Fighting Earthquakes with Pendulums: the Tuned Mass Damper

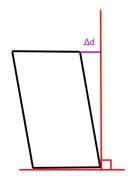
RESEARCH QUESTION: HOW DOES THE MASS OF A TUNED MASS DAMPER AFFECT THE DISSIPATION OF MECHANICAL ENERGY IN AN OSCILLATING STRUCTURE?

 $Physics\ Internal\ Assessment$

1 Introduction

19th of September, 1985: Mexico City was shaken to the ground by an 8.0 Mw earthquake, buildings collapsing onto citizens totalling 5,000 deaths. 19th of September, 2017: Mexico City was hit again by a 7.1 Mw earthquake; however, despite more skyscrapers towering over the city skyline than in 1985, they did not collapse (Vance 2017). As the ground was shaking beneath my feet, my safety and that of countless Mexicans trapped in swaying towers were rescued by the physics of damping mechanisms. One such damping mechanism is called a Tuned Mass Damper, which is a "system for damping the amplitude in one oscillator by coupling it to a second [damped] oscillator" causing energy from the earthquake to be "transferred' to the second oscillator" (Orloff 2012). Given their rarity in industrial skyscrapers, I wanted to investigate the effectiveness of a pendulum TMD in combating earthquakes like the ones I experienced in Mexico. However, I discovered that architects often chose lighter dampers to save costs, leading to very little research on what the mass of a TMD should be to most effectively dampen oscillation. Hence, I decided to investigate the effect of mass on the dampening of the structure by asking the research question: How does the mass of a Tuned Mass Damper affect the dissipation of mechanical energy in an oscillating structure?

2 Theoretical Background



This investigation consists of three components: the earthquake, the structure, and the pendulum Tuned Mass Damper.

2.1 The Earthquake

An earthquake repeatedly displaces the base of a structure ("How Do..."). Changing points of reference, it displaces the top of the structure by Δd (meters) as seen in Figure 1. Simulating this effect, I will simplify the impact of an earthquake on a structure as a single initial displacement Δd_1 , after which no external force will be applied.

Figure 1: Initial displacement

2.2 The structure

Internal reinforcement of the structure applies a restoring force, F_s (Newtons), to bring the structure back to vertical equilibrium, a force denoted by $F_s \propto -\Delta d$. A force opposite and proportional to displacement oscillates the structure around its equilibrium, denoted by the vertical red line in Figure 2, resulting in simple harmonic motion.

2.3 The Dampening

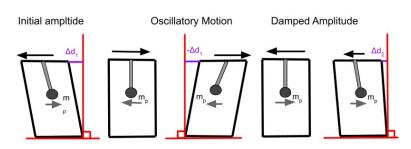


Figure 2: Animation frame of damping oscillation

As the structure oscillates, some of its mechanical energy, ME (Joules), is transferred into other forms of energy (eg. thermal energy) causing each successive amplitude (meters) of the structure's oscillation to be smaller than the previous one, known as damping as visualized in Figure 2. Now consider a pendulum at the top of the structure, the TMD. When it oscillates due to the displacement of the top of

the structure, it transfers energy from the structure into its own oscillation (Connor 2002). This occurs because the pendulum absorbs some of the structure's mechanical energy in order to oscillate itself.

Research Question: How does the mass of a Tuned Mass Damper affect the dissipation of mechanical energy in an oscillating structure?

Due to the additional damping of the TMD (caused by friction in the rotating hinge of the pendulum), it further dampens the structure's oscillation, meaning that the consecutive maximum displacement of the structure, Δd_2 (meters) will be lower than Δd_1 (meters). Similarly, the mechanical energy of the structure during the second period of oscillation will be smaller than the initial mechanical energy, $ME_2 < ME_1$, because of the dissipation of mechanical energy caused by the TMD. The mass of the pendulum, m_p (grams) will be modified to measure the effect on damping between ME_1 and ME_2 .

