

# Modern tools to reduce errors in blood collection and transport

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 Washington University in St. Louis  
SCHOOL OF MEDICINE

# Conflicts of interest

## Research Funding from:

- > Abbott Laboratories
- > Roche Diagnostics
- > Beckman Coulter
- > Siemens Healthineers
- > Biomerieux
- > Sebia
- > Qiagen
- > Cepheid
- > Werfen

## Consultation for:

- > Abbott Laboratories
- > Werfen
- > Cytovale

No conflicts in the area of preanalytics!!!

# Learning Objectives



01

Assess the impact of preanalytical error on specimen quality and diagnostic accuracy

02

Implement a multi-disciplinary approach to improve specimen collection and handling through the preanalytical phase to improve patient outcomes

# Lab error occurs frequently and are mostly preanalytic

01/2022 – 03/2023

45,812,367

Billable Tests

~11,000,000

Specimens

87,317

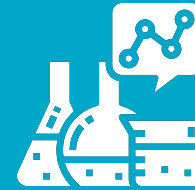
Errors

(0.79% of billable tests)



98.4%

Preanalytical  
Errors



1.1%

Analytical  
Errors



1.1%

Postanalytical  
Errors

# Can we capture and reduce these errors???

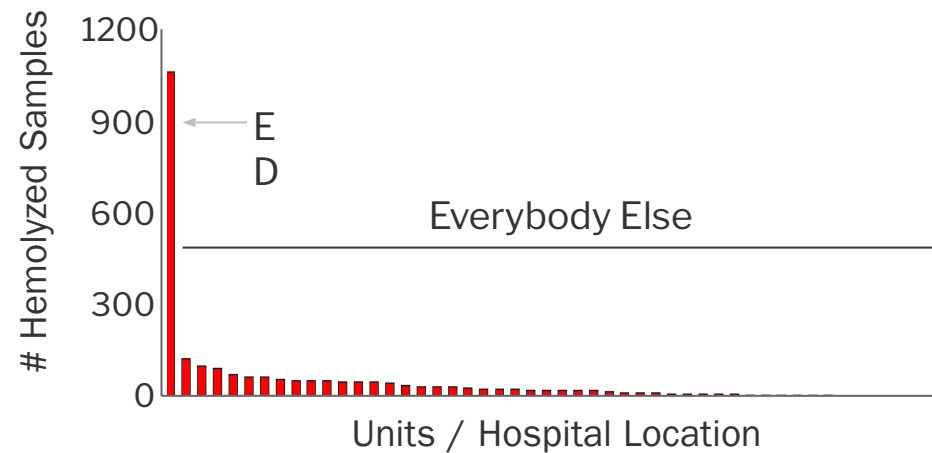
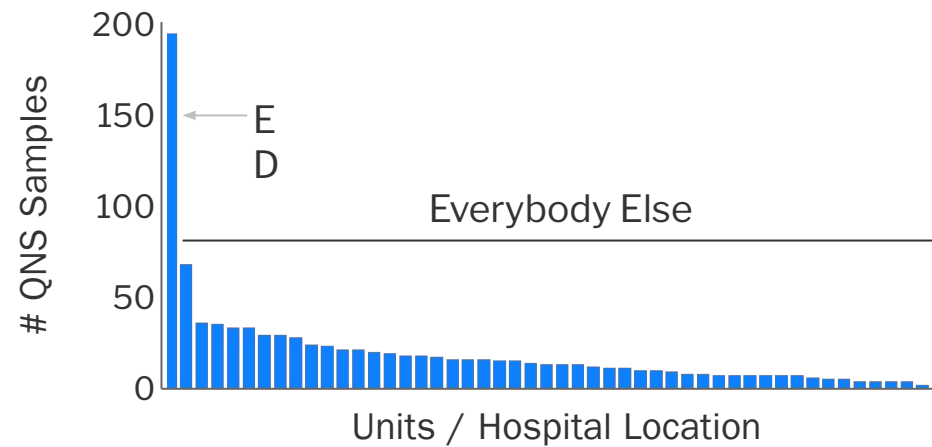
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Too old to test	550	0.6
Sample integrity	92	0.1
Requisition errors	62	0.1
Total	85,133	100



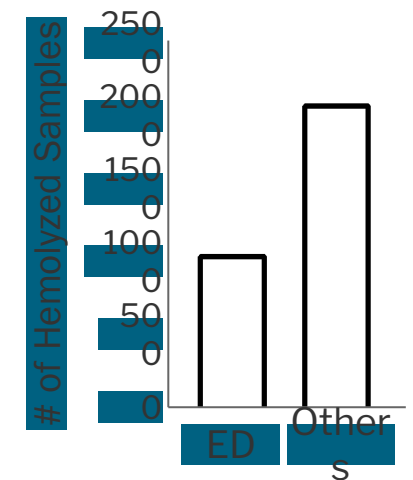
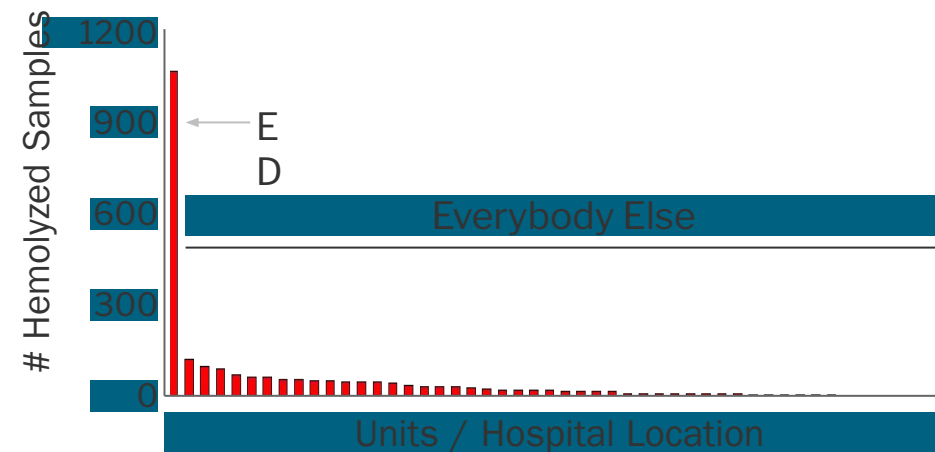
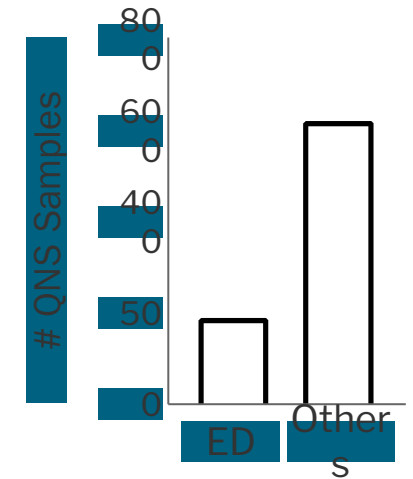
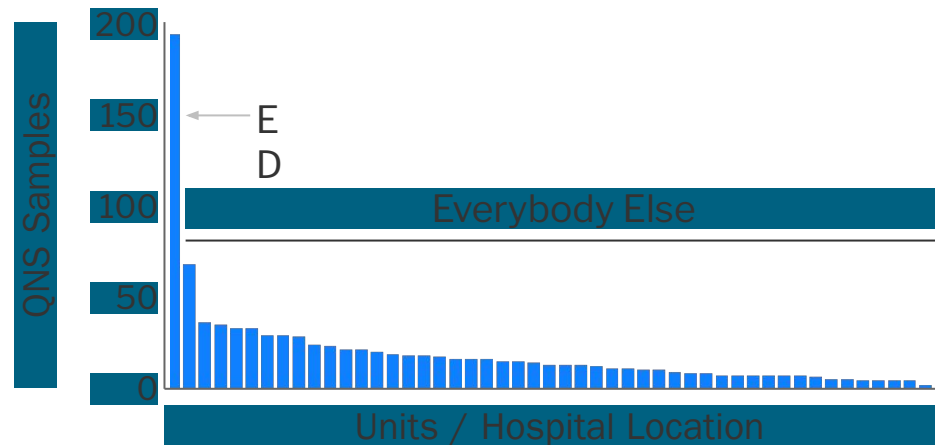
# I.

## Collection of Specimens from Phlebotomy

# Quantity not sufficient (QNS) and Hemolyzed Samples by Unit



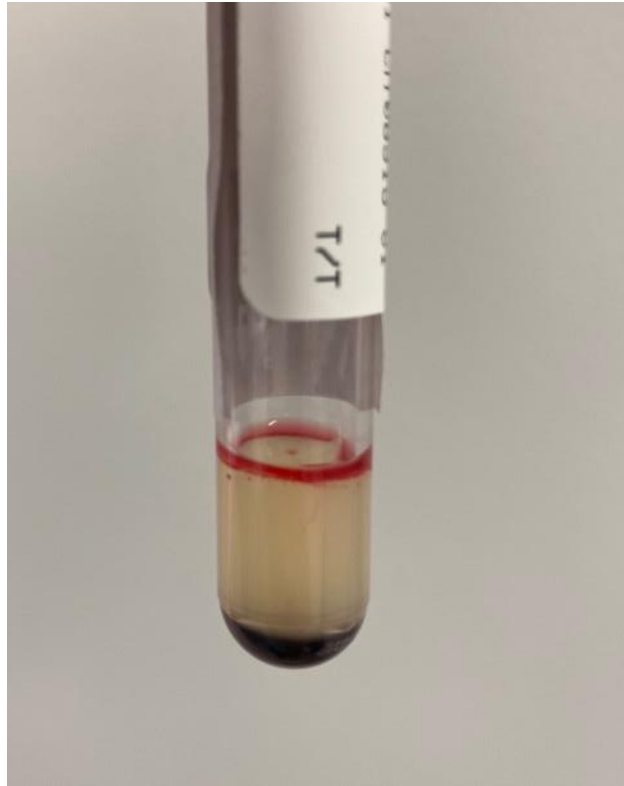
ED only  
accounts for  
25% of QNS and  
33% hemolyzed





