



PRESENTER

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## Background



**Why Does This Study Matter?** The issue of recurring injuries and its impact on teams and players. Specifically considering injuries to MLB (baseball pitchers in general), most studies focus on;

1. Frequency of injury,
2. Performance post-injury,
3. Injury detection, and
4. Strategies for injury prevention.

## Current Challenges

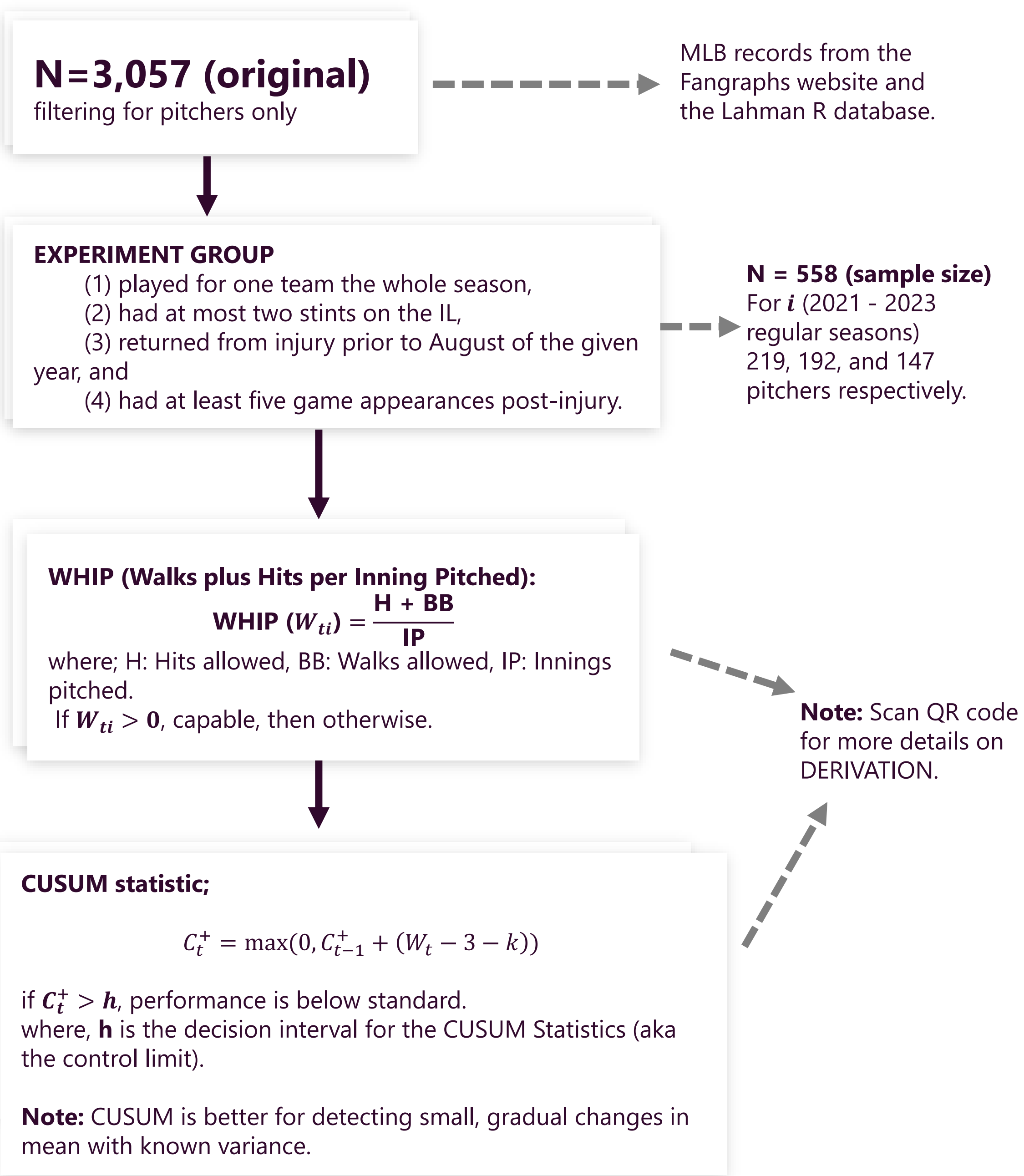
**Monitoring methods rely on *retrospective analysis* rather than *real-time* tracking.** This suggests that traditional approaches:

1. Analyze past injuries instead of detecting risks in real time.
2. May not be proactive in preventing injuries.
3. Could delay intervention, making it difficult to prevent re-injury before it occurs.