3 Deriving The Formula

To measure the total mechanical energy in the oscillating structure, I can use $ME = \frac{1}{2}k\Delta d^2 + \frac{1}{2}mv_2$ (Joules), the sum of its kinetic and potential energy, with some unknown constant k, mass m, and velocity v. However, at maximum displacement Δd_1 and Δd_2 , there is no kinetic energy, simplifying the equation to $ME = \frac{1}{2}k\Delta d^2$. To measure the percentage loss of mechanical energy due to damping, we can use the equation

$$\frac{ME_1 - ME_2}{ME_1} \times 100\%$$

which cancels out the unknown constant $\frac{1}{2}k$ resulting in the neat formula $\zeta = \frac{\Delta d_1^2 - \Delta d_2^2}{\Delta d_1^2} \times 100\%$ where ζ (%) is the percentage of lost mechanical energy. Δd_1 is a constant variable, equal to the initial displacement of the structure as a result of the earthquake. Δd_2 is the maximum displacement of the structure during its second period of oscillation, which allows me to measure the immediate loss in mechanical energy between the first two consecutive periods of oscillation. I can not measure Δd_2 in meters because it oscillates too fast for a human to measure, which is why I will use an accelerometer. The relationship between acceleration and displacement is $a = \omega^2 \Delta d$ where $\omega = 2\pi f$, where f is a constant, representing the frequency of the building's oscillation. I can therefore use

$$\Delta d_2 = \frac{a_2}{4\pi^2 f^2}$$

where a_2 is the maximum acceleration of the structure in the second period of oscillation. This results in the final equation for ζ (Equation 1, % of ME lost between first two consecutive periods) as:

$$\zeta = \frac{\Delta d_1^2 - (\frac{a_2}{4\pi^2 f^2})^2}{\Delta d_1^2} \times 100\% \tag{1}$$

4 Hypothesis

I hypothesise that the percentage loss in mechanical energy ζ (%) increases linearly as m_p increases because the larger mass of the Tuned Mass Damper can absorb more kinetic energy, since kinetic energy is dependent on mass, as seen in the formula $KE_p = \frac{1}{2}m_pv^2$. Hence, this could achieve the relationship $\zeta \propto m_p$ where the mass is proportional to the percentage of mechanical energy dissipated.

5 Setting up the Experiment

I had to build the structure using only materials in my garage while maintaining the rigidity-flexibility and height-width ratio of a real skyscraper. I approximated this by attaching plastic straws to each other with toothpicks totalling 30cm high. They are attached to 5cmx5cm wooden planks. For additional support it is screwed to a larger wooden platform, which reaches 5.00cm out from each side of the base of the structure. This allows the structure to oscillate while maintaining structural integrity. The platform will be placed next to a wall, so to displace by $\Delta d_1 = 5.00cm$ I will simply move the top of the structure until it touches the wall, separated by 5.00cm due to the platform size. Consider Figure 3 for reference.

Research Question: How does the mass of a Tuned Mass Damper affect the dissipation of mechanical energy in an oscillating structure?

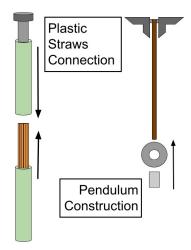


Figure 3: Assembling structure and TMD

For the TMD, I constructed a metal pendulum made out of old wire hanging on a screw tightened together with 90° metal mantles to cause friction (hence damping) when the TMD swings (Lourenco 2011). To modify the mass m_p , I will place standard 5g circular metal nuts into the wire, then fix them in place using a plastic slider that tightens around the wire. Finally, I had to find the constant value for the natural frequency of the swaying structure (f) to match it with the frequency of the pendulum so that both oscillate in the same time period, thus creating damped simple harmonic motion. To do this, I offset the structure by 5.00cm, released, and measured the period, $T = (0.700 \pm 0.005)s$, then solved for frequency, $f = \frac{1}{0.700s} = (1.43 \pm 0.01)Hz$. To synchronise the frequency of the TMD with f, I modified the length of the pendulum using $L = \frac{g}{4\pi^2 f^2}$, derived from the period of a pendulum.

$$\frac{1}{f} = 2\pi \sqrt{\frac{L}{g}} \rightarrow L = (0.122 \pm 0.002)m$$

where g is the gravitational constant $(9.81\frac{m}{s^2})$. Hence, the TMD will oscillate at the same frequency as the structure but in the other direction, thereby tuning the mass damper (Connor 2002). Lastly, I will attach a Raspberry Pi microcomputer to the top of the structure where I will measure acceleration using the SenseHat accelerometer sensor, where I will program a script to collect and store the data.

