

# Capturing ultra-rare CRISPR-Cas9 off-target editing events in single cells

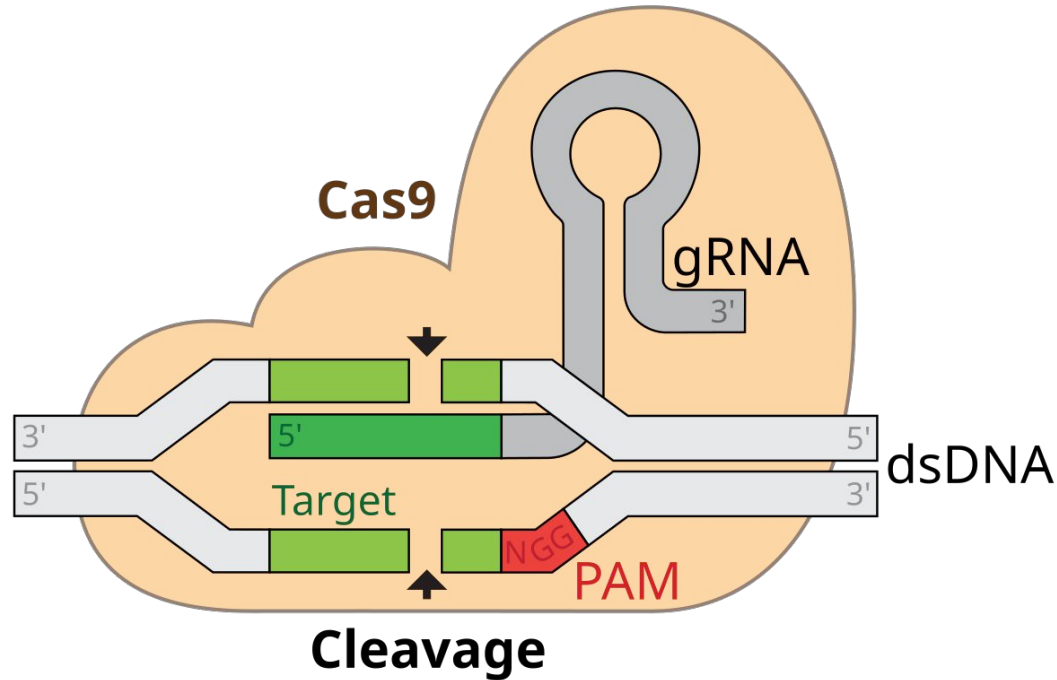
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UC LEADS  
UCSD SRC 2025  
8/13/2025



# What is CRISPR-Cas9?

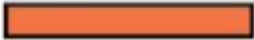


- Genome editing tool
- **Cas9** cuts DNA at specific sites
- A guide RNA (**gRNA**) leads Cas9 to the target
- Cas9 makes a double-strand break (**DSB**) near a **PAM (NGG or NAG)**
- DSBs repaired by imperfect repair mechanisms, which can introduce **indels**



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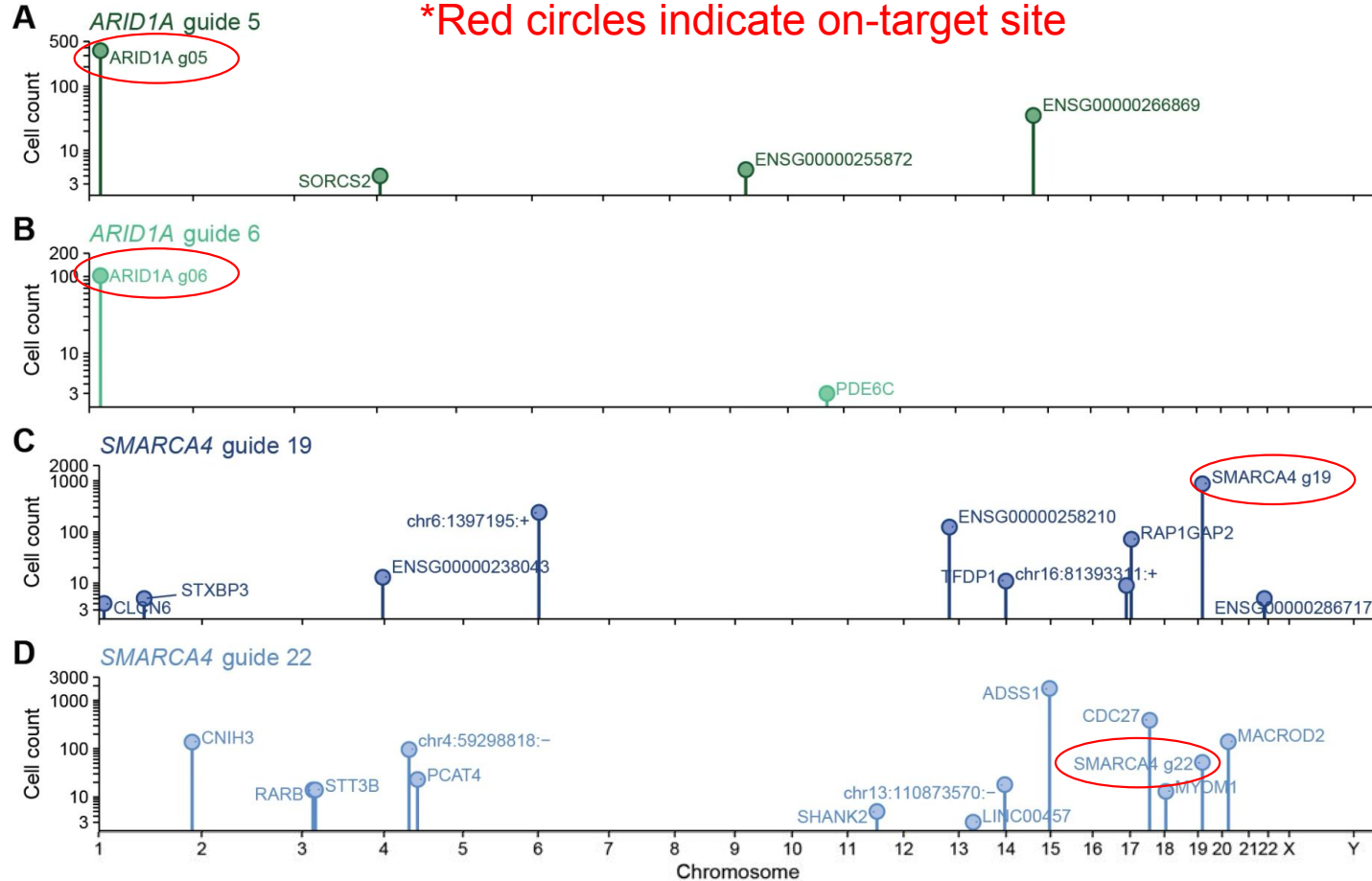
# CRISPR Often Cuts DNA at Unintended Sites

