

# CS 5A

Feb. 21

# Reminders

- Lab05, Worksheets 5 and 6 are released, **due Thursday Feb 27 at 11:59pm**
- Peer Eval Survey due tonight at 11:59pm, please fill these out if you haven't

# Correlation

Correlation measures the strength and direction of relationship between two variables

Ranges from -1 to +1

- +1: Perfect positive correlation
- -1: Perfect negative correlation
- 0: No correlation

# Correlation

Pearson correlation formula:

$$r = \Sigma((x - \mu_x)(y - \mu_y)) / (\sigma_x \sigma_y)$$

Code:

```
np.corrcoef(column_1, column_2)[0][1] -> r
```

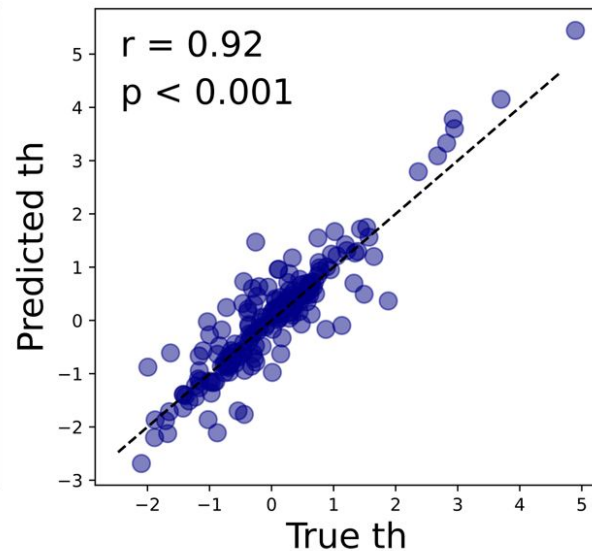
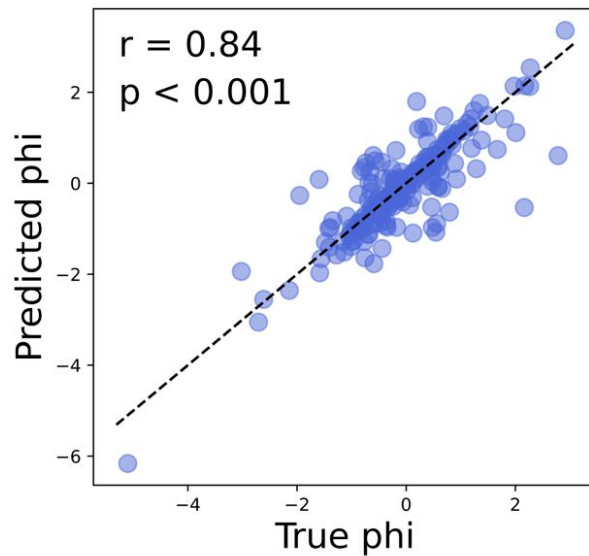
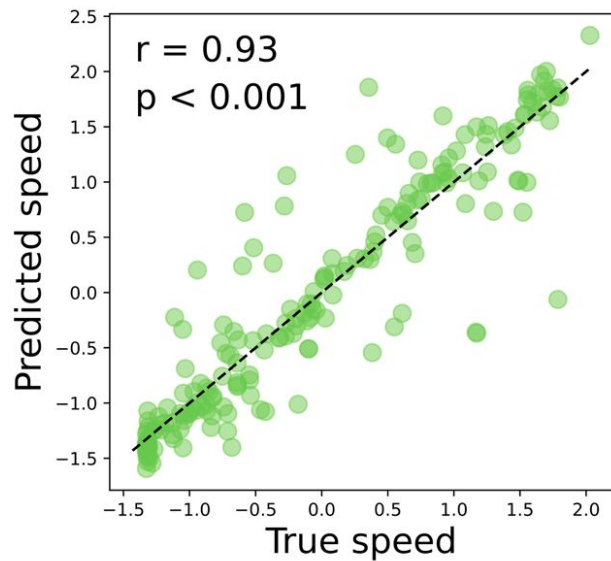
**Output:**

$$\begin{bmatrix} 1 & r \\ r & 1 \end{bmatrix}$$

where  $r$  is the Pearson correlation coefficient between  $x$  and  $y$ .

# Correlation

Visnav lateral Multitask Decoding - Speed + Eye Gaze (phi, th)



# Sampling Bias

## 1. Convenience Sampling Bias

- Selectively choosing data points can artificially increase correlation
- Can lead to misleading conclusions
- Common issue in research when data collection isn't random

```
np.corrcoef(data.column('quality'), data.column('alcohol'))
```

**Correlation: 0.4**

```
# Convenience sampling (selecting only high quality wines) biased_sample =  
data.where('quality', are.above(6)).where('alcohol', are.above(10)).take(np.arange(100))
```

*New correlation after convenience sampling*

```
biased_corr = np.corrcoef(biased_sample.column('quality'), biased_sample.column('alcohol'))
```

**Correlation: 0.7**

**Random Sampling creates a more representative correlation coefficient**

# Probabilities

$P(\text{Event}) = \text{Number of Favorable Outcomes} / \text{Total Number of Possible Outcomes}$

- Or more formally:  $P(A) = n(A) / n(S)$  where  $n(A)$  is count of event **A**, and  $n(S)$  is total sample size

```
matching_students = students.where('height', are.above(175)) \
    .where('weight', are.between(60, 70)) \ .where('age', are.below(22))

probability = matching_students.num_rows / students.num_rows
```

**Can sampling bias affect probabilities?**

# Law of Averages

## Law of Averages:

- Empirical probability approaches theoretical probability as trials increase

## Example - coin flipping

- theoretical probability is 50% heads, 50% tails
- if you flip 5 times, you might get 4 heads and 1 tail - 80%/20%!
- does this scale to 5000 coin flips?



# Lab05 Examples

# Simulating an event

```
def simulate_probability(n_samples):  
    return (students.sample(n_samples).where('height', are.above(175)) \  
            .where('weight', are.above(70)).num_rows / n_samples)  
  
# Run multiple simulations  
  
n_simulations = 10  
  
results = [] for _ in range(n_simulations): results.append(simulate_probability(100))  
  
true_prob = (students.where('height', are.above(175)) .where('weight', are.above(70)) .num_rows /  
students.num_rows)
```

# Simulation

[https://www.youtube.com/watch?v=SCNr\\_Lom5z8](https://www.youtube.com/watch?v=SCNr_Lom5z8)

The distribution will always tend towards a bell curve with more samples like this!