# ED collaboration goal: Reduce QNS samples

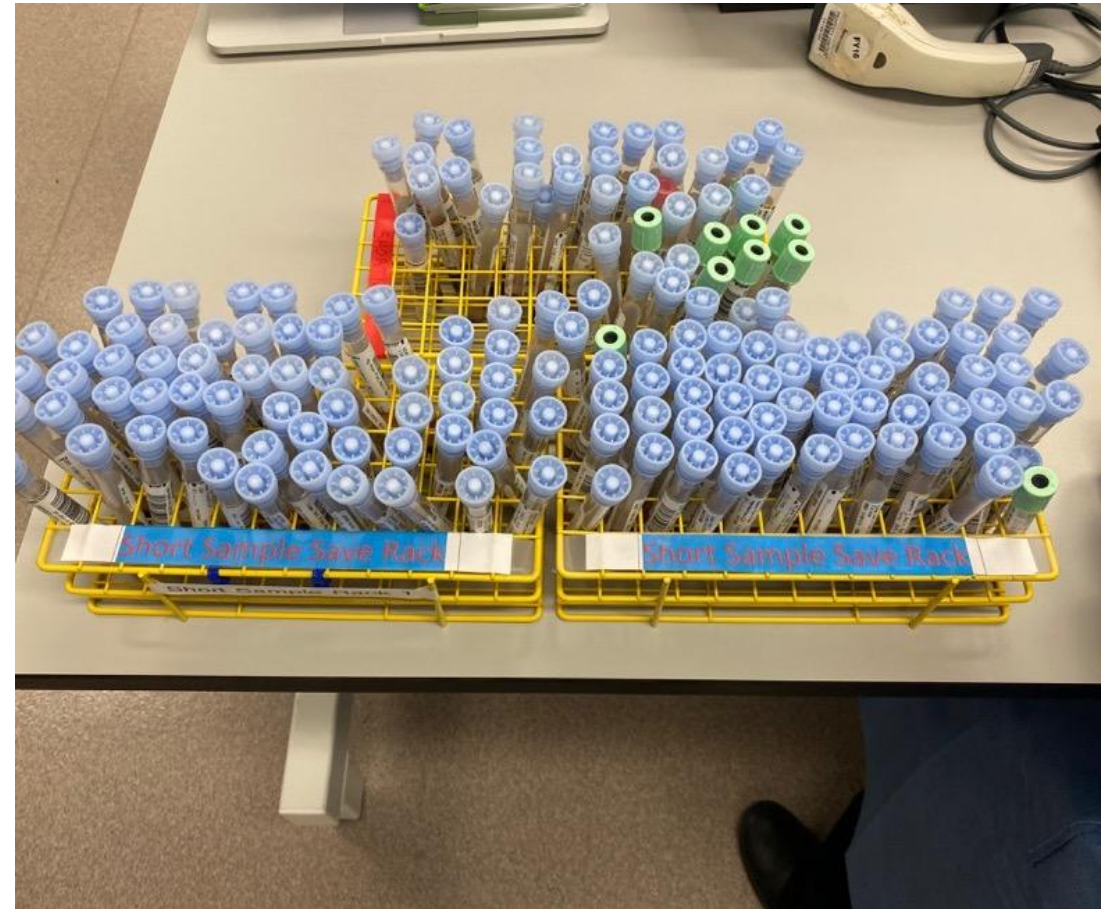
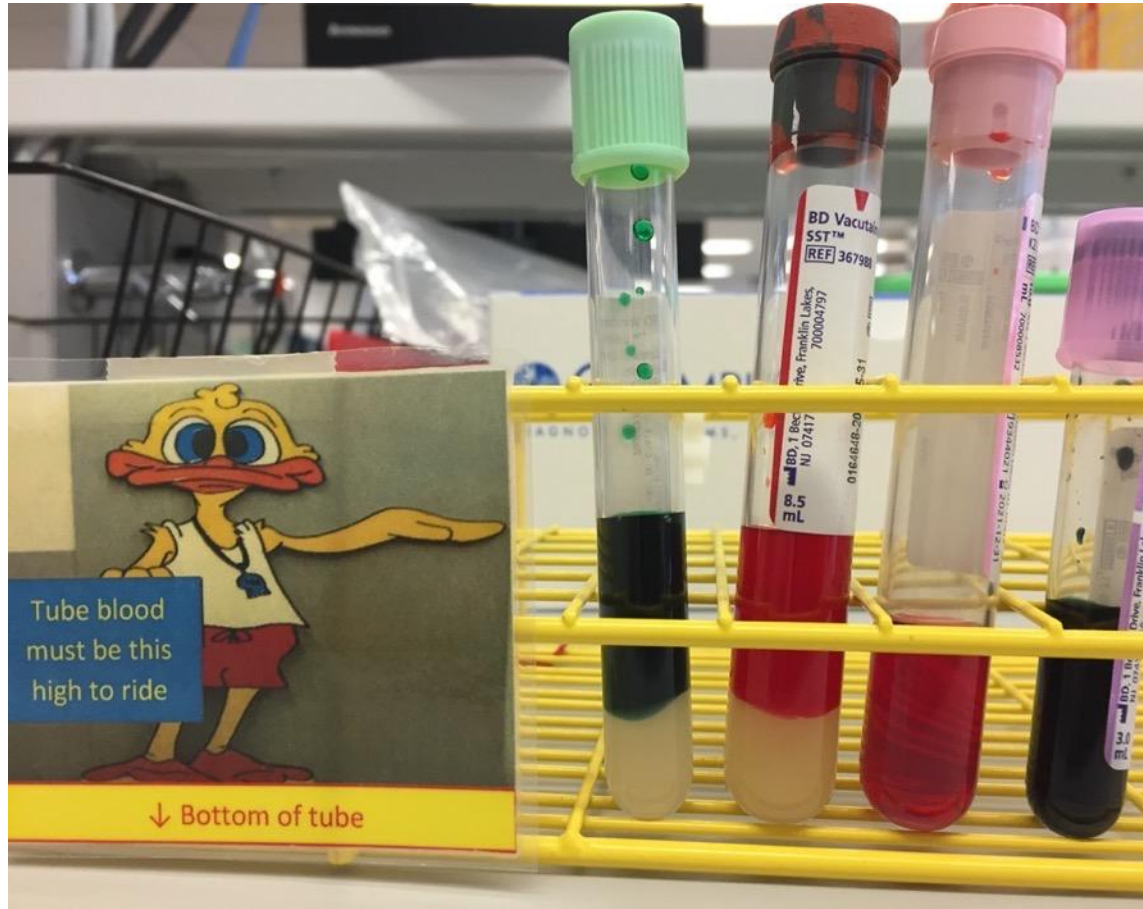
Of these



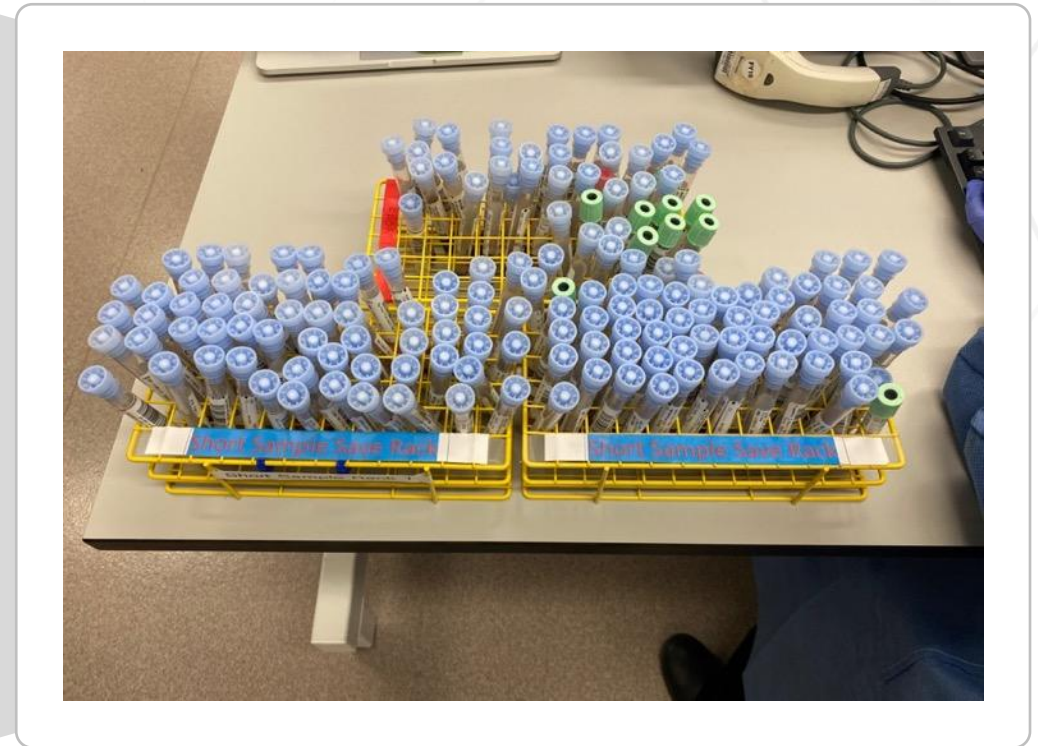
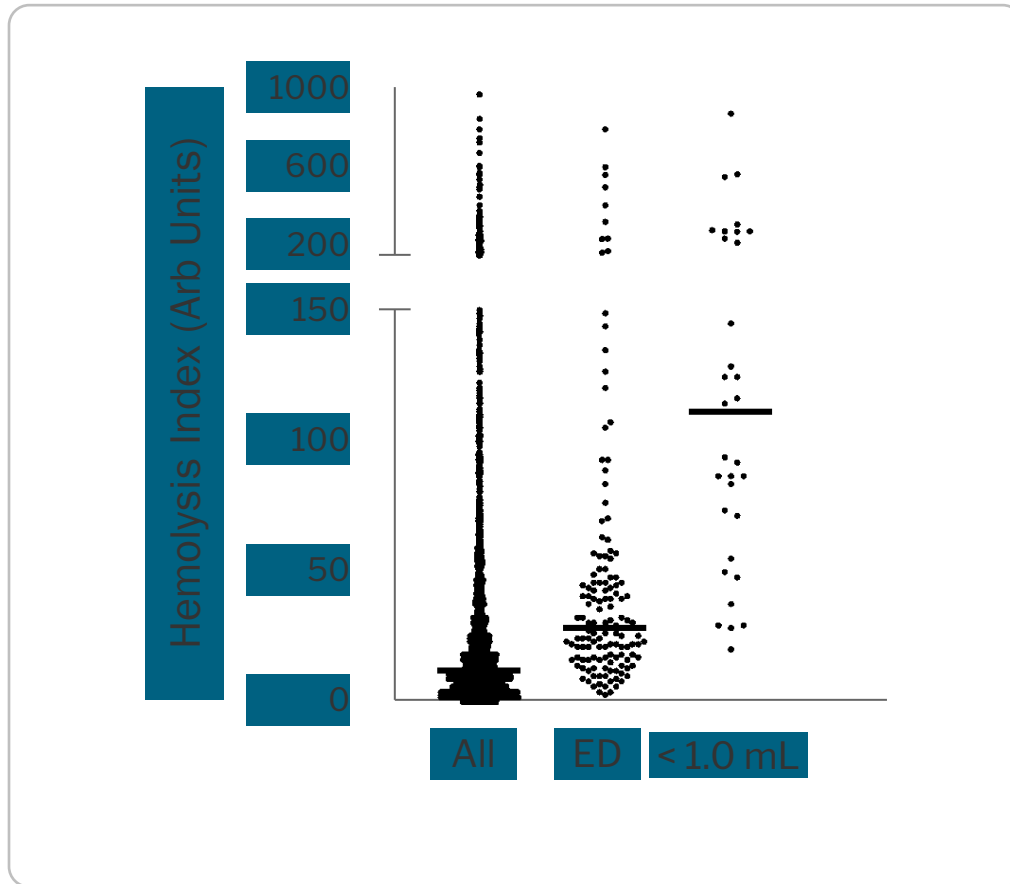
...and these



# QNS samples are common....



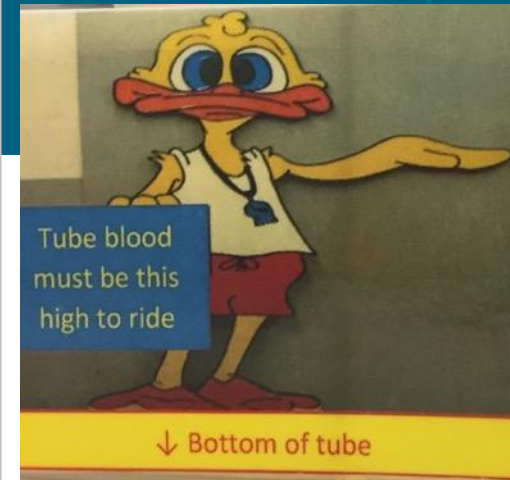
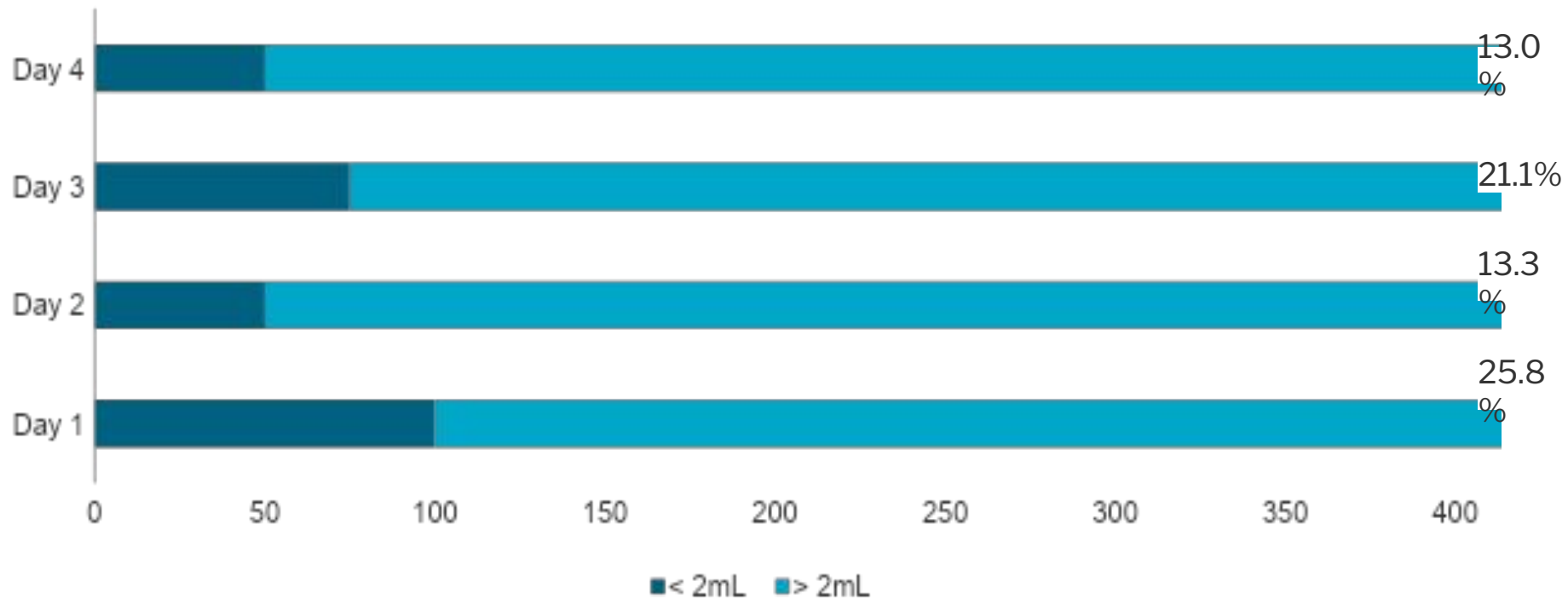
# QNS samples are more likely to be hemolyzed