- CRISPR can cut unintended sites with similar sequences to the guide RNA

	CHD3 guide 10 Target sequence PAM	Gene hg38 position (strand)	Cell count log2
5'	AATATGGAACCGGACCGGGT	<i>CHD3</i>	1825
5'	AATATGGAACCGGACCGGGT CGG	chr17:7890637 (+)	
	AATATGGAACCGGACCGGGT	<i>ENSG00000287360</i>	408
	AATGTGGAACAGGACCGAGT GGG	chr4:14359411 (-)	
	AATA-TGGAACCGGACCGGGT	<i>USP9X</i>	311
	CA-AGTGGAACCGAGACCGGGT AGG	chrX:41086513 (-)	

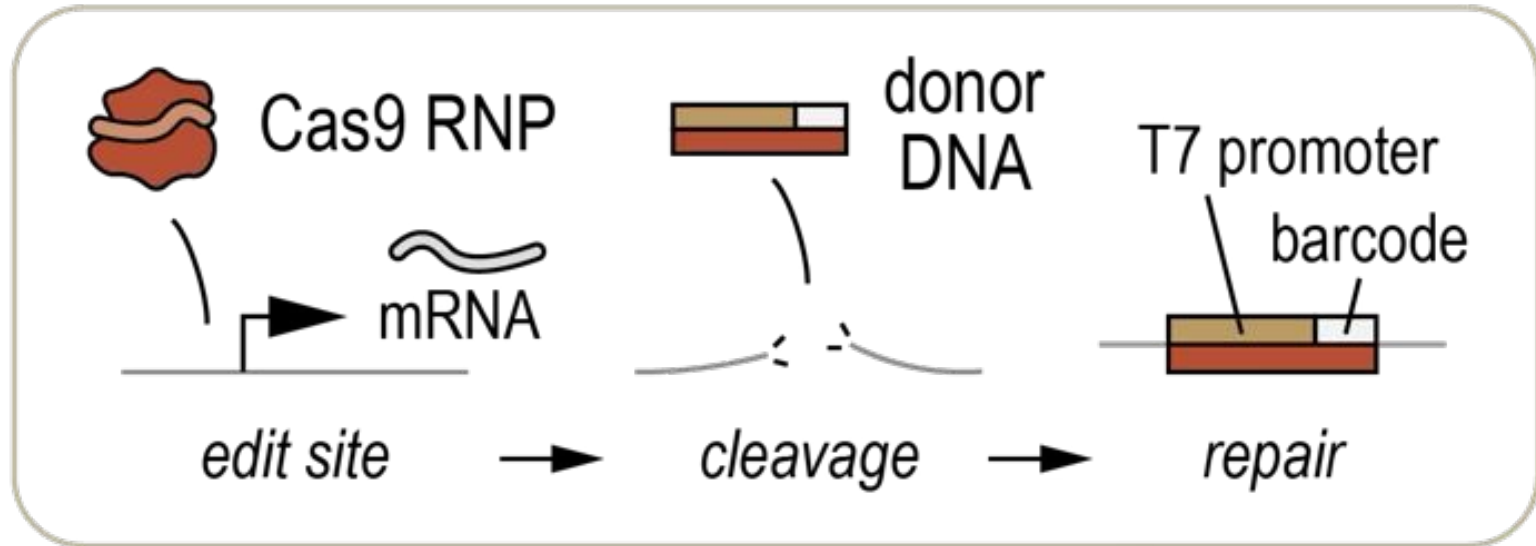
- Current tools are not effective at detecting/predicting off target edits