## Problem Statement

This work aims to develop a proactive system using CUSUM control charts to detect early signs of injury risk in MLB pitchers. By identifying gradual performance declines, the system will allow teams to intervene before injuries occur, reducing re-injury rates and improving player longevity. The goal is to find the optimal CUSUM threshold (*h*) for early injury detection, minimizing injury-related costs, improving player health, and ultimately saving team resources.

## Methods



## How Reliable is this Approach?

**Results**

- Injury Detection Range:  $h=1$  to  $h=20$  sensitivity and specificity for early injury prediction are presented.
- Game Appearances Before Injury, Game Appearances Before Signal, and Game Saved are presented in the plot for  $h=1$  to  $h=20$ .

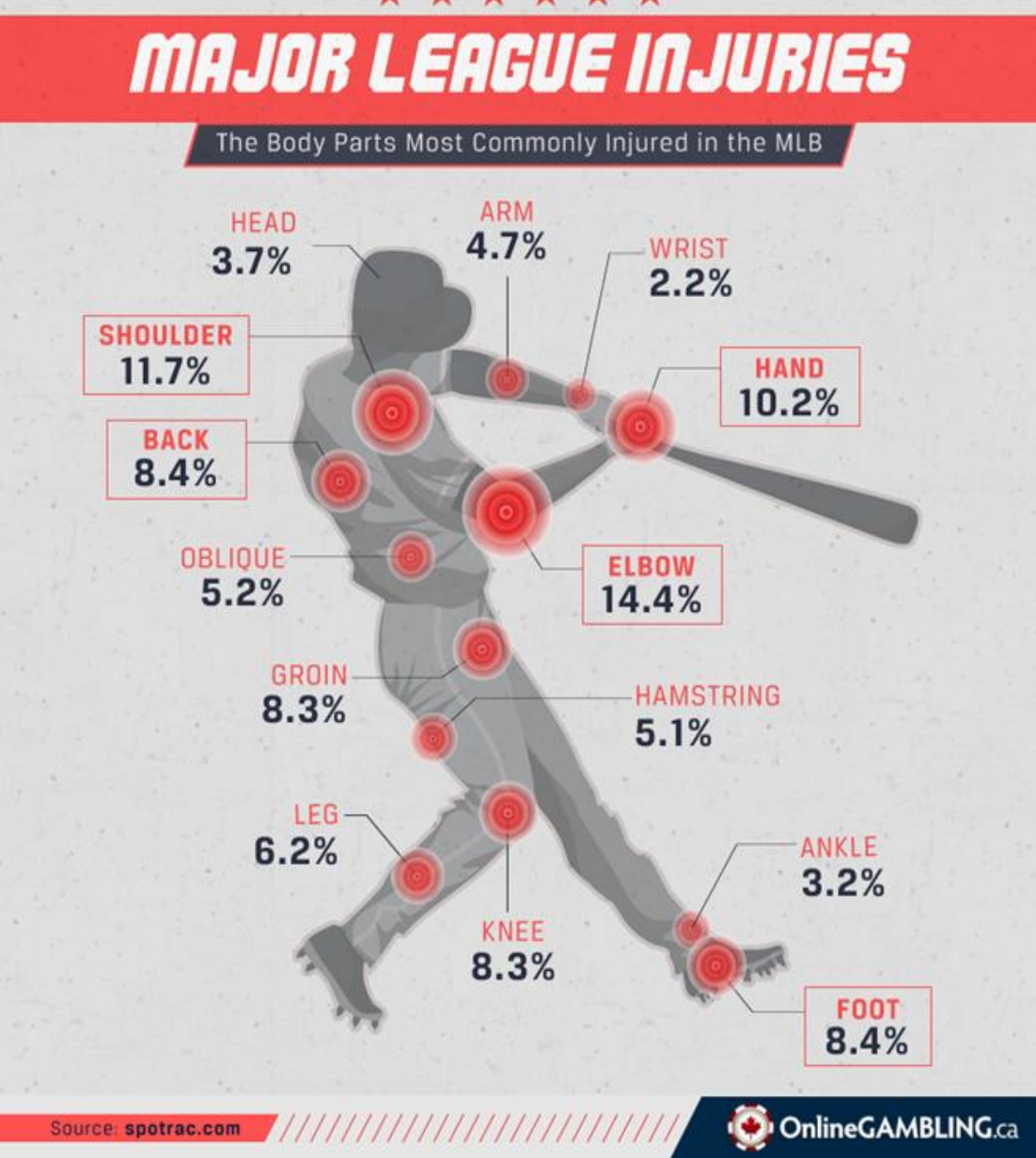
**Note:** Game Saved= Average Appearances Before Injury – Average Appearances Before Signal

**Analysis & Recommendation**

For Effective Monitoring: Using the  $h$  range of 7–12 provides a balanced approach for injury prediction. This is a **tradeoff between sensitivity and specificity**, where increasing the decision interval ( $h$ ) to be more specific also decreases the sensitivity.