5.1 Variables

Independent variable: Mass of the Tuned Mass Damper (m_p) . Modified with metal nuts. Increments over: 5, 10, 15, 20, 25, 30, 35, and 40 grams.

Dependent variable: Second crust of structure's acceleration (maximum acceleration in second oscillation) (a_2) .

Controlled Variables:

Initial displacement Δd_1 of the top of the structure. Must stay constant to simulate constant magnitude of earthquake in each trial. If inconsistent, initial acceleration will vary, resulting in different a_2 values throughout the trials. I will control this by releasing the structure from the same distance of 5.00cm (measured by placing top of structure parallel to wall). Moreover, I can anticipate minor inaccuracies because they will average out over the multiple trials.

<u>Structure's material and dimensions.</u> The structure itself should stay durable by not bending the straws nor damaging its structural integrity. A change in structural integrity could cause additional damping which would skew the damping effect, which would decrease a_2 irrespective of m_p . I can

anticipate this problem by monitoring the consistency of a_2 throughout the trials. I will control this variable using durable materials and not overloading the TMD (max weight 40g).

5.2 Materials

Material	Properties
8x Metal nuts	Weighing 5g each
8x plastic straws	Measuring 15cm each
32x wooden toothpicks	4x to attach straws together
1x copper wire	Rigid; $(12.2 \pm 0.2)cm$ long
10x metal screws	4x for base, 4x for ceiling, 2x for pendulum,
	1x for fixing base to platform
3x synthetic wood planks	2x 5cm by 5cm for base and ceiling of structure
	1x 15cm by 15cm for platform
2x metal handle for pendulum	Right-angled piece, screwed to ceiling, with holes
	for screw to go through, copper wire around it
1x adjustable slider	plastic slider to fix weights in place on pendulum
Raspberry Pi SenseHAT Accelerometer wirelessly connected to laptop	Sensor attached to top of structure $(\pm 0.001 ms^{-2})$

6 Method

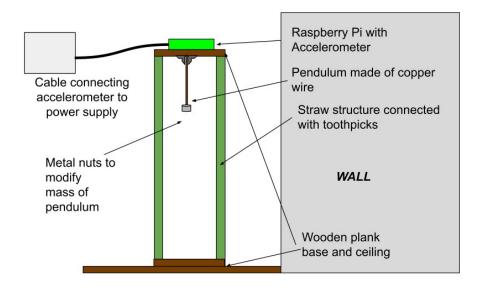


Figure 4: Experiment Setup

- 1. Ensure that the structure dimensions are not altered by checking if the structure is visually deformed or bent. Tighten the straws together if loose to increase consistency and reliability of data. If the straws bend or the structure is leaning to one side, then replace with new straws.
- 2. Connect to Raspberry Pi through wireless Secure Shell Connection from laptop through terminal (Ylonen 1996). Run custom accelerometer reader program to begin gathering raw data (see in Appendix B).
- 3. Displace the top of the structure until it reaches the wall, which by design of the platform is exactly 5.00cm away, removing likely human error possible when using ruler to displace thus

- maximizing accuracy of results. Ensure that it is fully in contact with wall before releasing to maintain Δd_1 precise and accurate for control variables to be consistent across trials.
- 4. Release the structure, avoid applying an additional force in the form of a push by letting go of structure gently and quickly. This maximizes accuracy in the data because the initial acceleration is constant across trials.
- 5. After the structure oscillates back and forth approximately twice, stop recording program, download data to computer from Raspberry Pi microcomputer via connection, then disconnect.
- 6. Record the measure of the amplitude in the acceleration graph $(a_2 \text{ in } ms^{-2})$ during its second period of oscillation displayed using my custom Python program using data visualization libraries (detailed later, consider Appendix C). Use highest recorded acceleration during second period of oscillation in order to minimize human error when reading the data thus maximizing accuracy.
- 7. Repeat steps 1 to 6 for trial 2, 3, 4, 5, 6, 7.
- 8. Detach slider from pendulum, attach another metal nut to the bottom of the pendulum to increment the mass through the values 5, 10, 15, 20, 25, 30, 35, and 40 grams. Tighten the slider to ensure they do not fall once fixed in place.
- 9. Repeat steps 1-8 for each increment in step 8.