# Enter Superb-seq



# Superb-seq Enables Single-Cell Detection of CRISPR Edits

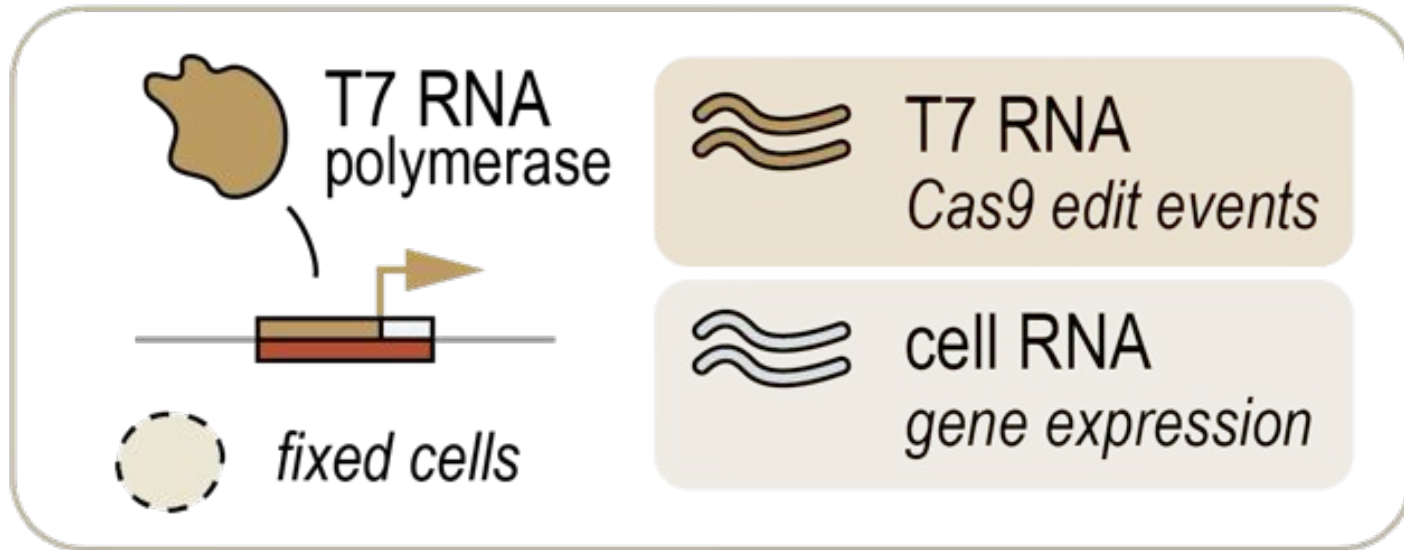
## 1 Cas9 edit labeling



*Homology-free labeling of on/off-targets*

# Superb-seq Enables Single-Cell Detection of CRISPR Edits

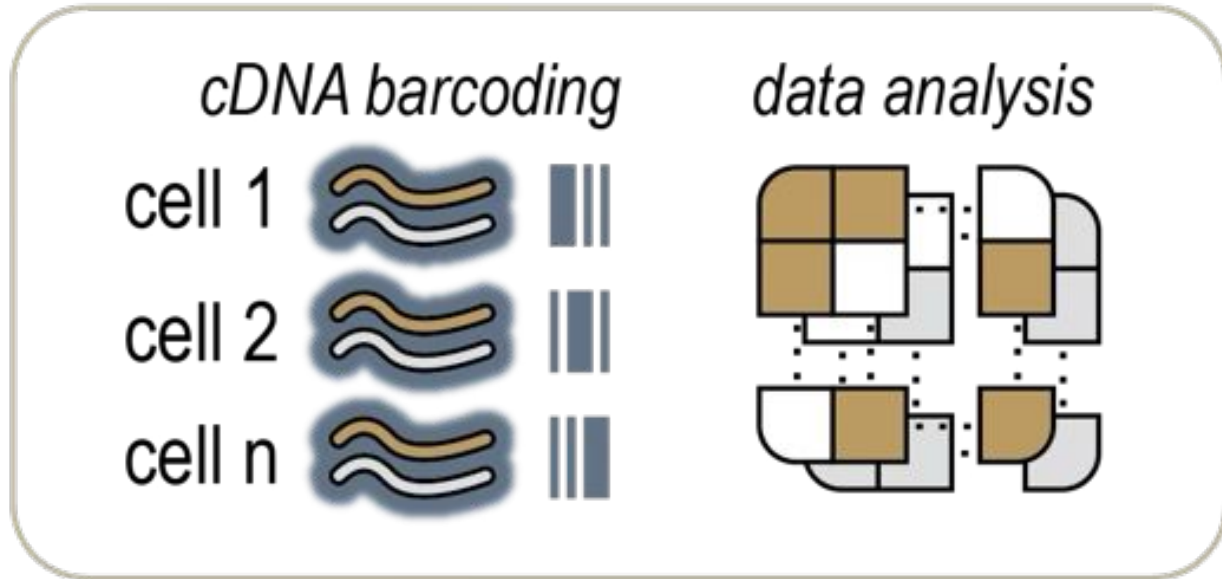
## 2 In situ transcription



*Single-cell genome amplification*

# Superb-seq Enables Single-Cell Detection of CRISPR Edits

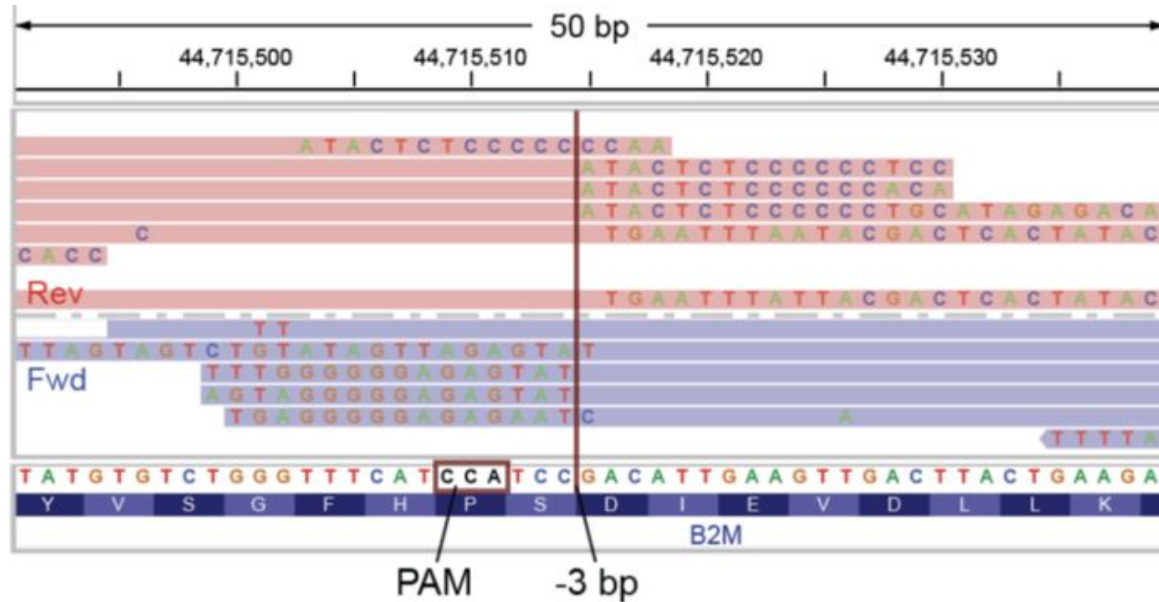
## 3 Combinatorial scRNA-seq



*Off-the-shelf kits, custom software*

# Sheriff Identifies Candidate CRISPR Edits from Superb-Seq Data

- Software that identifies CRISPR edits by locating the T7 Barcode (cut site marker)
- Uses a set of filters to remove false positives
- Requires edits to appear in  $\geq 3$  cells + bidirectional support
- Downside: Misses ultra-rare (1–2 cell) edits