Take Home: QNS samples = more hemolysis, more redraws

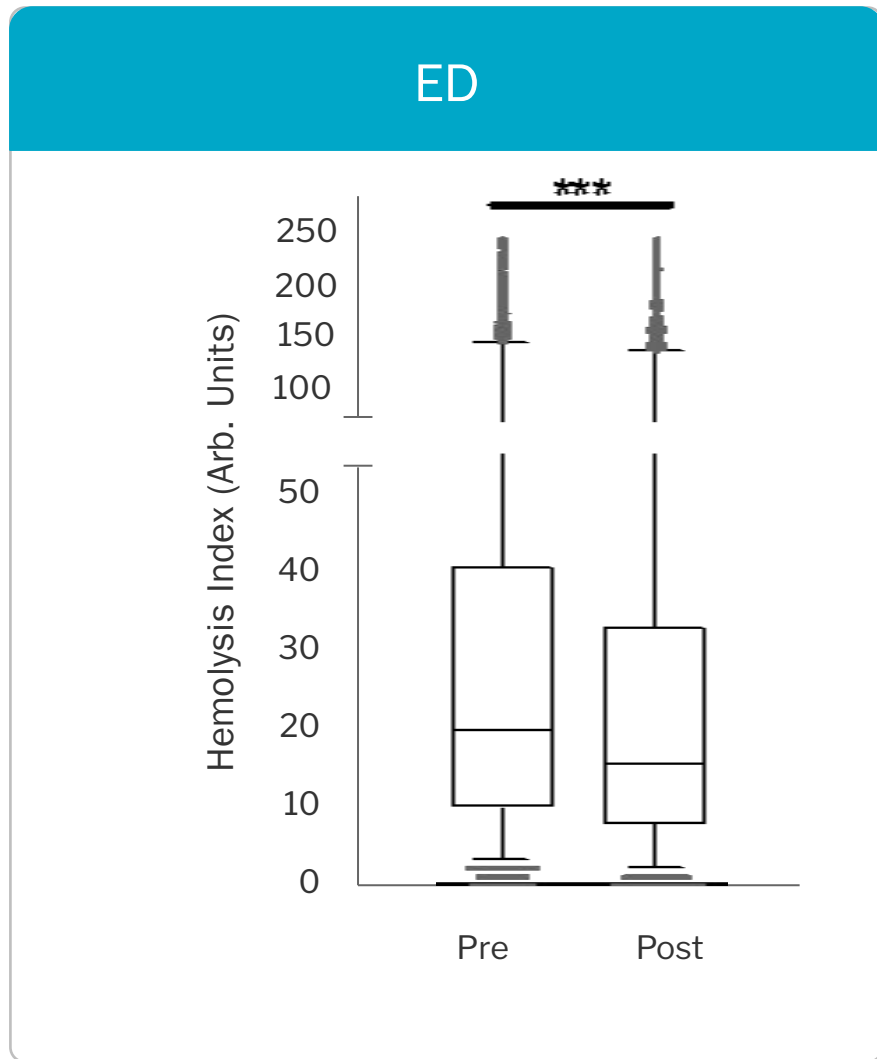
# Intervention- cancel all specimens < 2mL of blood

4 day grace period we measured and call back all samples < 2 mL to the ED

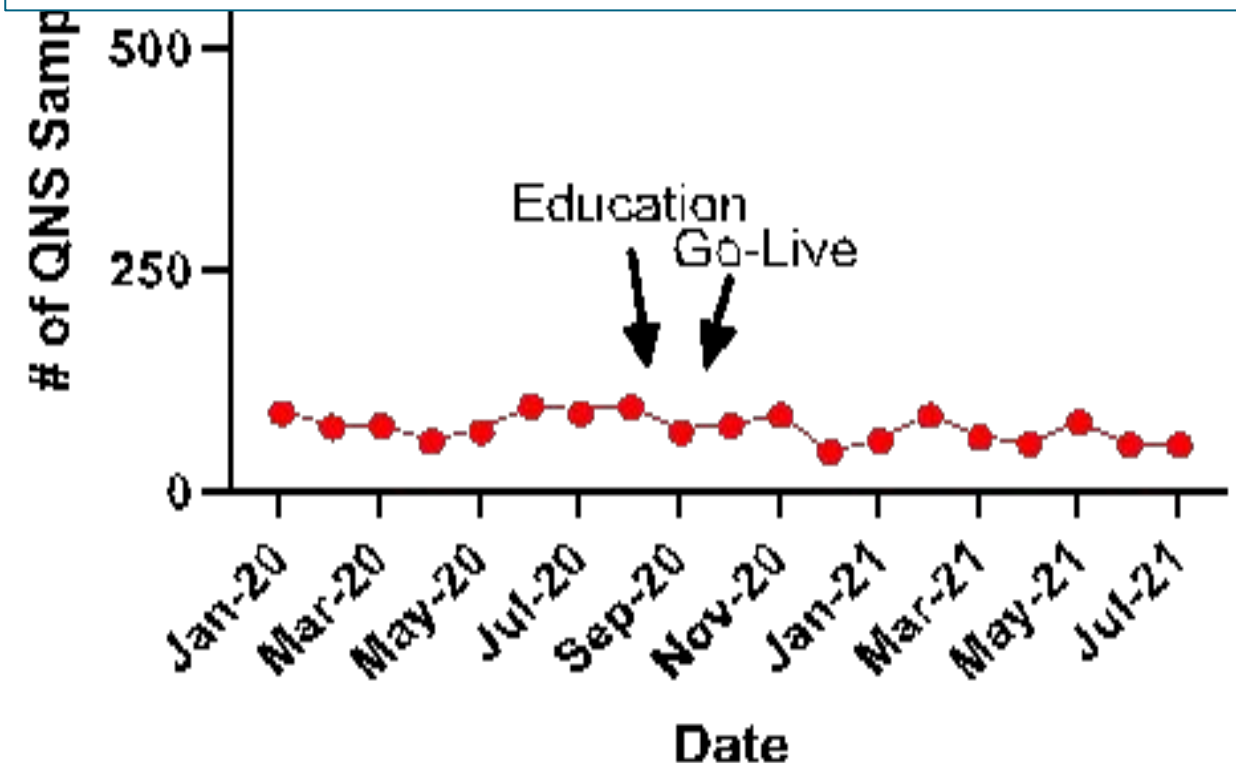




# Intervention reduced hemolysis, no change in QNS



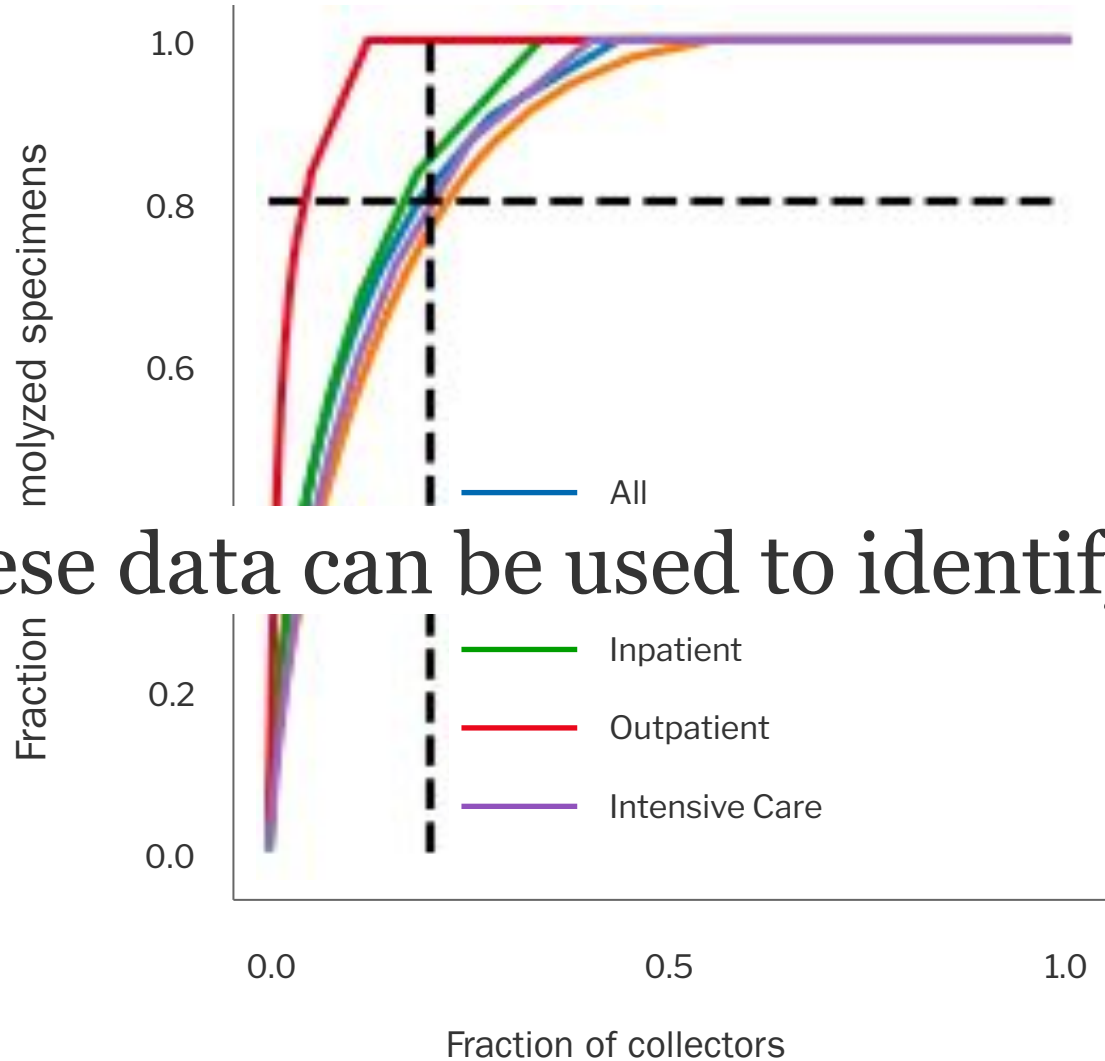
Shouldn't we be encouraging our staff to collect FULL Tubes?



# Use of positive patient identification to identify collectors



# 20% of collectors cause 80% of rejected samples

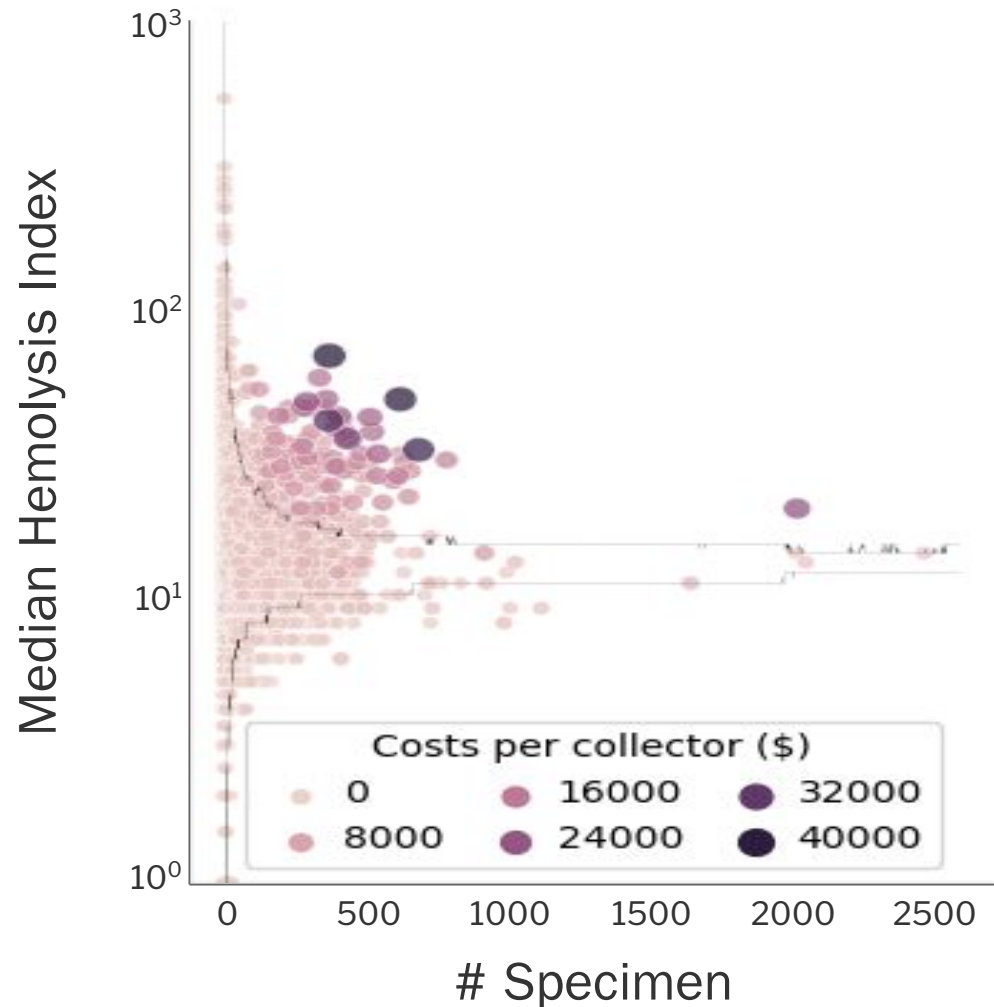


These data can be used to identify under performers!



