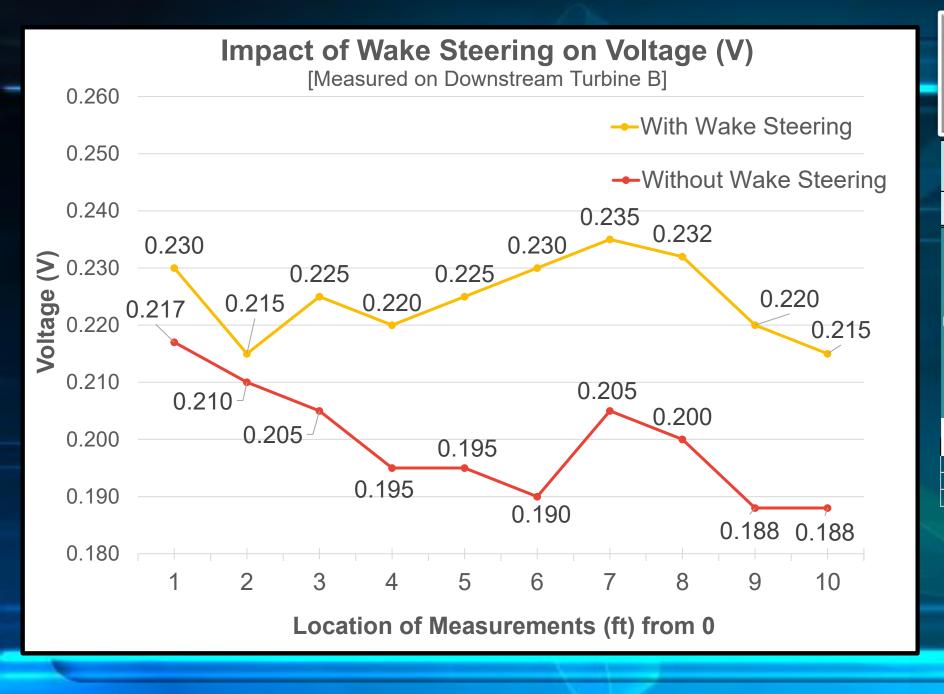




Green line represents the baseline wind speed without any turbines present. (No wind Interference)

Red line represents the wind speeds downstream from turbine A without wake steering efforts.

Yellow line represents the wind speeds downstream from turbine A with wake steering efforts. There is a clear increase in wind speeds downstream from turbine A with wake steering.



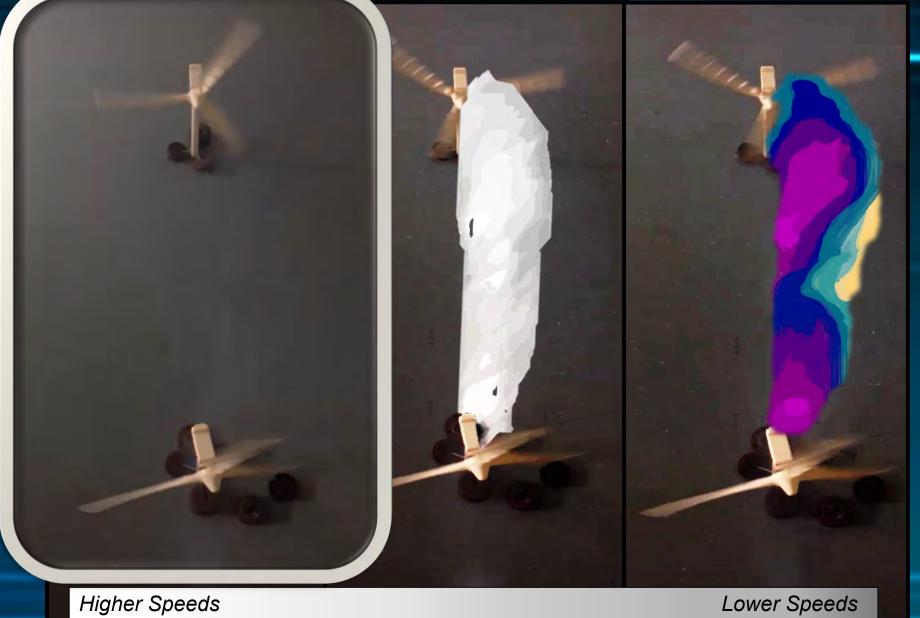
Voltage Data

Turbine Layout:		A B	A / B	Difference (%)
Wake Steering:		No	Yes	
	X ₀	0.250	0.210	-16.00%
	1	0.217	0.230	5.99%
	2	0.210	0.215	2.38%
	3	0.205	0.225	9.76%
Location of	4	0.195	0.220	12.82%
Measurements	5	0.195	0.225	15.38%
(ft from X ₀)	6	0.190	0.230	21.05%
	7	0.205	0.235	14.63%
	8	0.200	0.232	16.00%
	9	0.188	0.220	17.02%
	10	0.188	0.215	14.36%
			Increase %:	5%

 X_0 - includes Turbine A only. All other measurements are taken



Video Data



Curriculum Modules

 Observe wake steering effects on turbine models and analyze physics properties to optimize wind farm layouts, explaining how wind speed and generated energy vary by turbine location.

PB1: Physics in Wind Turbine Analysis Identify problems and apply solutions using proper tools and models.

B4B: AP Biology -Engineering Practices

A2B:
Statistics Analyzing
Functions in
Wind Turbine
Data

G10A:
Geometry Trigonometric
Ratios in
Right
Triangles

 Use trigonometric ratios to calculate yaw offset angles, at which wind strikes turbines and analyze resulting wake effects.

 Identify and interpret relationships in wind speed and turbine output data, record and analyze data.

Wake Steering: Conclusions and Results

Achieved:

Optimized wind farm layout, increased power production.

Effective Strategy:

Redirecting wakes with yaw angle adjustment.

Key Results:

- Higher power output for downstream turbines.
- Improved overall wind farm efficiency and economic potential.
- Enhanced design flexibility for land-constrained locations.

Validation:

Success demonstrated in field trials and simulations.

FUTURE STUDIES

Determining the effects of wake steering on column C and column D.

Graphing the effects of wind steering on plant growth.

Investigating the longterm impacts of wake steering on turbine longevity and maintenance. Exploring the economic implications of large-scale implementation of wake steering in wind farms.

ACKNOWLEDGEMENTS

This material is based upon work supported by the National Science Foundation under Award No. 2206864. Any opinions, findings, and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the National Science Foundation.

REFERENCES

- Balakrishnan, R. K. (2022). Maximization of the power production of an offshore wind farm.
- Howland, M. F. (2019). Wind Farm Power Optimization Through Wake Steering.
- Johnson, W. The Impact of Wind Farm Turbine Coordination for Power Production Optimization
- Kanev, S. (2019). Dynamic Wake Steering and Its Impact on Wind Farm Power Production
- Office of Energy Efficiency & Renewable Energy. (2022). Wind Turbines: The Bigger, the Better.
- Simley, E.(2024). The Value of Wake Steering Wind Farm Flow Control in US Energy Markets.
- Tri, B. (2023, August 14). Laying the foundation for wind turbines now and in the future.

Questions?