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Dr. Zekavat's 140 Lab  
Week 4

## Summary

This week we were given measurements of distance as a carriage fell to the ground. The measurements were taken by some mysterious device that has something to do with hertz. The task was to calculate average acceleration by first calculating all the pointwise accelerations, by first calculating all the pointwise velocities. We took our average acceleration and calculated the percent error with an accepted value for average acceleration, 980 cm/s/s. Then we took a slope of the velocity vs. time graph and found that it was about 10 cm/s/s off from our average acceleration.

## Data Sheet

Experiment M-5

Data Sheet

Mass of carriage (m): 1237.7 g

Frequency of spark (f): 30 Hz

Time per interval ( $\Delta t$ ): 0.033 s

$\frac{\Delta s}{0.033}$

$\frac{\Delta v}{0.033}$

Spot #	t	s	$\Delta s$	v	$\Delta v$	a
	sec	cm	cm	cm/s	cm/s	cm/s <sup>2</sup>
0	0	$s_0 = 0$				
1	0.033	$s_1 = 1.35$	$s_1 - s_0 = 1.35$	$v_1 = 40.9$		
2	0.066	$s_2 = 3.8$	$s_2 - s_1 = 2.45$	$v_2 = 74.2$	$v_2 - v_1 = 33.3$	$a_1 = 1009$
3	0.099	$s_3 = 7.2$	3.4	103.03	28.788	872.36
4	0.133	$s_4 = 11.8$	4.6	139.39	36.364	1101.93
5	0.166	$s_5 = 17.4$	5.6	169.697	30.303	918.27
6	0.199	$s_6 = 24.1$	6.7	203.03	33.333	1010.10
7	0.232	$s_7 = 31.9$	7.8	236.36	33.333	1010.10
8	0.265	$s_8 = 40.7$	8.8	266.667	30.303	918.27
9	0.298	$s_9 = 50.6$	9.9	300	33.333	1010.10

Average acceleration: 981.405 cm/s<sup>2</sup>

Percent error: 0.14336 %

↑ add all  $\Delta s$   
8

100 x  $\frac{980 - \text{average acceleration}}{980}$

## Calculations

```

1 from typing import Tuple
2 import pandas as pd
3 import numpy as np
4 import altair as alt
5
6 delta_t = 0.033
7
8 def mkdf(delta_t: float) -> Tuple[pd.DataFrame, pd.DataFrame, pd.DataFrame]:
9     positions = pd.DataFrame({
10         "t": [0, 0.033, 0.066, 0.099, 0.133, 0.166, 0.199, 0.232, 0.265, 0.298],
11         "s": [0, 1.35, 3.8, 7.2, 11.8, 17.4, 24.1, 31.9, 40.7, 50.6]
12     })
13
14     velocity = pd.DataFrame({
15         "delta_s": [position.s[k+1] - position.s[k] for k in range(position.shape[0] - 1)]
16     })
17     velocity = velocity.assign(v=velocity.delta_s / delta_t)
18
19     acceleration = pd.DataFrame({
20         "delta_v": [velocity.v[k+1] - velocity.v[k] for k in range(velocity.shape[0] - 1)]
21     })
22     acceleration = acceleration.assign(a=acceleration.delta_v / delta_t)
23
24     return position, velocity, acceleration

```

```

[39] 1 avg_accel = acceleration.a.sum() / acceleration.shape[0]
      2 percent_error = abs(100 * (980 - avg_accel) / 980)

```

```

[40] 1 avg_accel

```

```

☐→ 981.4049586776857

```

```

[44] 1 percent_error

```

```

☐→ 0.14336313037609458

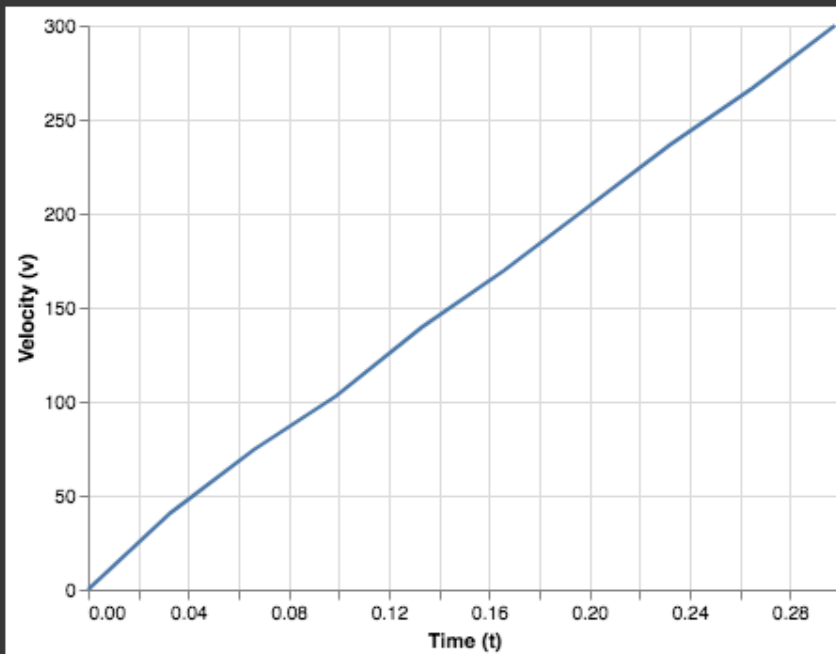
```

## Analysis & Conclusion

It was fun getting a sense of how velocity and acceleration is calculated under my fingernails. It is disappointing to do it during Covid because I would have liked to see the measuring instrument in action. I was surprised that the slope of the line of best fit wasn't closer to the average acceleration. In conclusion, position, velocity, and acceleration are mathematically related by a chain of derivatives. We can approximate this chain of derivatives by taking pointwise differences and dividing by delta t. Using a linear regression, the slope of the line of best fit of the velocity points is within 1% of the pointwise-calculated average acceleration.

# Velocity vs. Time Graph

```
1 velocity_prime = pd.DataFrame({"delta_s": [np.nan], "v": [0]})
2 alt.Chart(
3     position.join(pd.concat((velocity_prime, velocity)).reset_index())
4 ).mark_line(
5 ).encode(
6     x=alt.X("t", title="Time (t)"), y=alt.Y("v", title="Velocity (v)")
7 )
```



```
[70] 1 from sklearn.linear_model import LinearRegression
2     model = LinearRegression()
3
4     model.fit(v_vs_t.t.values.reshape(-1, 1), v_vs_t.v)
5
6     print(f"Slope of line of best fit: {model.coef_[0]}")
```

```
☐ Slope of line of best fit: 991.2646120415806
```