7 Risk assessment and ethical concerns

Ethical and Safety Concerns This experiment is free of ethical concerns because it does not involve other beings and is free of safety concerns as it is not dangerous for any parties involved.

Environmental Concerns All materials are reusable except for the plastic straws which will be properly recycled when disposed of, implying limited environmental concerns. Therefore, this experiment has no significant risk or ethical concerns associated with it and carefully approaches any environmental concerns.

8 Data and Analysis

	Acceleration at Second Crest of Acceleration Graph ms^{-2} $\Delta ms^{-2} = \pm 0.001$							
Mass (m_p) g $\Delta g = \pm 0.01$	Trial 1	Trial 2	Trial 3	Trial 4	Trial 5	Trial 6	Trial 7	Mean Acceleration (a_2) ms^{-2}
5.00	2.013	2.019	2.022	2.015	2.023	2.021	2.054	2.019 ± 0.005
10.00	1.856	1.851	1.862	1.854	1.861	1.865	1.858	1.858 ± 0.007
15.00	1.733	1.735	1.737	1.739	1.730	1.741	1.729	1.735 ± 0.006
20.00	1.632	1.631	1.630	1.634	1.625	1.634	1.642	1.633 ± 0.008
25.00	1.529	1.530	1.532	1.520	1.519	1.531	1.533	1.528 ± 0.007
30.00	1.450	1.453	1.444	1.439	1.437	1.439	1.437	1.443 ± 0.008
35.00	1.401	1.412	1.396	1.398	1.395	1.409	1.404	1.402 ± 0.008
40.00	1.385	1.381	1.369	1.384	1.376	1.380	1.382	1.380 ± 0.008

Table 1: Raw data showing mass(g) and acceleration(ms^{-2})

Collected data (in Table 1) is gathered to four significant figures with uncertainty $\pm 0.001 ms^{-2}$. The mean acceleration is calculated in the rightmost column by adding all trials and dividing by 7, with uncertainty calculated using $\frac{(Range)}{2}$. The calculated uncertainty is larger than the device's uncertainty, which reflects human error caused by minor variations in control variables, although it is low as it remains at three decimal places and is averaged out throughout the 7 trials for increased precision.

However, trial 7 of the first increment highlighted in bold, 2.054, is an **outlier** as it falls above the upper limit of the interquartile range,

$$1.5 \times IQR + Q_3 = 1.5 \times 0.008 + 2.023 = 2.035$$

Therefore, it will not be used in any calculations as to maintain high precision throughout the analysis. This could have been caused by an additional force placed onto the structure when releasing it from Δd_1 . There are no other outliers in the data.

Figure 5 is an example of Trial 2, $m_p = 5.00g$, of the raw data measured using the Raspberry Pi SenseHAT, visualised using the MatPlotLib and Numpy packages in a custom Python script I made displaying raw data gathered from the accelerometer (Hunter 2007; Harris et al. 2020; Van Rossum and Drake 2009). The release of the structure is shown in the initial spike, after which follows dampened SHM motion. a_2 is gathered by taking the value at the crest of the second period of oscillation, which corresponds with the maximum value at the second peak on the acceleration graph (example: a_2 occurs at t = 1.49 with acceleration $2.019ms^{-2}$. The pattern produced resembles damped simple harmonic motion, which further validates that the experiment was set up properly to simulate an earthquake's effect on a building.