Mark Zaydman, MD, PhD  
Assistant Professor, Wash  
U

# Statistical model predicts underperforming collectors



**Prediction accuracy (future 20%):**

Sensitivity = 71%

Specificity = 93%

Accuracy = 88%

**Other approaches:**

Most recollections

Average -Highest HI



Next steps:  
Automated  
feedback...

Barnes-Jewish Hospital Emergency Department

## Automated Collector Evaluations

### HEMOLYSIS FEEDBACK REPORT

Month: December 2022  
Collector: Mark A Zaydman (PPID: 1234567)

Total # of samples	82
# hemolyzed samples	42
Hemolysis rate	5%

Performance relative to peers

Average

Below Above

CLICK TO CONTINUE

# Takeaways: Mitigating and detecting hemolysis

01

Hemolysis is a major cause of preanalytical error

- Consider working with nursing and ED to establish methods to detect collection underperformers

02

QNS rates may considerably impact TAT and hemolysis



# II.

## Specimen Transport

# Specimens are transported by pneumatic tube systems (PTS)

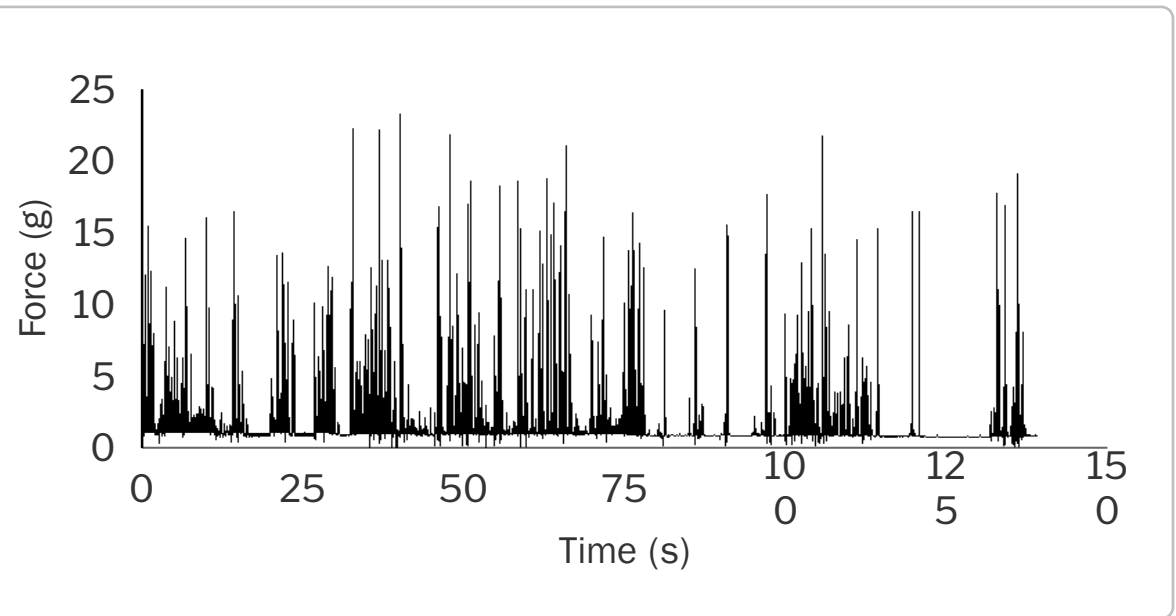
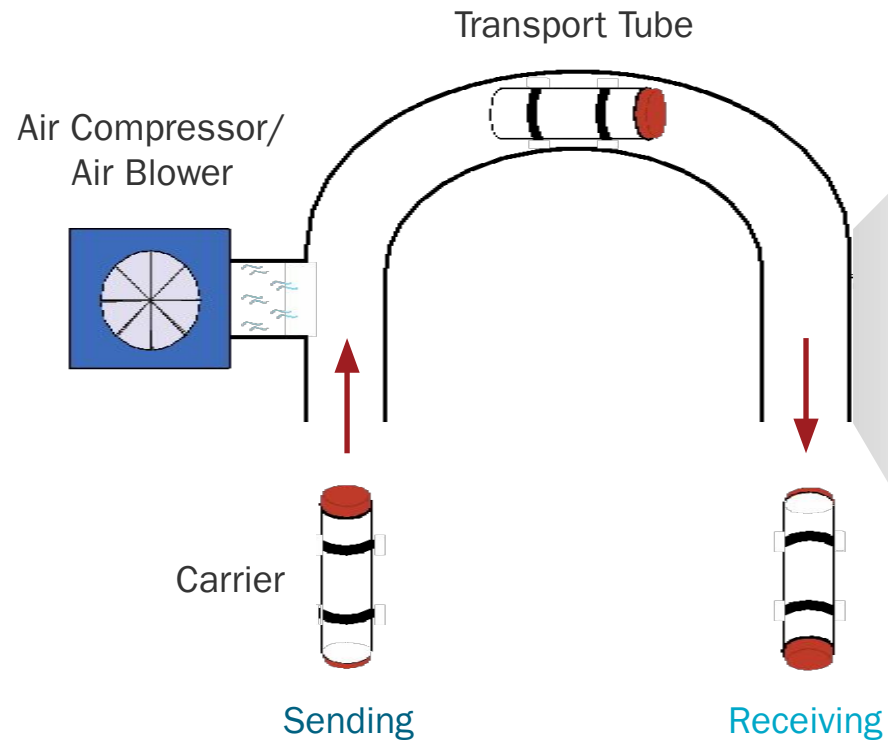


Invented in 1850's to transport telegraphs

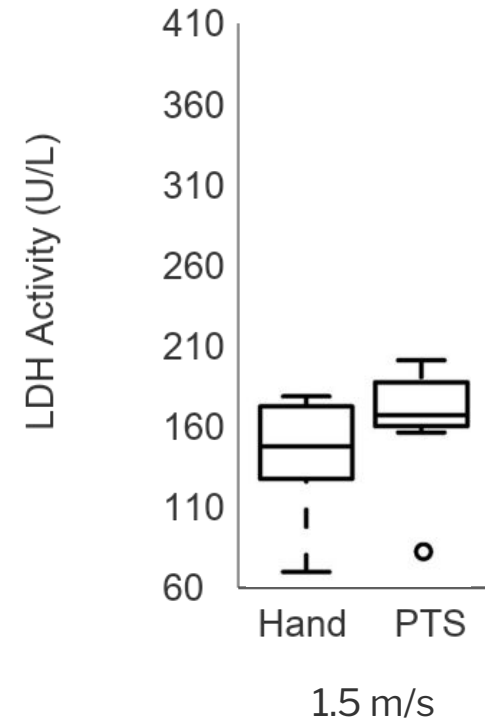
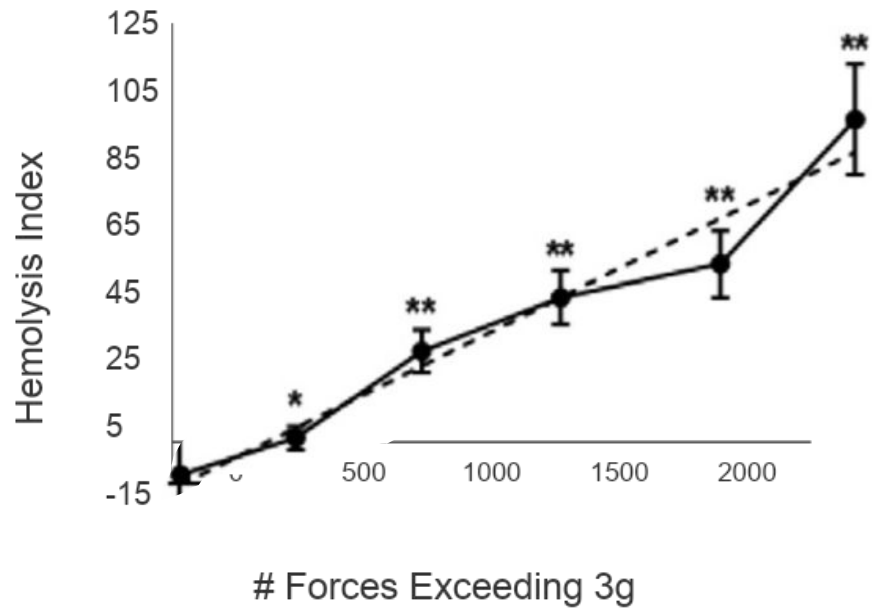
PTS provides rapid transport of patient specimens to laboratories