# Project Goals and Hypothesis

## Goal:

- Detect ultra-rare CRISPR editing events from Superb-seq data

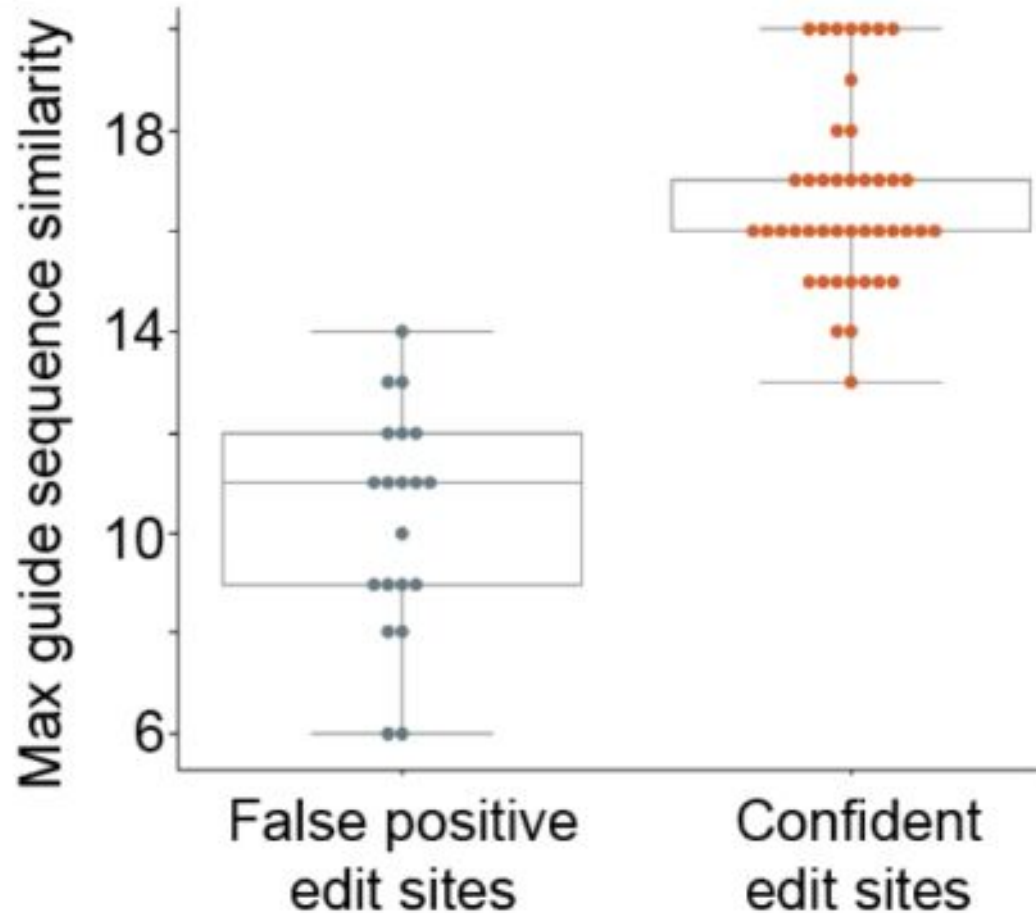
*Why? To gain deeper insights into CRISPR editing and the interpretability of single-cell perturbation screens*

## Hypothesis:

- I hypothesize that some of these ultra-rare candidate edits represent real CRISPR-Cas9 off-target events.

# High False Positive Rate Even With Standard Filtering

- With standard filters set, 20/63 calls were still false positives
- Loosening the filters will only produce more