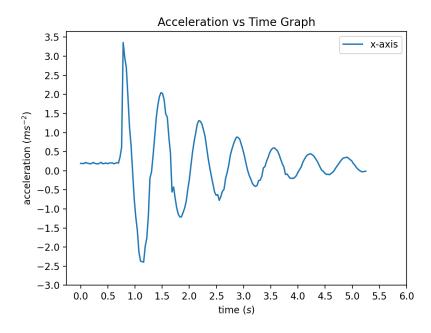


Figure 5: Trial 2 Raw Accelerometer Data Plotted Using Matplotlib script

With the a_2 values retrieved from the acceleration versus time graph of each trial and increment, the data can now be processed from acceleration to % of dissipated mechanical energy.

8.1 Processing Data

		Р	ercentage	e of Dissi _l	pated Med	chanical I	Energy ζ		
$\operatorname{Mass}(m_p)$ g $\Delta g = \pm 0.01$	Trial 1	Trial 2	Trial 3	Trial 4	Trial 5	Trial 6	Trial 7	ζ Uncertainty $\Delta\%$	$\begin{array}{c} \text{Mean } \zeta \\ \% \end{array}$
5.00	75	75	75	75	75	75		±1	75 ± 1
10.00	79	79	79	79	79	79	79	±1	79 ± 1
15.00	82	82	82	81	82	81	82	±1	82 ± 1
20.00	83.7	83.7	83.7	83.6	83.8	83.6	83.5	± 0.9	83.6 ± 0.9
25.00	85.7	85.6	85.6	85.8	85.8	85.6	85.6	± 0.8	85.7 ± 0.8
30.00	87.1	87.0	87.2	87.3	87.3	87.3	87.3	± 0.7	87.2 ± 0.7
35.00	88.0	87.8	88.0	88.0	88.1	87.8	87.9	± 0.7	87.9 ± 0.7
40.00	88.2	88.3	88.5	88.2	88.4	88.3	88.3	± 0.7	88.3 ± 0.7

Table 2: Processed data showing mass(g) and percentage of dissipated mechanical energy(%)

The Percentage of Dissipated Mechanical Energy is calculated using $\zeta = \frac{\Delta d_1^2 - (\frac{a_2}{4\pi^2 f^2})^2}{\Delta d_1^2} \times 100\%$ and then formatted to the decimal places of its uncertainty, seen in Table 2. For example, trial 1 at $m_p = 5.00g$ is calculated using:

$$\zeta = \frac{0.05^2 - (\frac{2.013}{4\pi^2 \times 1.43^2})^2}{0.05^2} \times 100\% = 75.\underline{13} = 75\% \text{ (no decimal place)}$$

To calculate uncertainty of ζ , the following formula is derived from uncertainty rules applied to ζ , where a_2 is the mean acceleration (raw data). To save space, the derivation can be seen in Appendix A, but is simply the uncertainty rules applied to ζ . However, the TMD's frequency varied between recorded trials. This inconsistency adds to the uncertainty of the processed data because the equation for ζ assumes a constant frequency f. The values varied between 1.41Hz and 1.45Hz, therefore the uncertainty of the frequency will be modified to $\Delta f = \frac{(Range)}{2} = \frac{1.45-1.41}{2} = 0.02Hz$. Now, the uncertainty can be calculated:

$$f \pm \Delta f = (1.43 \pm 0.02)Hz, \quad \Delta a_2 = \pm 0.001ms^{-2}, \quad \Delta d_1 = 0.05m$$

$$\therefore \Delta \zeta = (\frac{100}{16\pi^4 \Delta d_1^2}) \times (\frac{2a_2 \Delta a_2}{f^4} + \frac{4a_2^2 \Delta f}{f^5})$$

Then, to calculate the uncertainty of the same example data point:

$$\Delta \zeta = (\frac{100}{16\pi^4 0.05^2}) \times (\frac{2 \times 2.013 \times 0.001}{1.43^4} + \frac{4 \times 2.013^2 \times 0.02}{1.43^5}) = 1.4\% = 1\% \text{ (one significant figure)}$$

The uncertainty decreases over higher values of m_p since acceleration decreases, although it is much larger than in the raw data since acceleration is squared and the uncertainty of the frequency is considered. Consequently, the first three increments are formatted to zero decimal places to account for the larger uncertainty, whereas the later trials are formatted to one decimal place because of lower uncertainty. While it may seem like precision is lost because of one less significant figure, the differences across means are significantly larger in the first three increments which accounts for the increase in uncertainty. Therefore, precision is still maintained. Additionally, the uncertainty turns out to equal the same value across each increment, hence it is a column in the data table to improve readability. The mean percentage of dissipated mechanical energy is calculated by averaging the trials, however, since uncertainty is constant throughout each trial, the same uncertainty will be used in the mean. This ensures that accuracy is not artificially increased by using the mean of the data.