Reduce turnaround time by ~10 minutes

# PTS generates extreme accelerations during transport

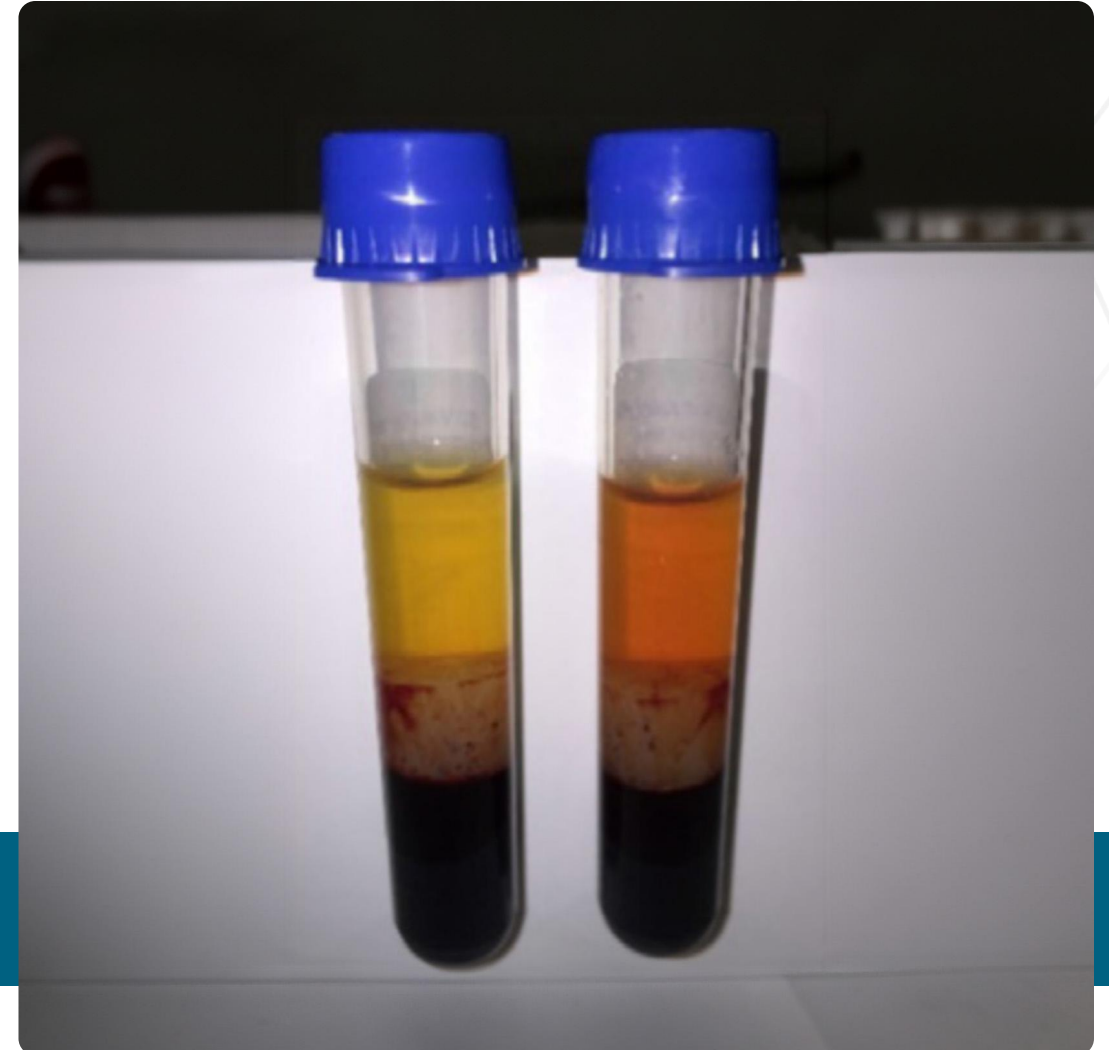


# Specimen transport by PTS increases hemolysis



# Analytes impacted by PTS transport- mostly intracellular

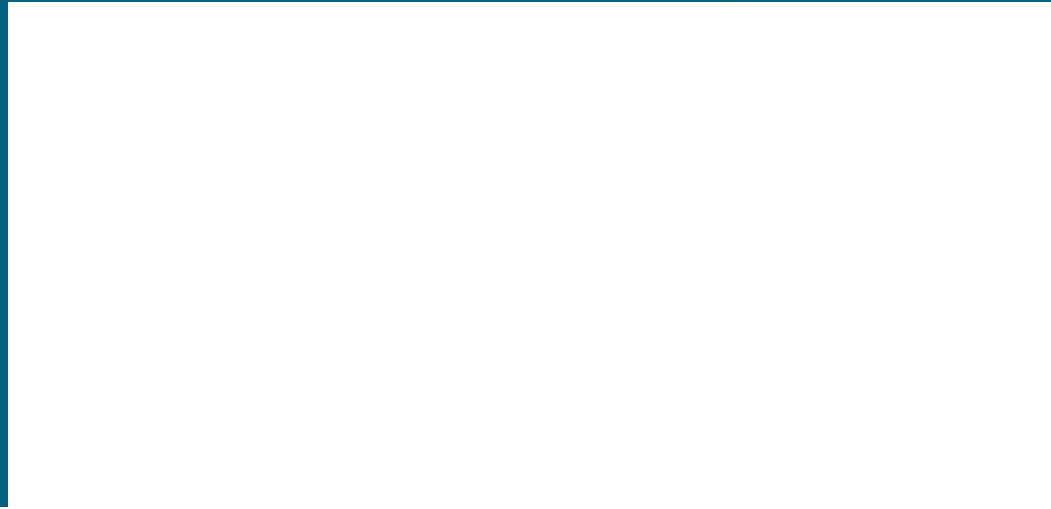
	Mean (N = 30)	
	Control	Trans
Na (mmol/liter)	141.1	140.1
Cl (mmol/liter)	103.9	103.8
CO2 (mmol/liter)	24.5	24.4
Ca (mg/dL)	9.59	9.58
P1 (mg/dL)	3.06	3.08
Bilirubin (mg/dL)	0.54	0.54
Uric acid (mg/dL)	6.20	6.19
K (mmol/liter)	4.45	4.56
Hb (mg/100 ml)	4.76	12.64
LDH (U/liter)	98	148



And known to be impacted by pneumatic tube transport!

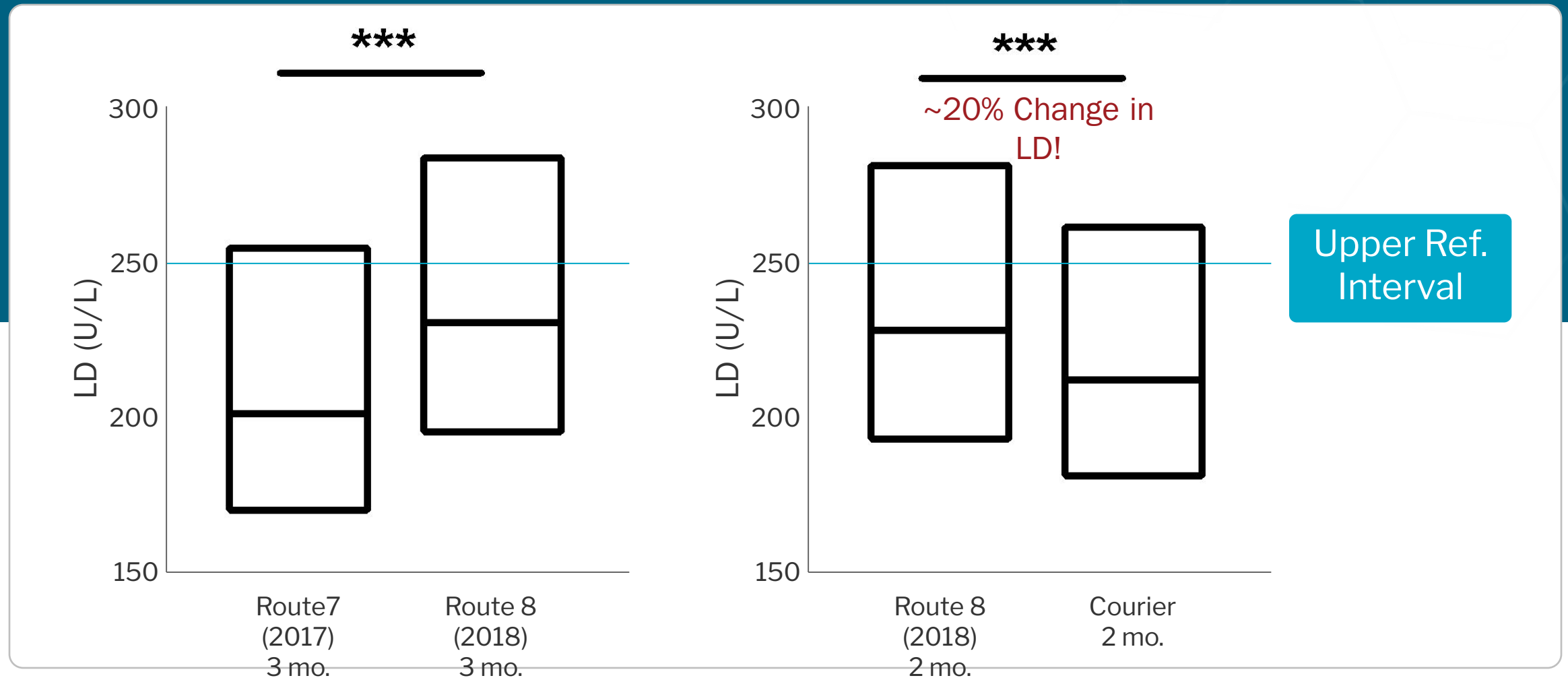
Steige H *et al.* Clin Chem. 1971; 17:12.  
Mullins GR *et al.* Clin Chimica Acta. 2016; 462.  
Nybo M *et al.* Clin Chem. 2018; 64:5.

# Typical PTS study design: Compare paired samples





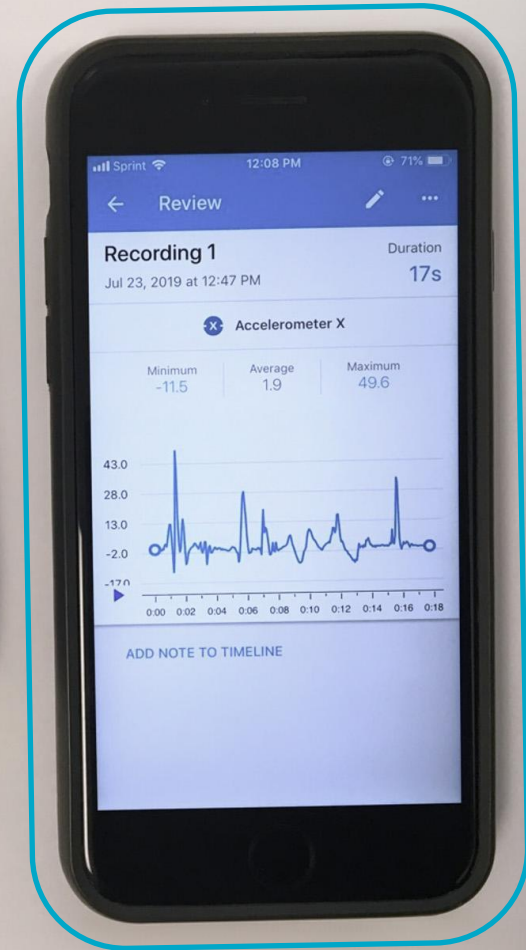
# Leveraging your data to assess PTS performance



# 3-axis accelerometers measure forces from PTS transport

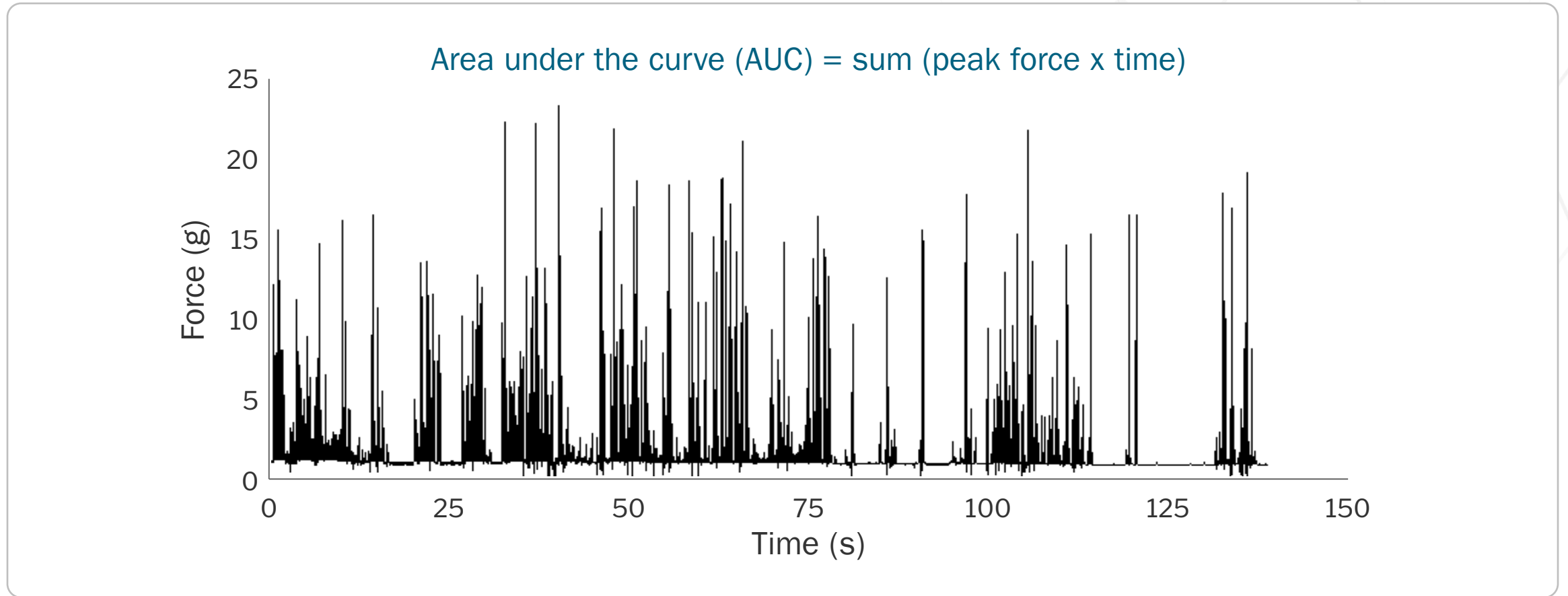


Accelerometer



iPhone 6s running  
Google Science Journal

# Dataloggers monitor the number of accelerations over time



Start of run



End of run

# Validation method



- Collect 5 tubes of blood
- Healthy subjects
- 15 total subjects
- 3 different days
- 2 different routes

1 tube walked to lab  
(control)