**Why:** The cost of misclassifying a struggling pitcher as injured is low (e.g., imaging by a doctor between games).

**Strengths:** This approach to the solution is clear, with easy implementation.

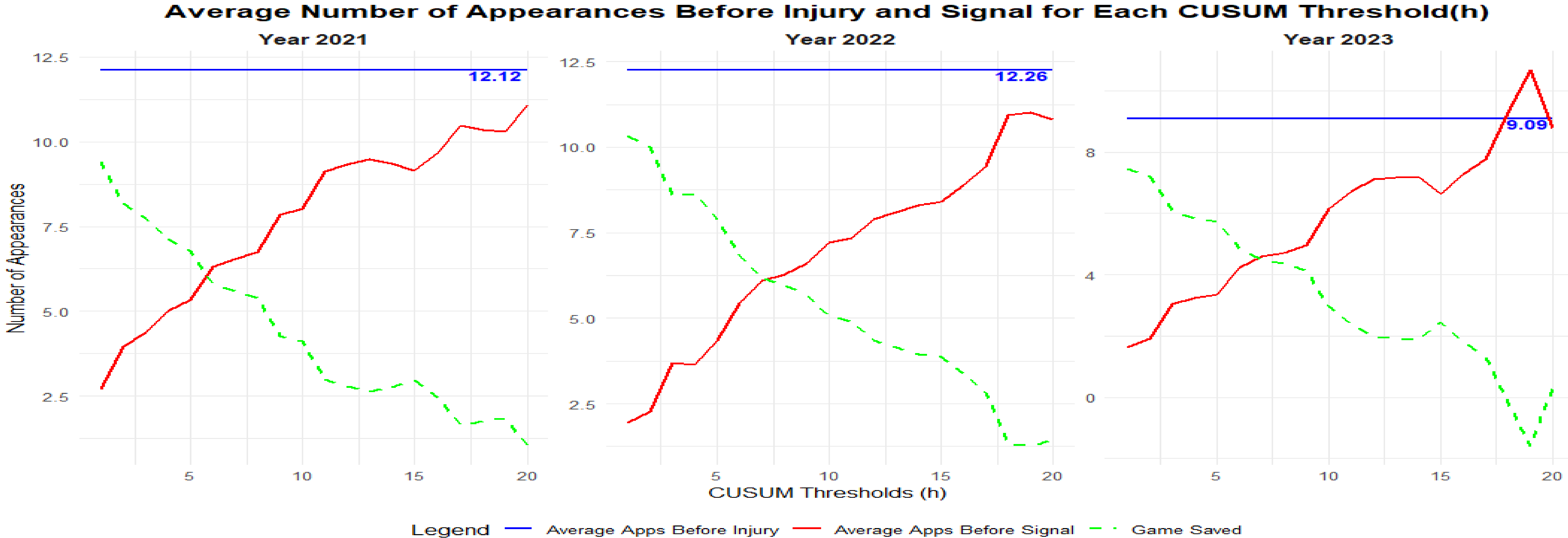