Moreover, standard deviation of the processed data yields $\sigma = 0.1\%$ or less for each increment of m_p (hence it is not shown in a data table to avoid repetition). This value is very low, thus further validating the precision of the processed data. This data can now be graphed to understand its trend.

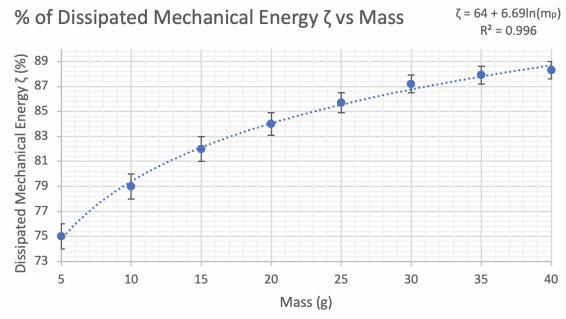


Figure 6: Graph of % of Dissipated Mechanical Energy vs Mass

The processed data follows a strong logarithmic relationship, as it takes an exponentially increasing amount of mass to dissipate larger percentages of mechanical energy. The logarithmic correlation relates $\zeta = k + A \ln m_p$, where k and A are constants. Using software, the line of best fit is

$$\zeta = 64 + 6.69 \ln m_p$$

with a high coefficient of determination of $R^2 = 0.996$, which indicates that the line of best fit accurately resembles the data. Moreover, the line of best fit passes through every error bar, which reflects that the data is precise. This would not have been the case if the processed data uncertainty would have been calculated using the mean of the trials, which reflects how accounting for discrepancy in frequency was important and properly taken into account in the previous section.

The equation for ζ can be rearranged to solve for the percentage of mechanical energy damped because of the change in mass which gives a better understanding on the influence of the TMD,

$$\zeta - 64\% = E_{TMD} = 6.69 \ln m_p$$

Where E_{TMD} represents the energy dissipated by the TMD. This value is proportional to 6.69 ln m_p , where the coefficient 6.69 magnifies the effect of mass on E_{TMD} and thus dissipates more mechanical energy.

9 Conclusion

The experiment measured the acceleration during the second peak of oscillation to deduce the percentage of mechanical energy dissipated between the first two consecutive periods. The data does not support the hypothesis, which suggests a far more interesting conclusion about the nature of the damper. Although it was hypothesized that since the kinetic energy of the damper is directly proportional to the mass then it would similarly be proportional to dissipated energy, $KE_p \propto m_p \propto \zeta$, the experiment reveals that the dissipated mechanical energy is actually a logarithmic function $\zeta = k + A \ln m_p$ for some constant k and k. By measuring k0 and processing the raw data into k0 using equation 1, which using software concluded the final equation

$$\zeta(\%) = 64\% + 6.69 \ln m_p$$

Firstly, this conclusion suggests that m_p and ζ increase logarithmically rather than linearly, which translates into heavier TMDs providing increasingly less additional damping for larger masses. This suggests that although the kinetic energy of the TMD increases linearly with mass, less energy overall is transferred into the TMD when the mass is higher, so the TMD moves less and thus absorbs less of the mechanical energy ($KE = \frac{1}{2}m_pv^2$, velocity v is smaller). Moreover, it suggests that a lot of the kinetic energy of the TMD is transferred back into the kinetic energy of the structure, because the energy is dissipated primarily through the friction of the screw on the TMD which does not entirely dissipated the TMD's kinetic energy.