Remaining  
specimens ship  
through PTS



Remove one specimen



Remaining  
specimens ship  
through PTS

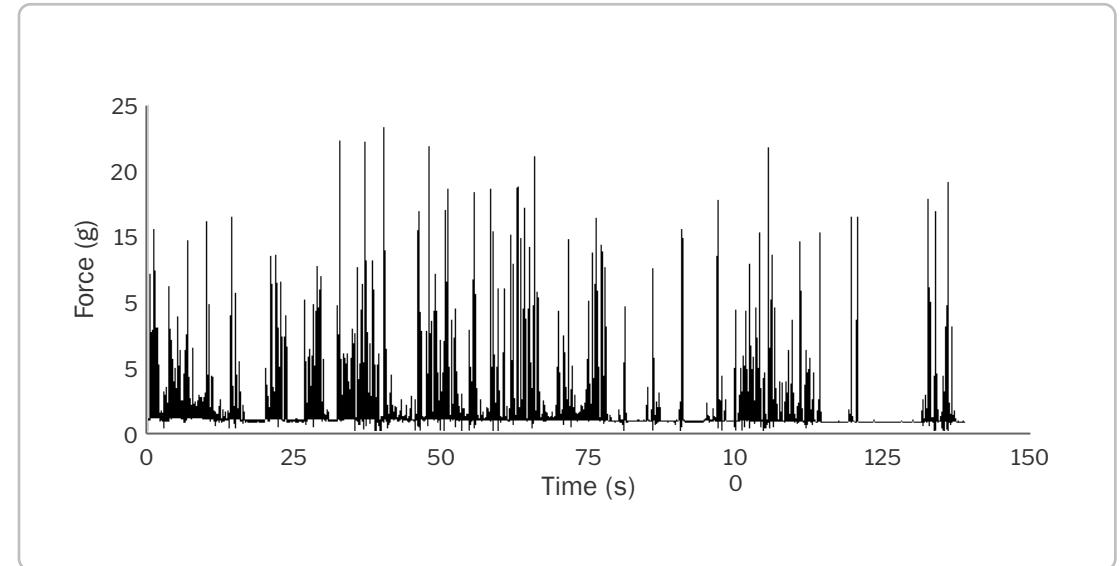
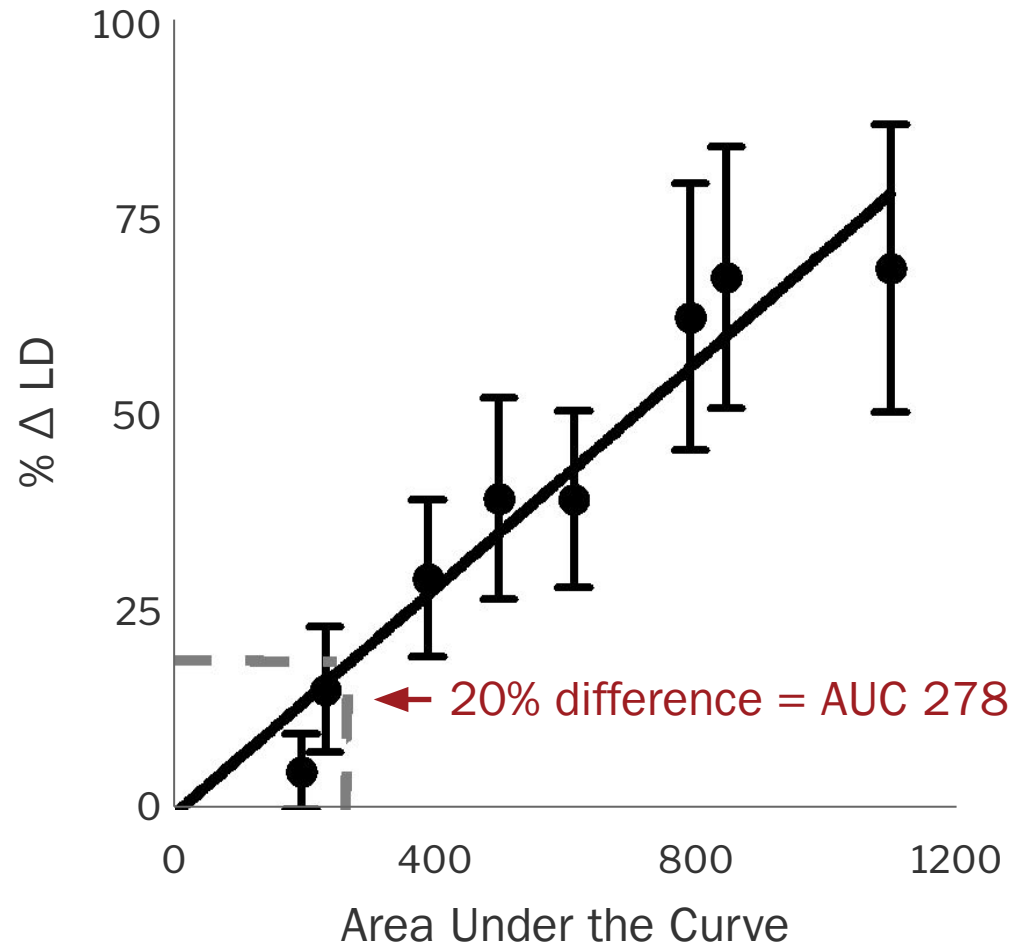
..... Up to 4 times  
X 2 routes



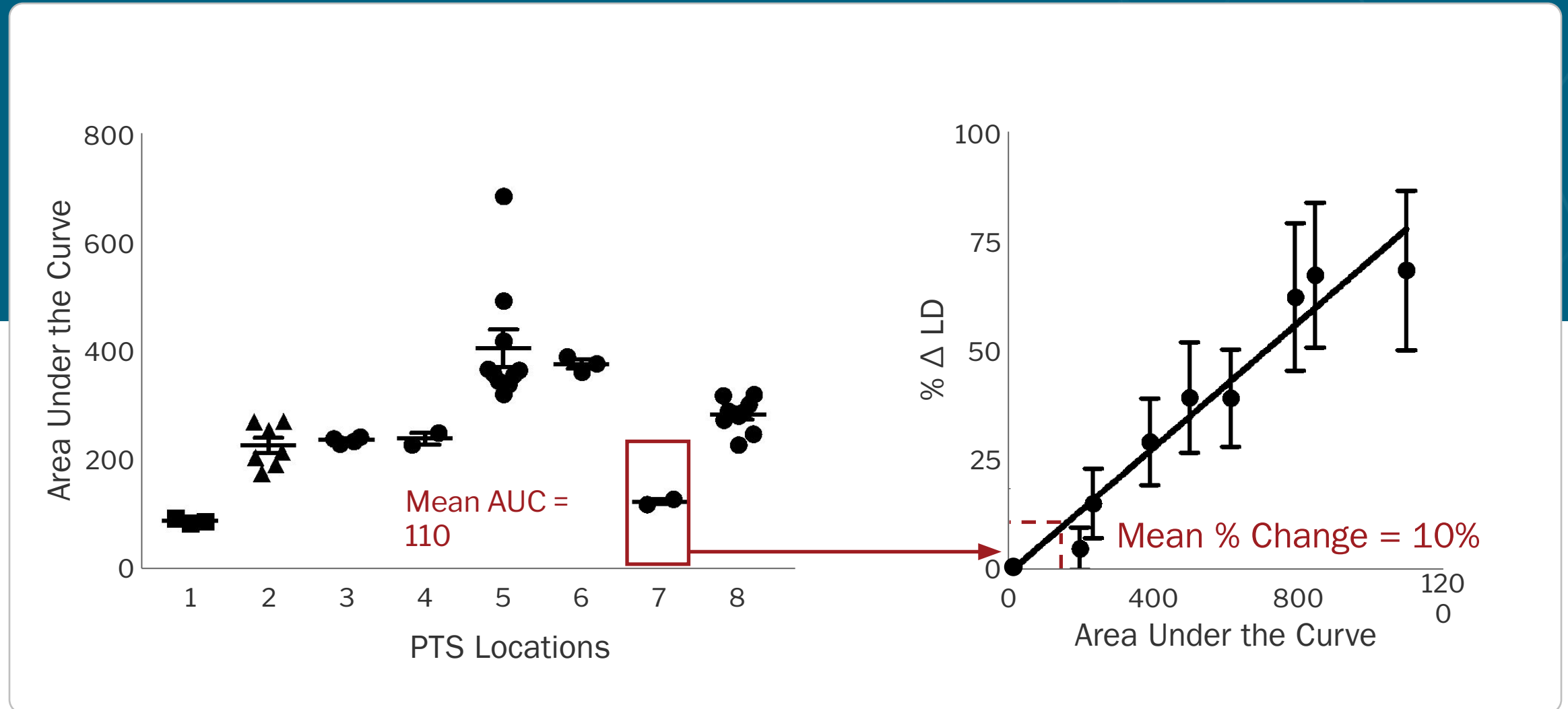
Analyze K<sup>+</sup>, LD, and HI

**OR ANY ANALYTE OF INTEREST**

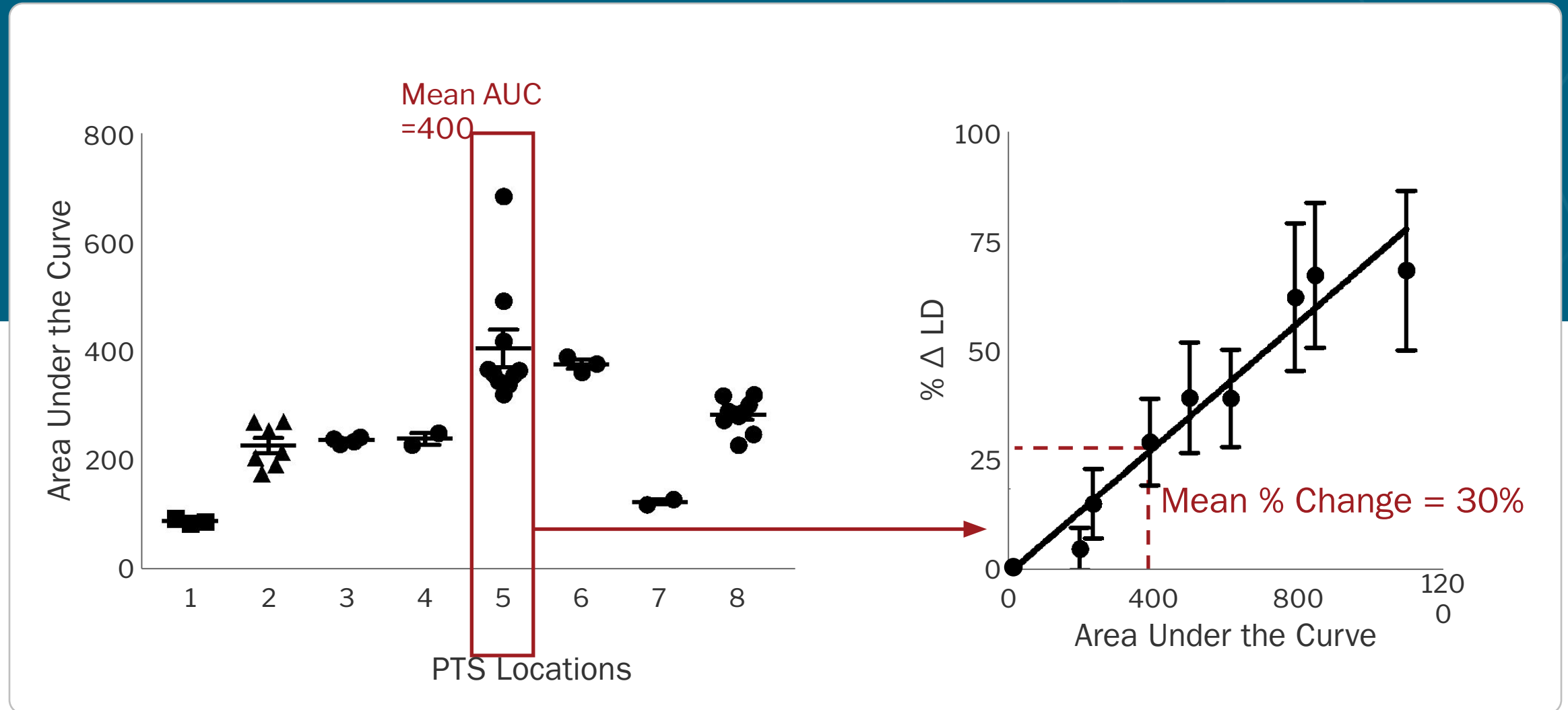
# Correlating PTS parameters with change in LD



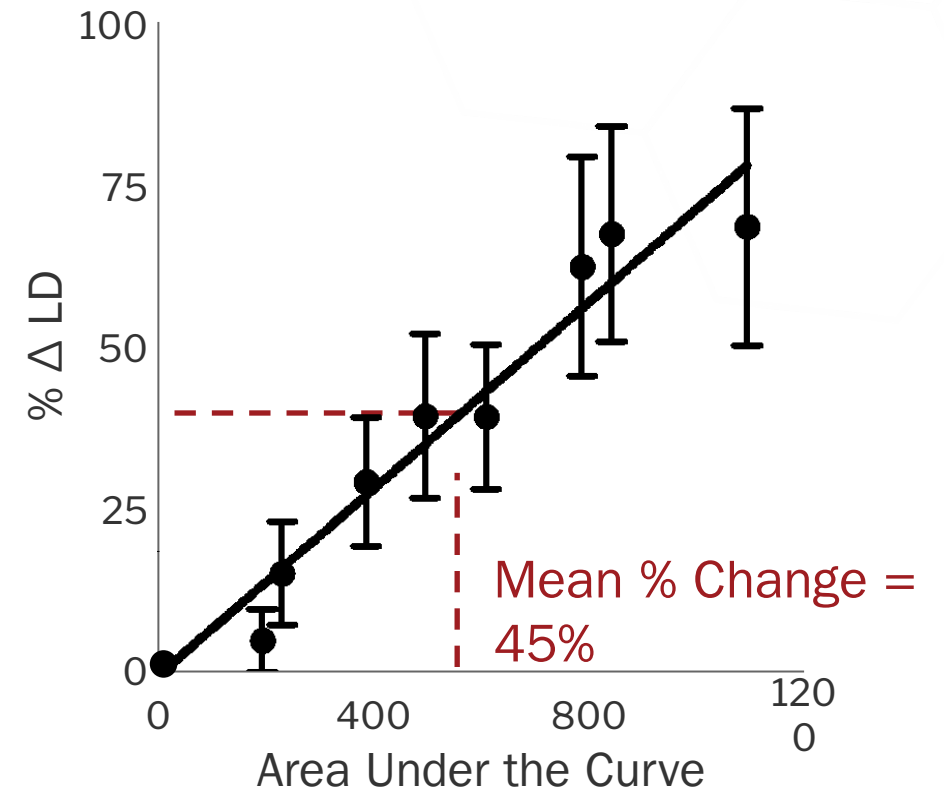
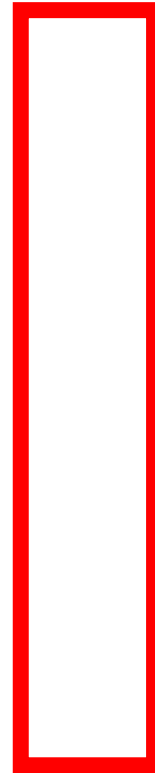
# Correlating PTS parameters with change in LD