# How well does this approach predict re-injury risk?



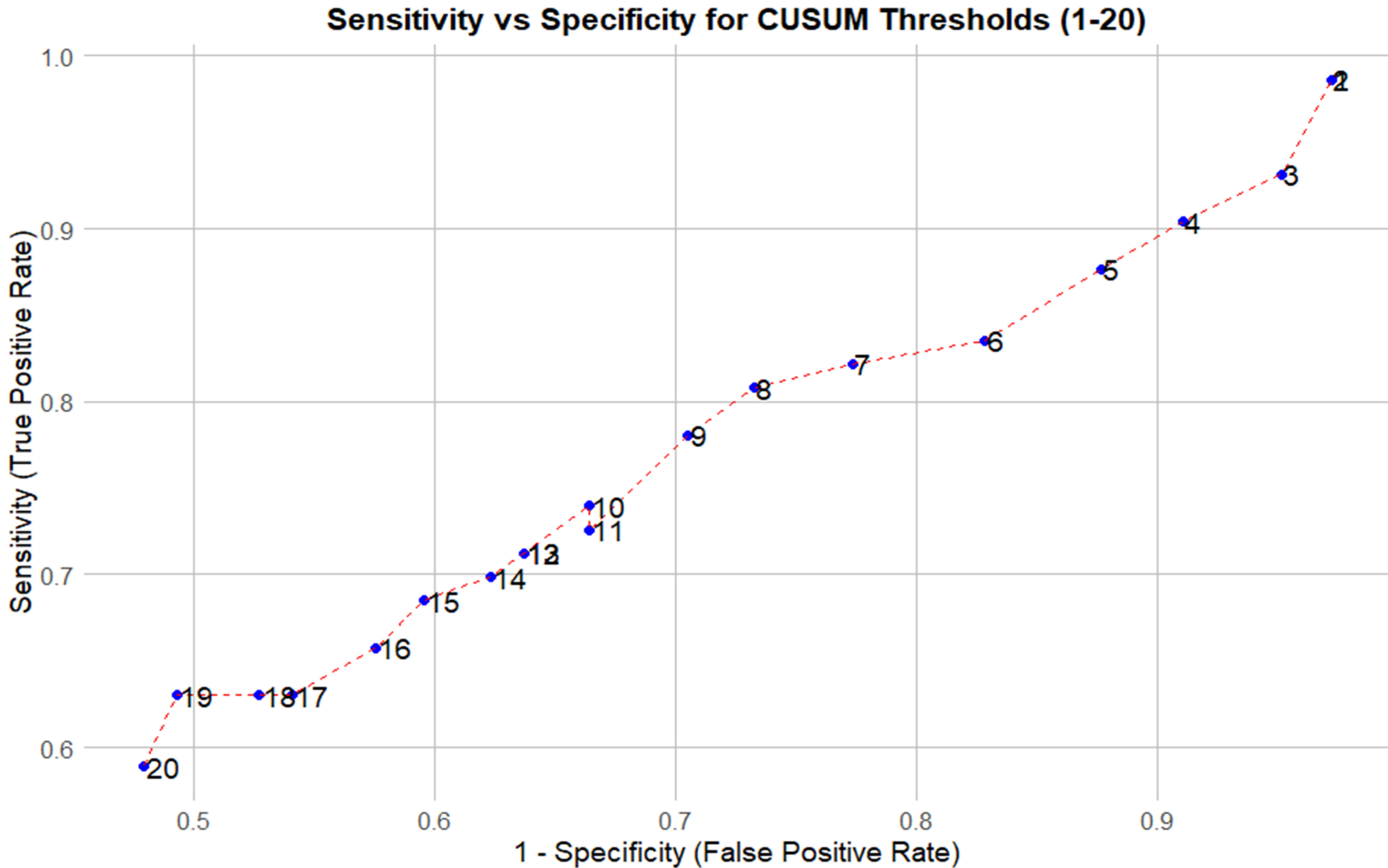
**KENNESAW STATE UNIVERSITY**  
COLLEGE OF COMPUTING AND SOFTWARE ENGINEERING  
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## A Quality Monitoring Approach to Evaluating Re-Injury Likelihood to MLB Pitchers