Secondly, the constant k=64% indicates that a TMD only provides additional damping to the initial loss of internal energy through friction within the structure. This coincides with established knowledge that imperfect oscillators are damped as energy is transferred to their surroundings through heat and internal friction. Although this function is inaccurate as m_p approaches 0 because ζ would approach $-\infty$, for small masses like $m_p=1g$ the function demonstrates how a large percentage of mechanical energy is dispersed regardless of the TMD.

Therefore, the conclusion answers the research question in that the mass of a Tuned Mass Damper affects the dissipation of mechanical energy in an oscillating structure in a logarithmic fashion, as for increasingly larger masses, less additional energy is dissipated. Therefore, engineers would find it redundant to increase the mass past a threshold that would keep civilians safe inside a building during a skyscraper, because additional mass would not affect damping much yet would cost more to implement. Also interestingly, the experiment reveals how even a small Tuned Mass Damper can have a large effect on the dissipation of mechanical energy. Therefore, these methods should be pursued in skyscrapers that may not have much space for larger masses, because even relatively low masses can radically improve the safety of civilians in an oscillating skyscraper after an earthquake.

10 Further Investigation

Further investigation into different types of dampers would assist in establishing architectural conventions that mitigate the effect of earthquakes and thereby save lives. Along with a tuned mass damper, there are other methods of damping worth investigating and comparing to the TMD to understand which one is more effective to be put in place. A tuned liquid damper, for instance, is another method of damping oscillations in buildings using sloshing liquids, but has similarly little research done on it (Bozer 2021). It would be worth comparing these two methods, along with other damping systems, to evaluate the effectiveness of each. This would strengthen the conclusion made by this investigation as maybe other systems are less effective and therefore TMD's with increasingly higher mass are the only way to provide effective additional damping.

11 Weaknesses and Improvements

Source of error	Significance for conclusion	Improvements		
and effect		to reduce error		
Systematic errors influencing accura				
Deformation of structure as plastic straws slightly bent forwards by the end of the experiment.	Overall could have artificially increased the dissipated mechanical energy as trials progressed skewing the data towards lower values than real value. Not that significant because trials in each increment do not get progressively lower, which indicates that deformation did not have a large effect.	Replace the straws with long plastic beams that would offer similar rigidity -flexibility but without deformation because of stronger structural integrity. Moreover, this would remove the need for toothpicks connecting the straws, which would mean that the bending would be dispersed evenly across the beam and therefore approximate an ideal oscillating structure.		
Random errors influencing precision				
Accelerometer sensor on Raspberry Pi had small random spikes that deviated from damped simple harmonic motion, seen in Figure 5 at t = 1.6s	Not largely significant because the spikes are small compared to the amplitude of the damped simple harmonic motion, can be calculated with standard deviation between a_2 trial values which ensured that any outliers caused by random spikes in sensor were removed.	Conduct experiment with more precise sensor that would be more consistent when measuring acceleration.		
Unintentional force applied by hand when releasing structure from the displaced position against the wall.	Largely reduced by the careful methodology used as the structure was released from behind held against a wall, although this was the most significant source of error and likely caused the outlier in trail 7 row 1. This would have artificially increased the acceleration of the object, as F = ma, which would skew the results into different initial acceleration values and therefore higher data gathered than real values.	Use electro-mechanical earthquake simulator that would shake the structure by a set force instead of depending on human contact. Using a mechanical oscillator which simulates an earthquake with constant frequency and displacement would greatly increase the resemblance of the experiment to a real earthquake, and would validate the data with more precision due to increased regulation over control variables and no additional force.		

Table 3: Table showing sources of error, their effect, and improvements

12 Strengths

Strengths	Effect on Conclusion
Digital sensor was used to record accelerometer data.	The structure moved significantly faster than a human could detect and measure, therefore using a digital sensor improved the precision and accuracy of the data and conclusion.
Accelerometer was attached to 'RPi 3' microcomputer which allowed for simple data extraction using a WiFienabled secure connection without need for disassembly to extract data.	Remote data extraction allowed me to not physically interfere with the TMD structure and thereby not alter any controlled variables like bending, hence increasing precision.
Programmed my own script to process raw data into visual graph.	Since the raw data consisted of thousands of accelerometer snapshots, finding a_2 would have taken large amounts of time if not for my own script that visualized acceleration versus time, making finding a_2 as simple as quantitatively locating the second peak of the graph, then extracting that acceleration.
Straws reinforced with toothpicks were chosen for the structure.	Allowed structure to bend without completely falling apart, imitating the motion of real-life skyscrapers. The elastic property allowed the damping to be measured while maintaining structural integrity, thus producing damped simple harmonic motion, which is assumed in the equations used for energy.