# Correlating PTS parameters with change in LD



# Using datalogger to assess new PTS routes





# Takeaway: laboratories should consider assessing the PTS

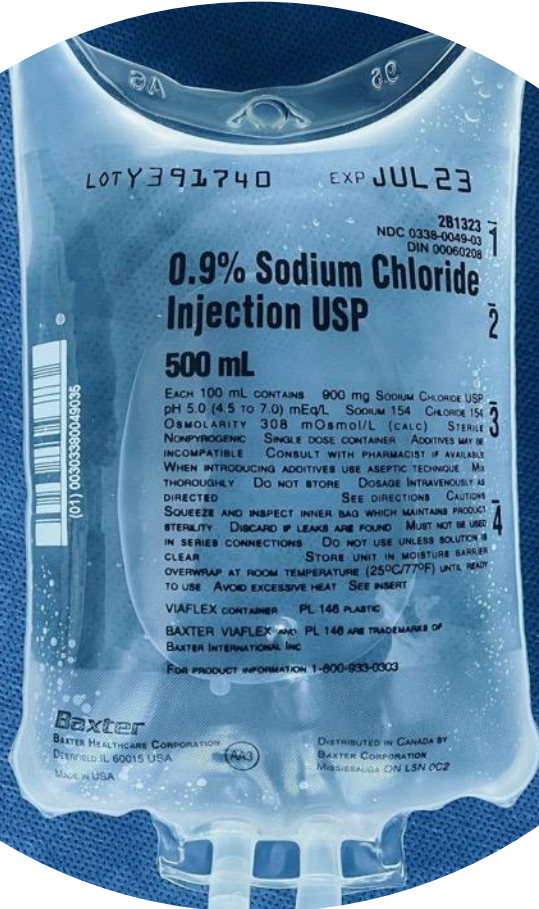
01

## Methods to assess for PTS performance

1. Paired specimens (Walked and sent through PTS)
2. Use of retrospective data from your laboratory
3. 3-axis accelerometers

# III.

## Detecting Contamination from Intravenous Fluids (IVF)



# How frequent is IVF contamination?

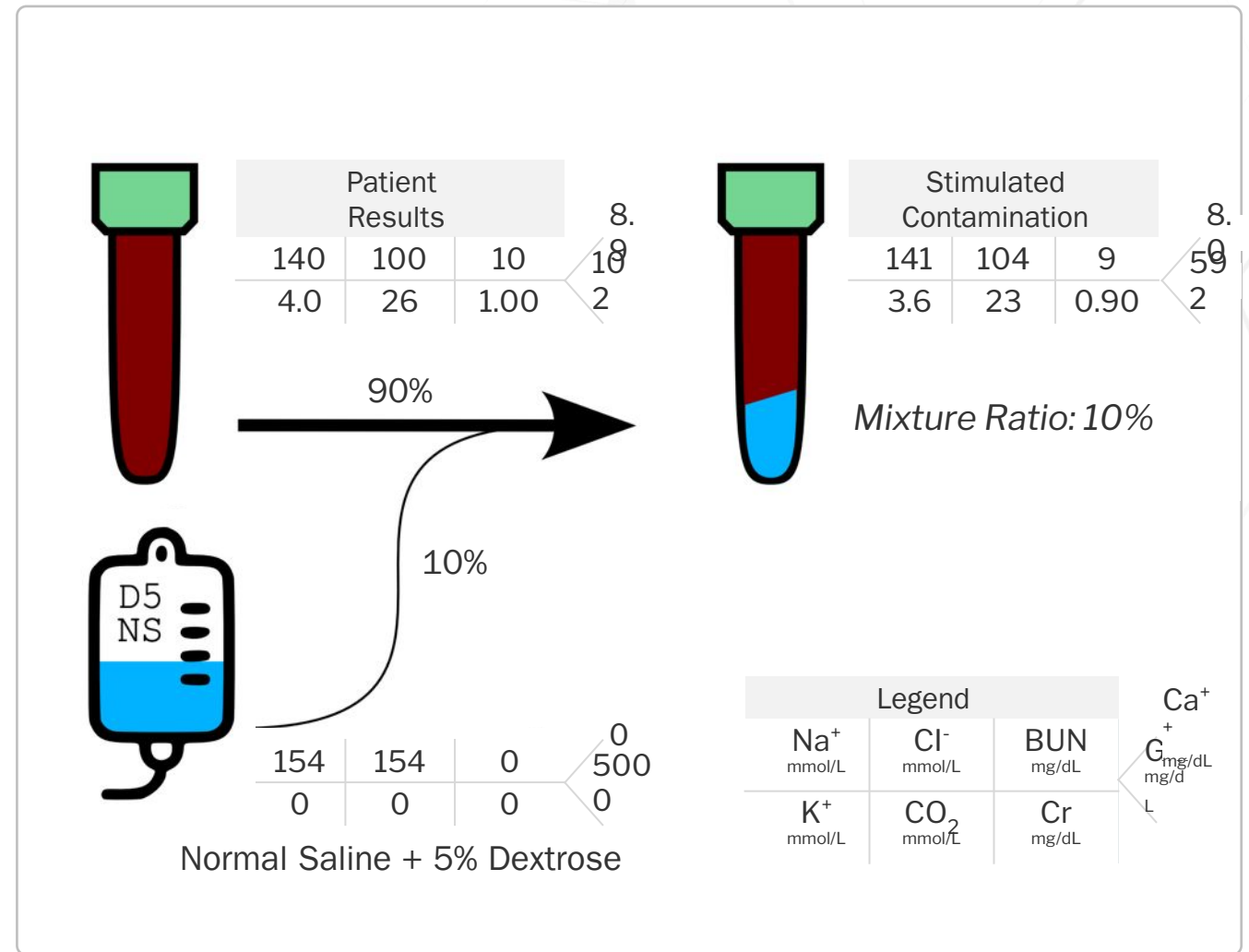
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Not on ice	1,369	1.6
IV contamination	1,122	1.3
Too old to test	550	0.6
Total	85,133	100

# How hard is it to detect IVF contamination?

## Study Design

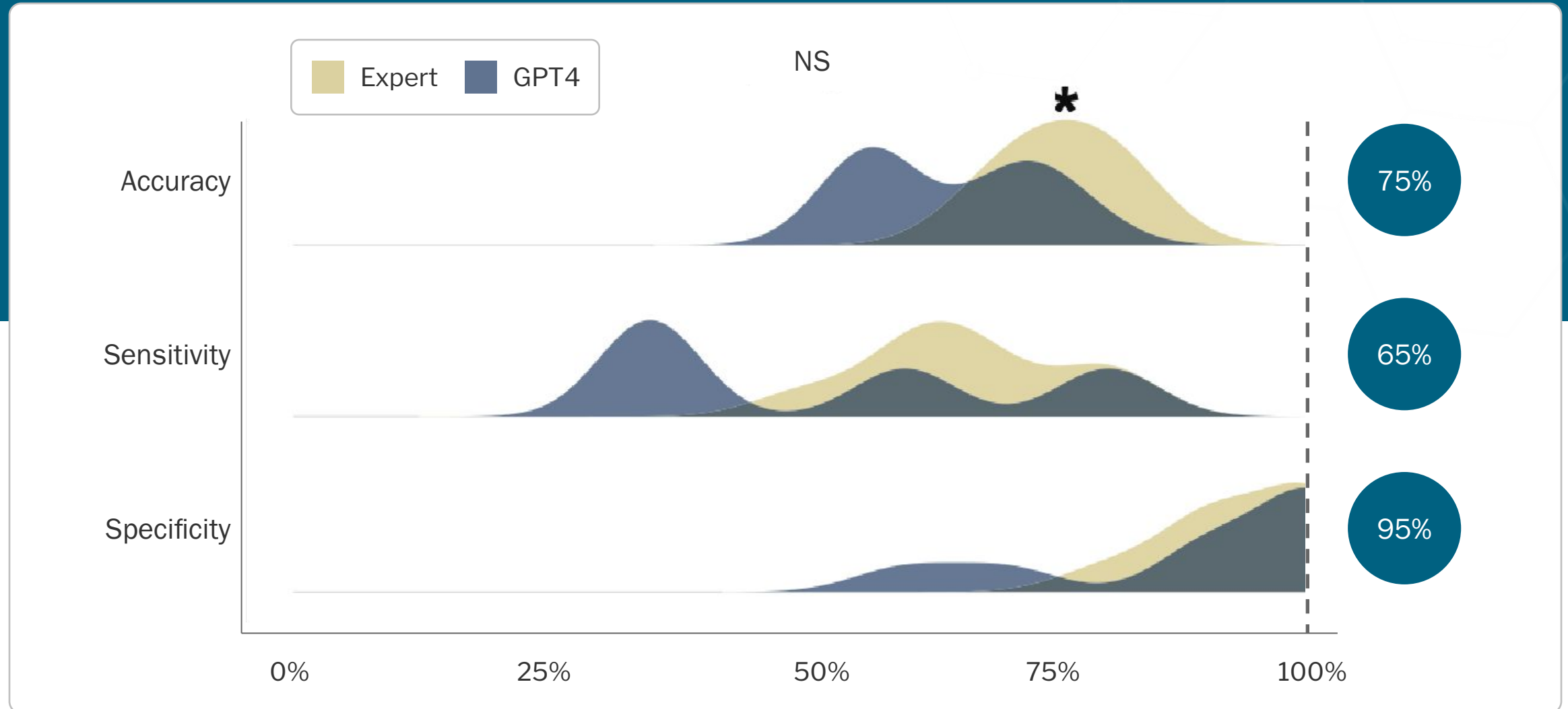
- 2 PhD Directors
- 2 PhD Clinical Fellows
- 1 MD Lab Medicine Residents
- 2 MD Internists
- 1 Laboratory Technologist
- GPT-4 (AI based Large Language Model)

Gave them results from 60 Basic metabolic panels



Are they contaminated??

# Humans are bad at detecting IVF contamination



# Delta checks are commonly used to distinguish potential error

Table 1. Selected Comparison Delta values and Repeat Criteria for Delta Check

Test	Criteria for repeat (differences between consecutive results on the same patient)
Albumin	~ 15 g/liter
Calcium, total	~0.25 mmol/liter (1.0mg/dl)
Potassium	~2.0 mmol/liter and no hemolysis
Protein, total	10 g/liter
Sodium	~20 mmol/liter



CLIN. CHEM. 21/11, 1648-1653 (1975)

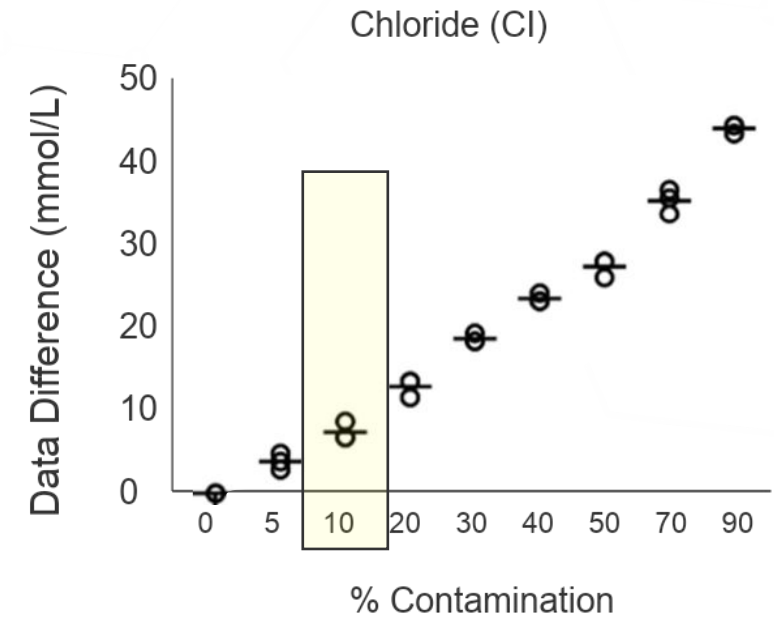
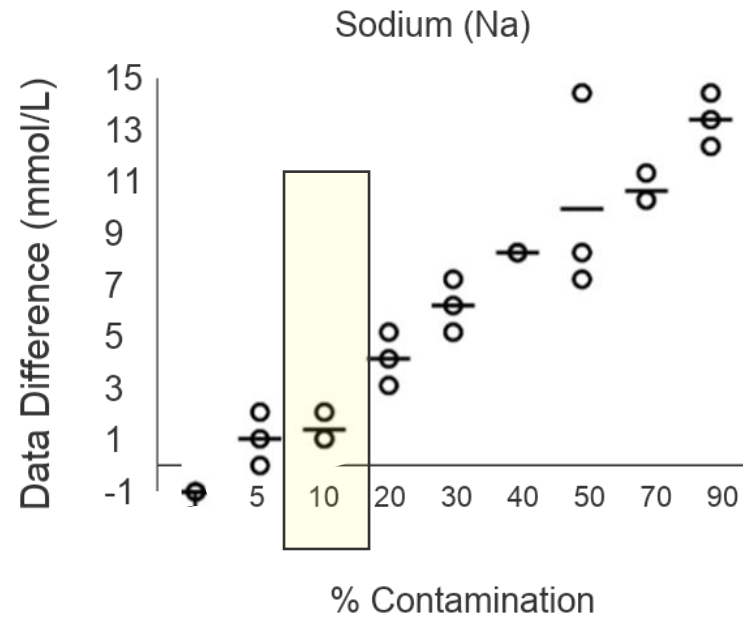
## Patients as Their Own Controls: Use of the Computer to Identify "Laboratory Error"