# Methods Overview

Step 1. Gather Candidate Sites

Step 2. Assign and Score Guide Match

Step 4. Compute  
P-Values

Step 3. Construct Null Model

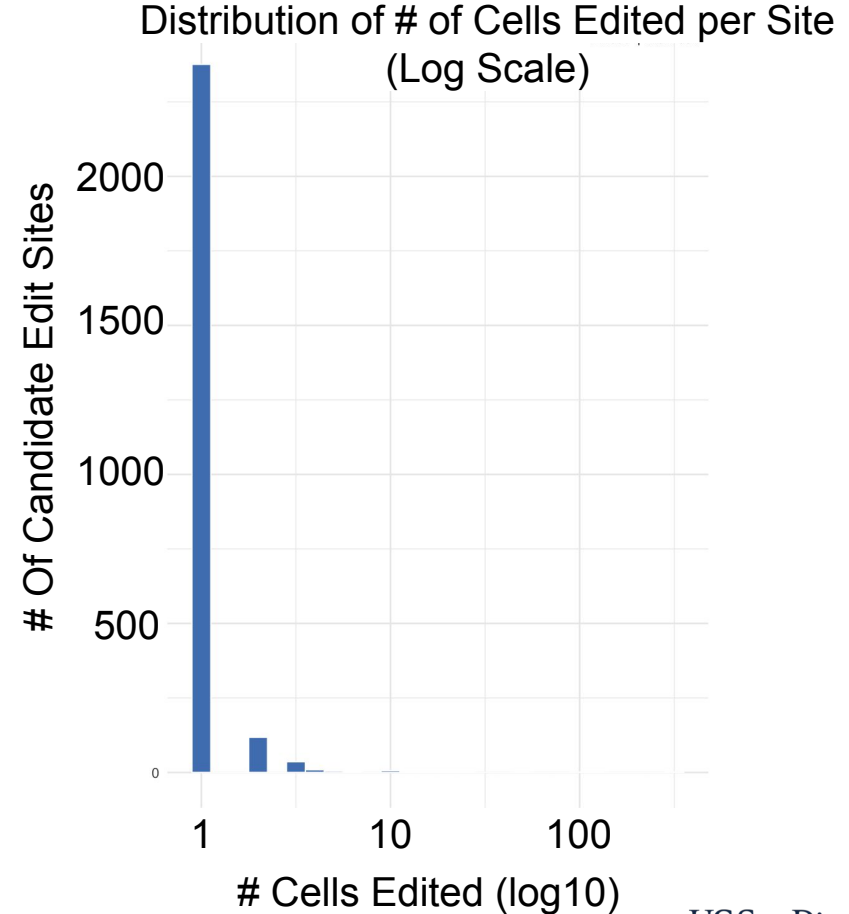


\*Note: Only analyzed data from chromosome 1 to allow reasonable runtime\*

# Relaxed Sheriff Reveals Thousands of Rare Candidate Edits

Sheriff Setting	# Edits Detected
Strict (default)	6
Relaxed (mine)	2562

- Problem: How do we differentiate the quality of all these new edit sites?



# Identifying Likely Off-Targets via Guide Homology Scoring

- **Candidate Site to Guide Homology Scoring**
- **Method:**
  - Extract **10bp window** around candidate site
  - identify all **valid PAM (NGG or NAG) sequences**
  - score guides against the **adjacent 20 bp** with **global sequence alignment:**
    - base match = +1
    - mismatch = -1
    - gap open = -2
    - gap extend = -0.5

Guides Considered:

- ARID1A g05
- ARID1A g06
- SMARCA4 g19
- SMARCA4 g22

SMARCA4 guide 22 Target sequence PAM	Gene hg38 position (strand)
 <p>ACCGAGGGTGGAGGTAGCGT ACCGAGGGTGGAGGTAGCGT TGG</p>	<b>SMARCA4</b> chr19:10986484 (-)
 <p>ACCGAGGGTGGAGGTAGCGT TGGGAGGGAGGAGGTAGCGT CGG</p>	<b>ADSS1</b> chr14:104729483 (+)

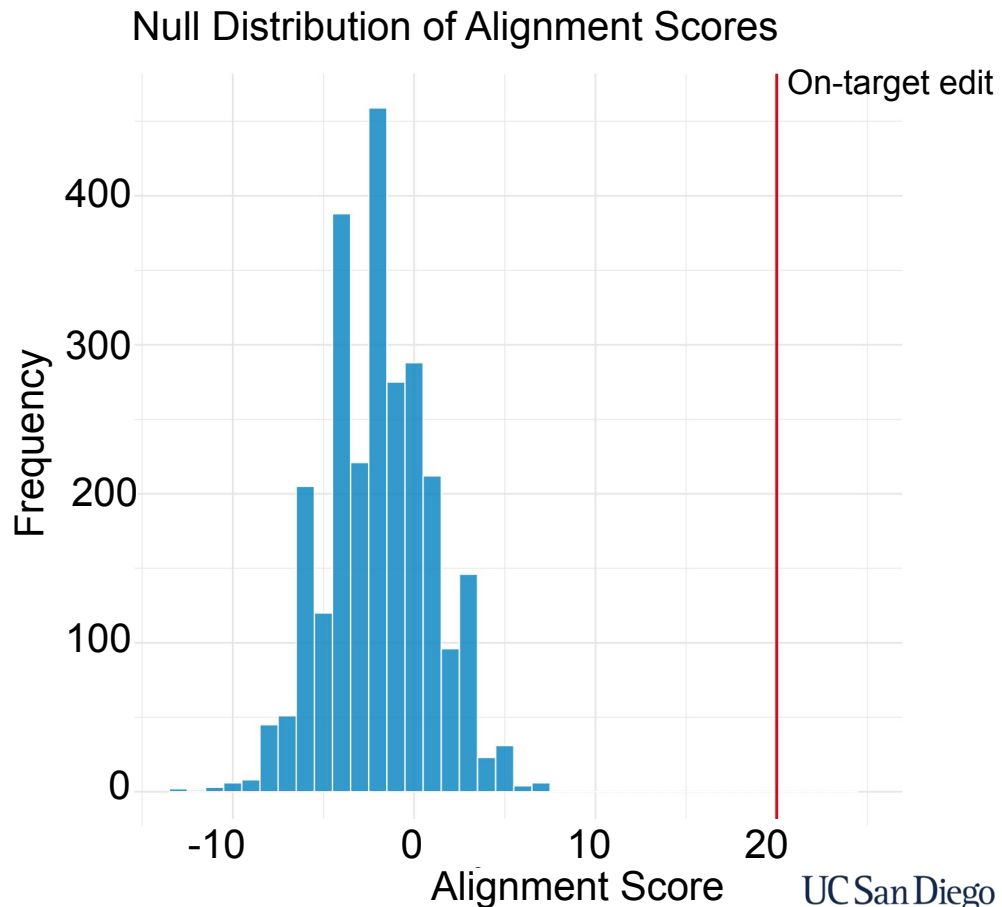
# Building a Null Model to Evaluate Off-Target Evidence

## Problem:

- Many guide-edit matches could happen by random sequence similarity.