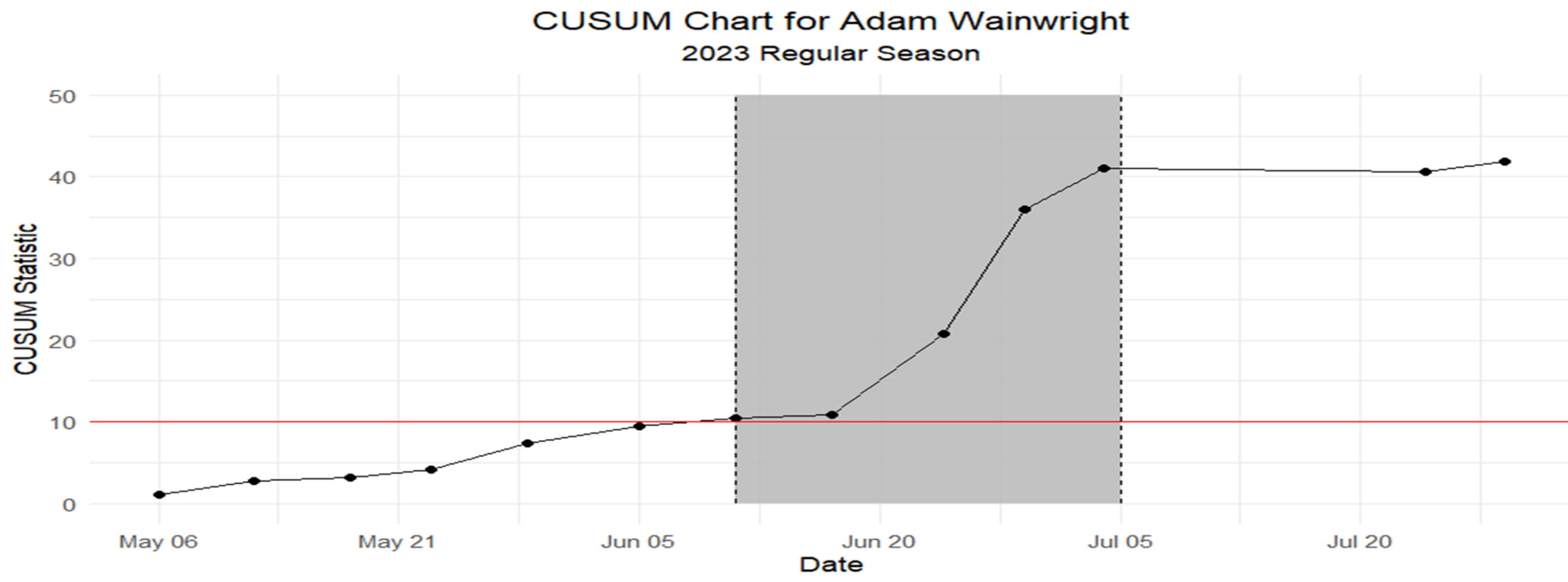
CUSUM Thresholds for Detecting Injuries

How different h-values affect detection rates.



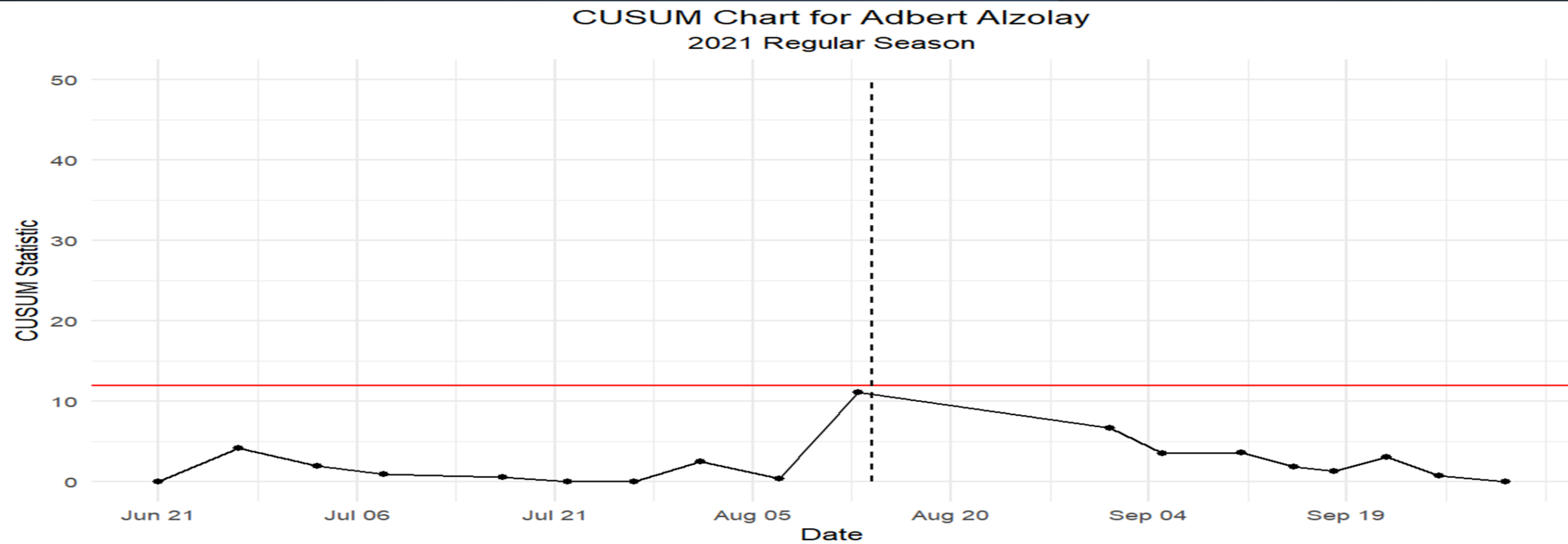
Case Study 1

A pitcher flagged by CUSUM before re-injury.



Case Study 2

A pitcher Never flagged by CUSUM After re-injury.



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