Table 4: Table showing strengths of the experiment and their effects

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13 Appendix A

To find the uncertainty, $\Delta \zeta$, where ζ defined as:

$$\zeta = \frac{\Delta d_1^2 - (\frac{a_2}{4\pi^2 f^2})^2}{\Delta d_1^2} \times 100\%$$

Since constant constant terms can be ignored when calculating for uncertainty, ζ can be simplified to:

$$(\frac{100}{16\pi^4 \Delta d_1^2})(\frac{{a_2}^2}{f^4})$$

Now, using the three uncertainty rules from the data booklet, $\Delta \zeta$ can be calculated using:

$$\begin{aligned} x &= a_2^2 \rightarrow \frac{\Delta x}{x} = \mid 2\frac{\Delta a_2}{a_2} \mid \rightarrow \Delta x = 2a_2\Delta a_2 \\ y &= f^4 \rightarrow \frac{\Delta y}{y} = \mid 4\frac{\Delta f}{f} \mid \rightarrow \Delta y = 4f^3\Delta f \\ \\ z &= \frac{x}{y} \rightarrow \frac{\Delta z}{z} = \frac{\Delta x}{x} + \frac{\Delta y}{y} \rightarrow \Delta z = (\frac{a_2^2}{f^4})(\frac{2a_2\Delta a_2}{a_2^2} + \frac{4f^3\Delta f}{f^4}) \rightarrow \Delta z = \frac{2a_2\Delta a_2}{f^4} + \frac{4a_2^2\Delta f}{f^5} \\ & \zeta = (\frac{100}{16\pi^4\Delta d_1^2})z \\ \\ \therefore \Delta \zeta &= (\frac{100}{16\pi^4\Delta d_1^2})\Delta z \rightarrow \Delta \zeta = (\frac{100}{16\pi^4\Delta d_1^2})(\frac{2a_2\Delta a_2}{f^4} + \frac{4a_2^2\Delta f}{f^5}) \end{aligned}$$

14 Appendix B

Program running on microcomputer to record sensor data.

```
\#\ import\ libraries\ for\ accelerometer , time, and storing
from sense_hat import SenseHat
import time, json
# initialize sensor
sense = SenseHat()
# periodically record acceleration data and save to file
data = []
while True:
    \mathbf{try}:
        time. sleep (0.01)
        acceleration = sense.get_accelerometer_raw()
        y = acceleration ['y']
        print("[+] _ gathering _ data")
        data.append(y)
    except KeyboardInterrupt:
        with open('data.json', 'w', encoding='utf-8') as f:
             print("[+]_Saving_to_json_file")
             json.dump(data, f, ensure_ascii=False, indent=4)
        exit()
```

15 Appendix C

Program running on laptop to visualize raw data.

```
\#\ import\ data\ processing\ and\ visualization\ libraries
import matplotlib.pyplot as plt
import json
import numpy as np
# load data
data = []
with open("data.json", "r") as f:
    data = json.load(f)
\# use timestep of sensor to understand real-life time elapsed
TIMESTEP = 0.029
x = list(range(1, len(data)+1))
x = [x*TIMESTEP for x in range(len(data))]
\# set up graph
plt.plot(x, data, label="x-axis")
plt.xticks(np.arange(min(x), max(x)+1, 0.5))
plt.yticks(np.arange(round(min(y1))-1, round(max(data))+1, 0.5))
plt.xlabel(r'time_{-}(\$s\$)')
plt.ylabel(r'acceleration (\$ms^{-2}\$)')
plt.title('Acceleration_vs_Time_Graph')
plt.legend()
plt.show()
```