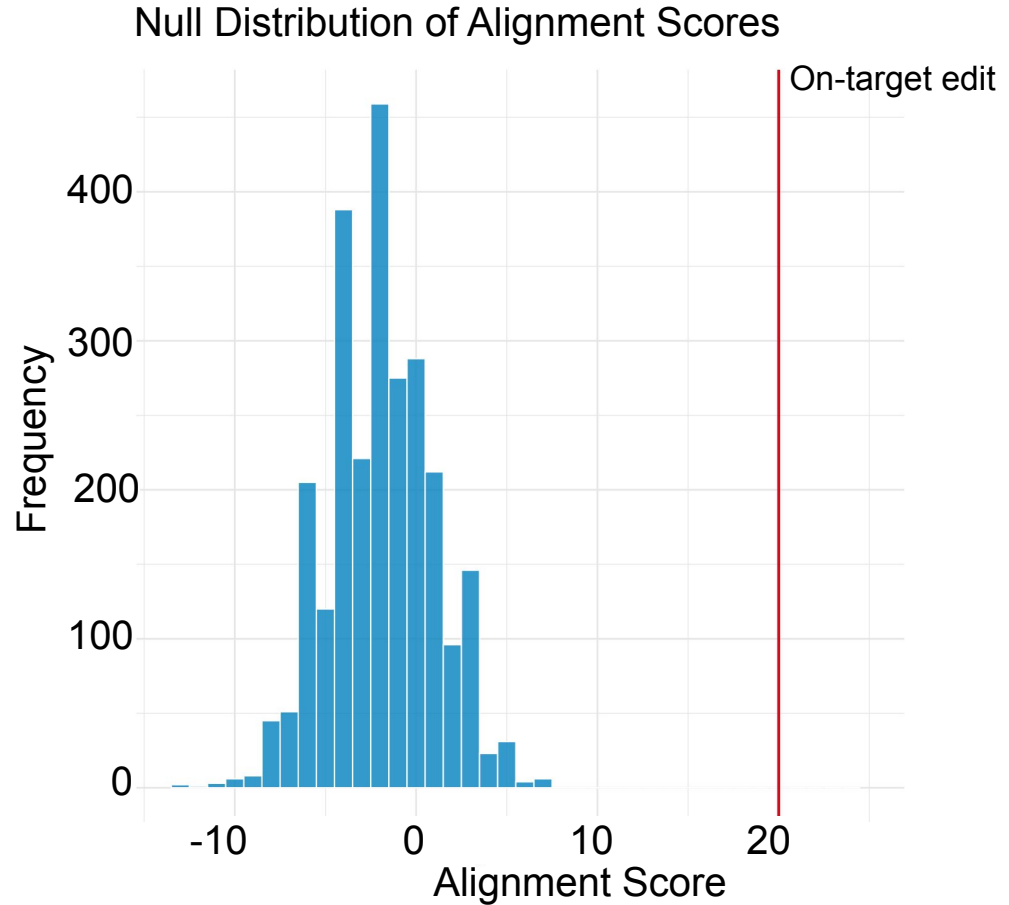
## Solution:

- Build null model
  - Scramble guide sequences
  - Rerun guide homology scoring



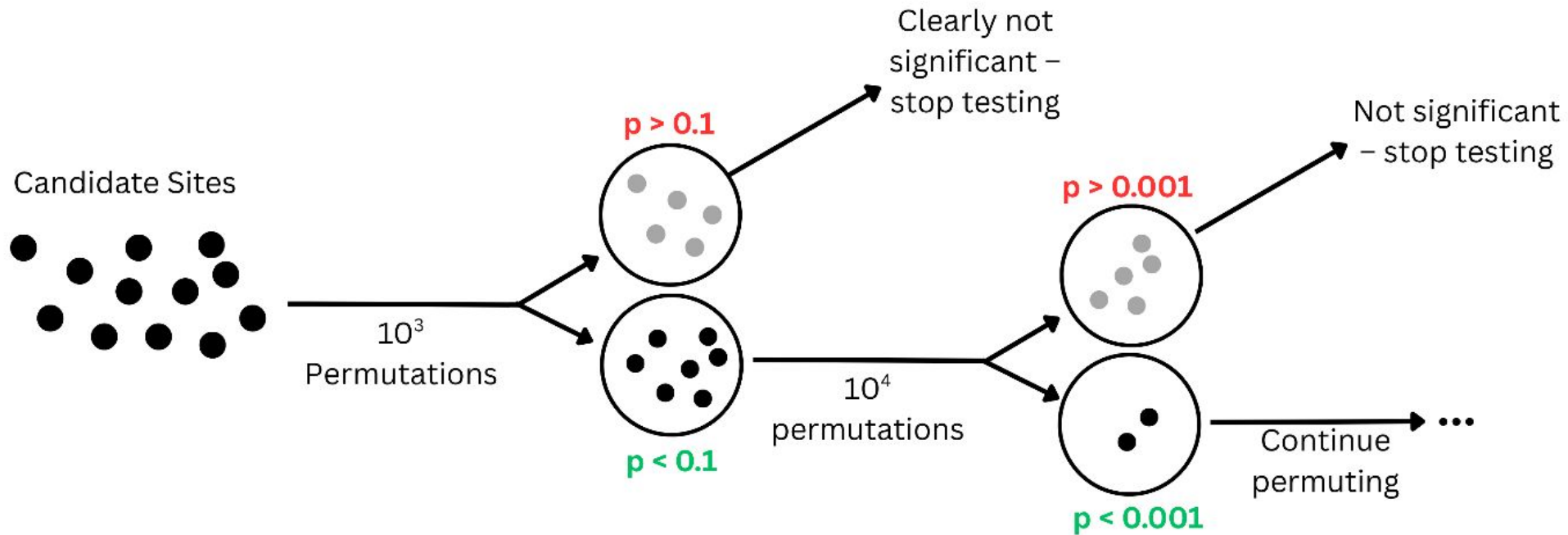
# Empirical P-Value

Proportion of scrambled scores  $\geq$  observed score



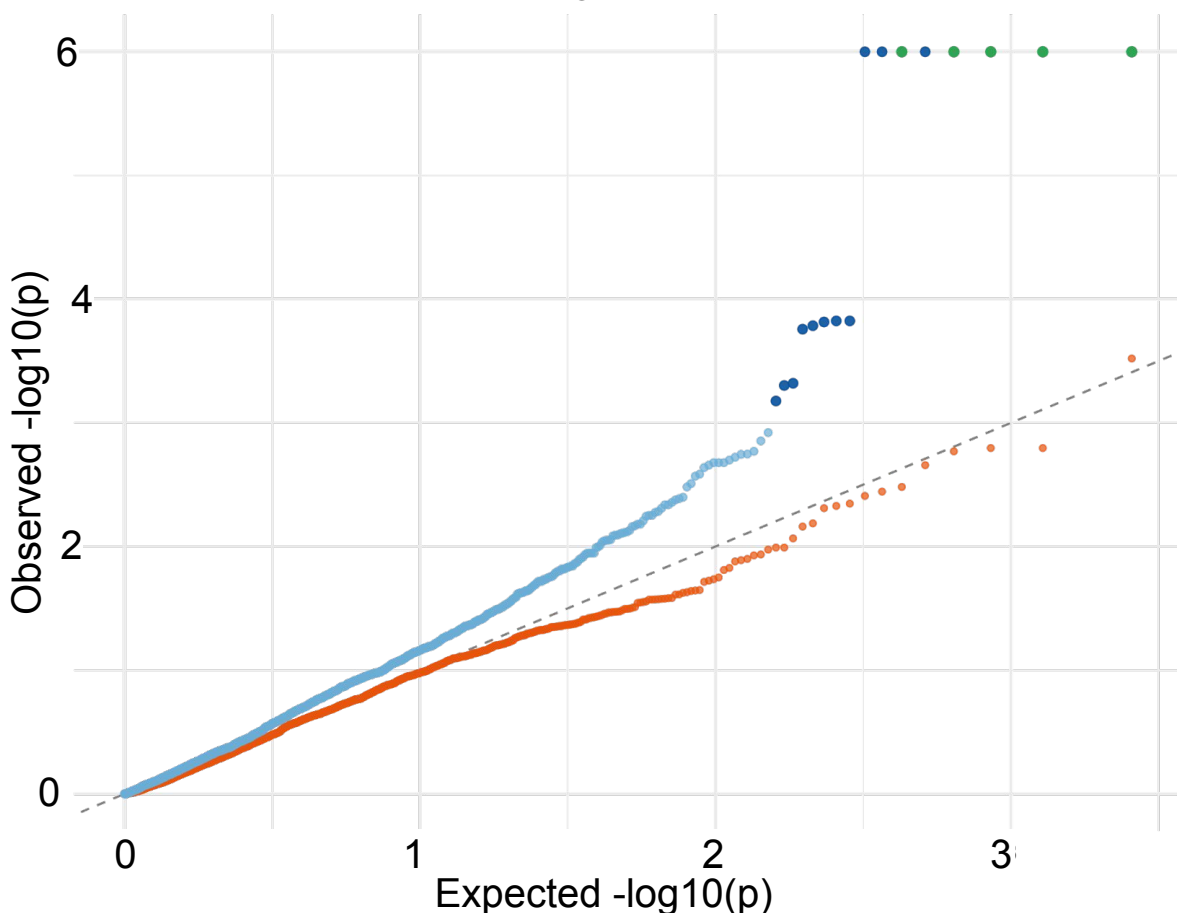
# Adaptive Permutation P-Value

Increases permutations for sites with low p-values  
to refine significance



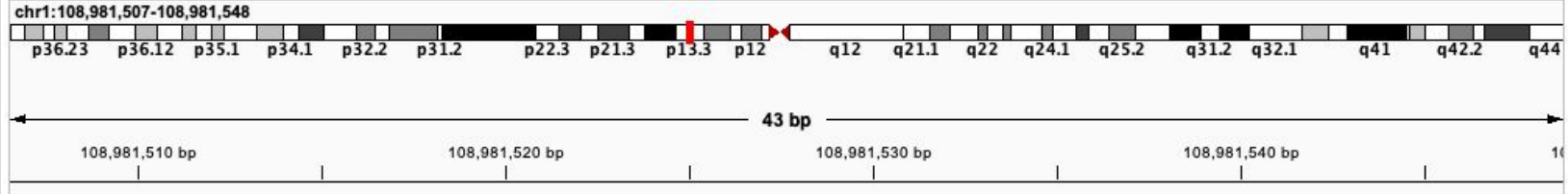
# 11 New Ultra-Rare Off-Targets Pass Statistical Thresholds

QQ Plot: Real vs Scrambled ( $-\log_{10} p$ ) (Adaptive Empirical P-Values)



Significant Sites	Count
Confirmed (prior work)	5
New (this study)	11

- Confirmed (prior)
- Real (other)
- Real ( $p < 0.001$ )
- Scrambled



False Positive:  
Sequencing error

IGV\_readable\_t7\_only.bam



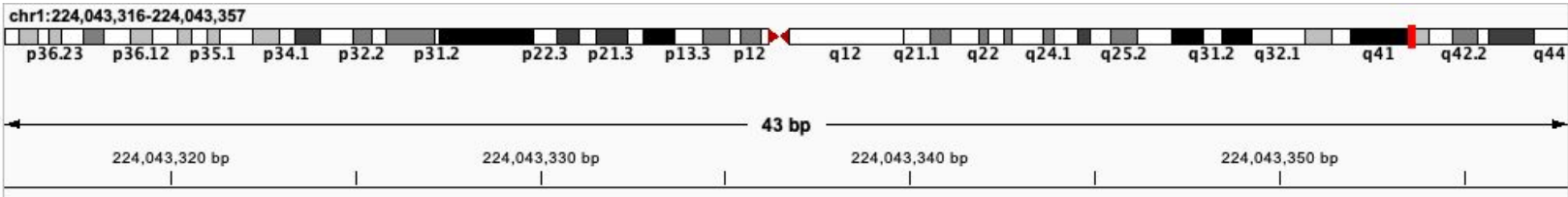
Sequence



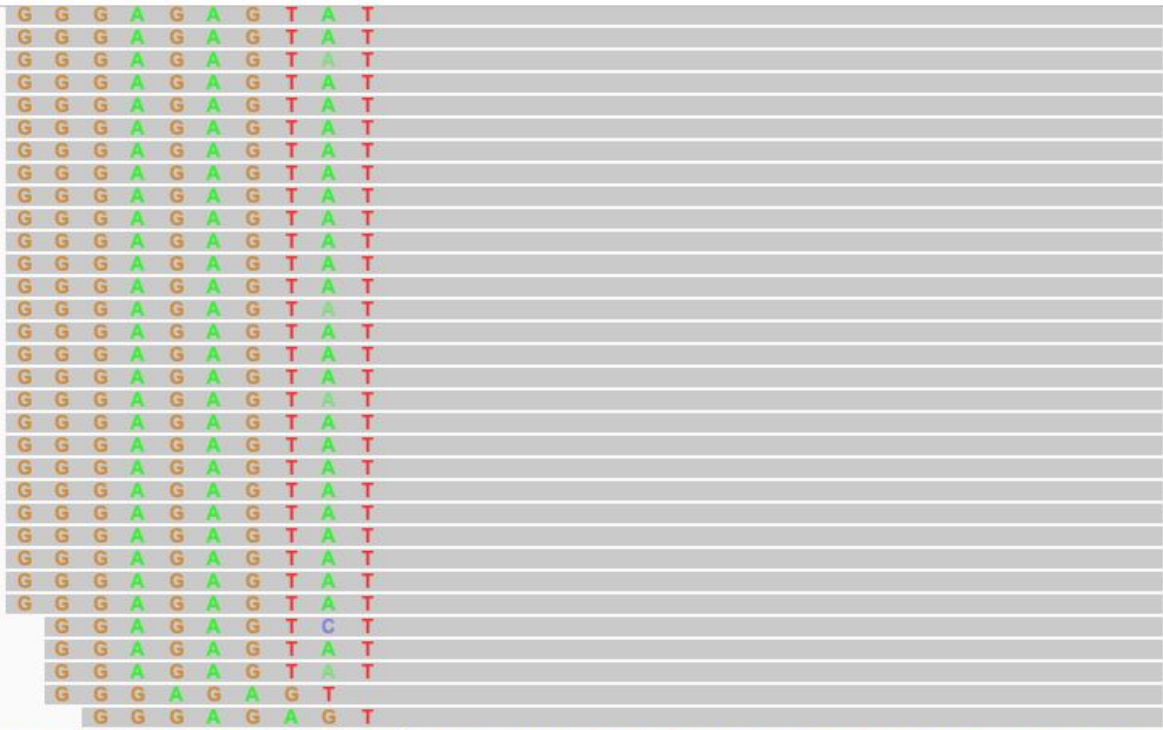
G G G A C A T T A T A T T T A A T C T C C C T G G G C C T T G G T T C C C T A A G C

Refseq All

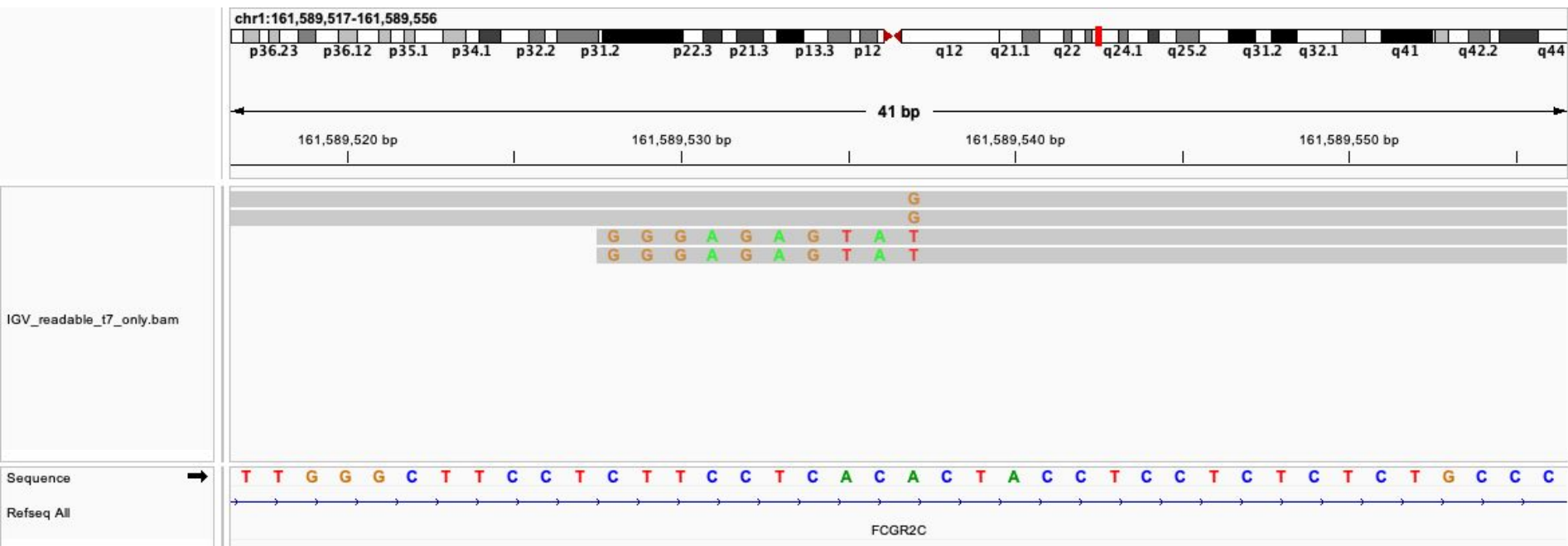
WDR47



Real edit:  
 Occurred in 23 cells, but  
 lacked bidirectionality



# Real Edit: Single cell edit event



# Conclusions and Future Directions

## Conclusions:

- Statistical analysis revealed 11 new significant candidate off-target sites
- 2 sites were manually validated as real edits missed by Sheriff
- Confirms existence of ultra-rare Cas9 off-target edits overlooked by standard Sheriff pipeline

## Future Directions:

- Refine method to better detect rare edits and reduce false positives.
- Expand analysis to the rest of the genome and more guides
- Incorporate gene expression to assess functional impact of edits

# Acknowledgements

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- **You (the audience)** – for your time and attention

**Thank you!**  
**Any questions?**