Jack H. Ladenson

# Multianalyte Delta Checks to assess for IVF Contamination



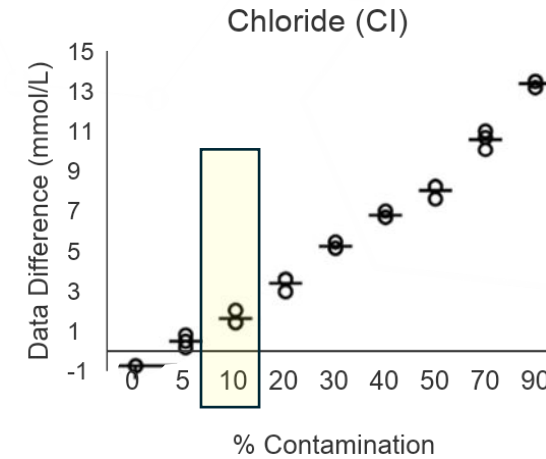
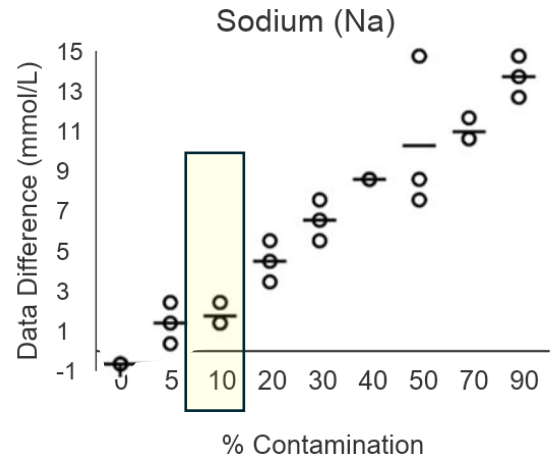
Added normal saline to blood at increasing proportions



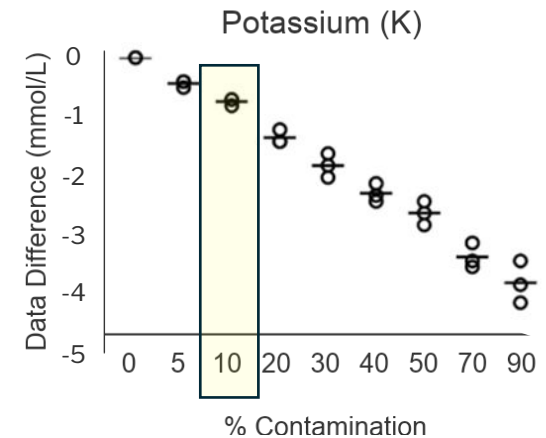
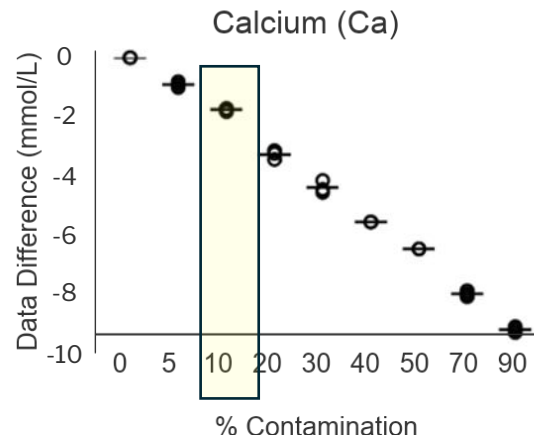
## Fluid Compositions of Normal Saline

	Sodium	Chloride	Potassium	CO2	Creatinine	BUN	Calcium	Glucose
	154	154	0	0	0	0	0	0

# Multianalyte Delta Checks to assess for IVF Contamination



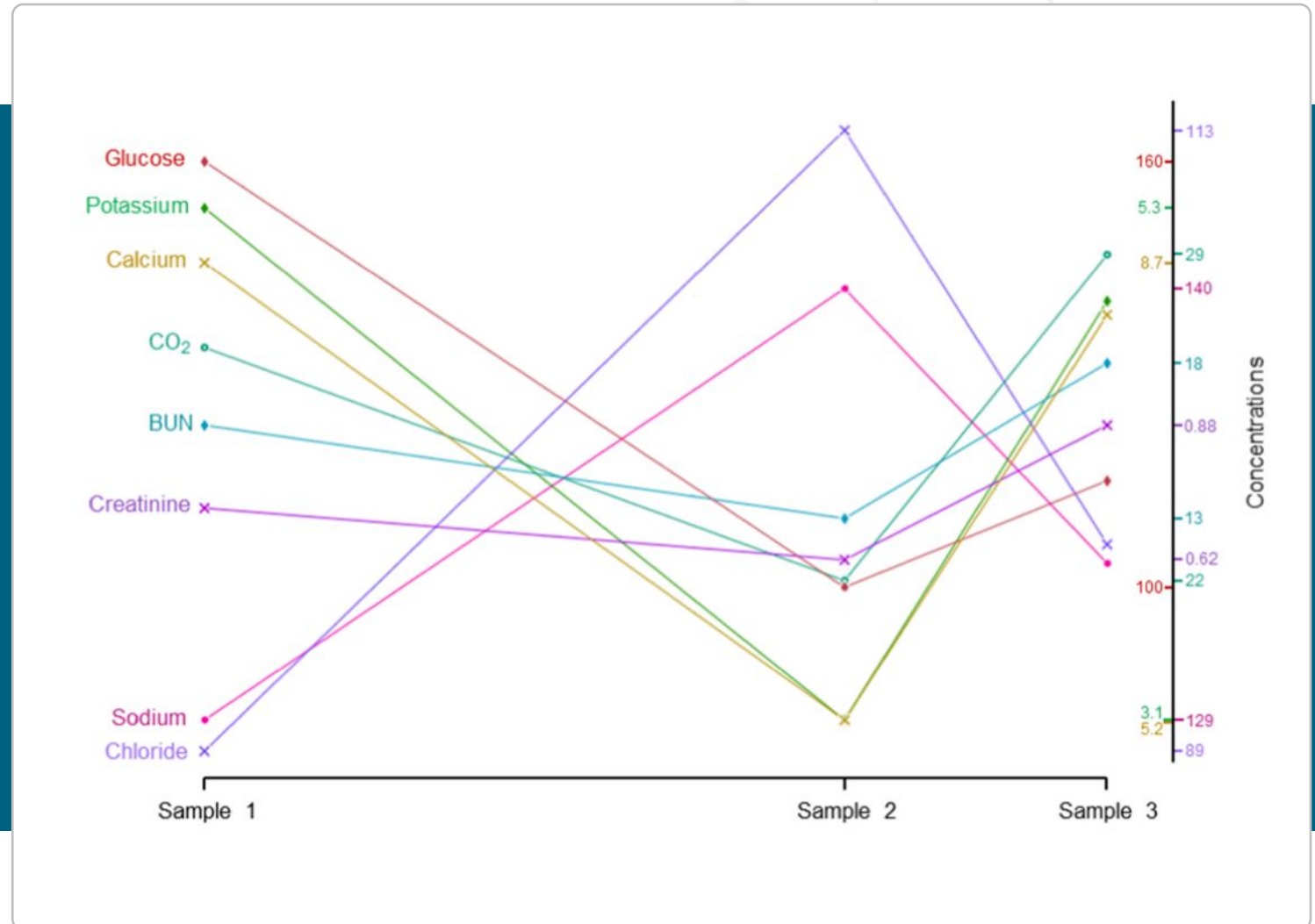
Normal Saline Contamination of 10% =  $\Delta \text{Cl} > 7.7$ ,  $\Delta \text{K} > -0.7$  and  $\Delta \text{Ca} > -1.7$





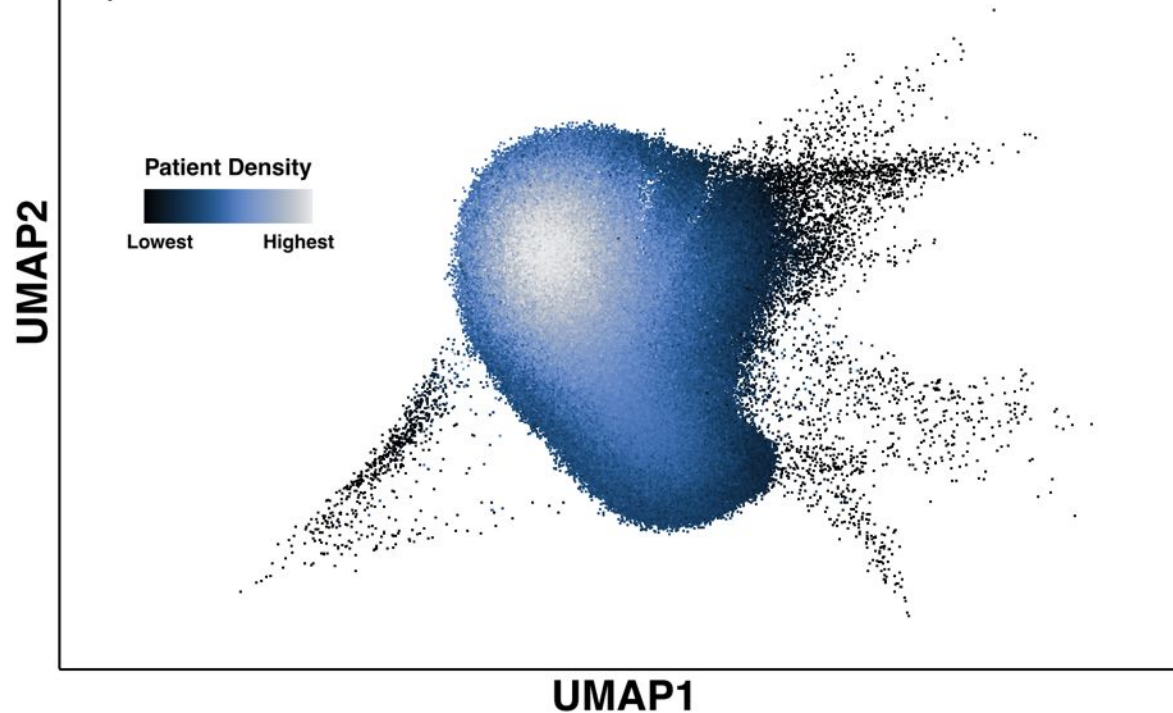
# Anomaly with Resolution (AWR) with IVF Contamination

- Reviewed 10 patients in which the rule would have fired
- All exhibited the AWR pattern
- All receiving NS at time of specimen collection



# Emerging approaches for detecting IV fluid: Machine learning

**UMAP Embedding of Population-Level Variation**  
Inpatient Basic Metabolic Panels



		<b>WashU</b>		MCC	0.747
		Expert Review		Sens	0.858
		-	+	Spec	0.993
Pipeline	-	8509	18	PPV	0.657
	+	57	109	NPV	0.998
Contamination Rate per 1000			<b>14.61</b>	Acc	0.991

# Key Takeaways:

- 01 IVF contamination is likely common in hospitals but hard to detect
- 02 Delta check rules can be implemented and their impact maximized by leveraging studies or your own data
- 03 Ways of interfacing machine learning algorithms into the LIS are needed

# Conclusions

- 01 Preanalytical error is common (~0.8% of all samples in a core lab) and is underrecognized.
  - Better tools are needed to identify preanalytical error
- 02 The LIS can be leveraged to identify common preanalytical error
  - However labs needs better access to the LIS to apply these rules
- 03 Working with other departments including the ED and nursing can help reduce preanalytical error



# Thank you!



Nick Spies, MD  
Yanchun Lin, PhD  
Hannah Brown, PhD  
Mark Zaydman, MD,  
PhD  
Ann Gronowski, PhD  
Abe Qavi, MD, PhD

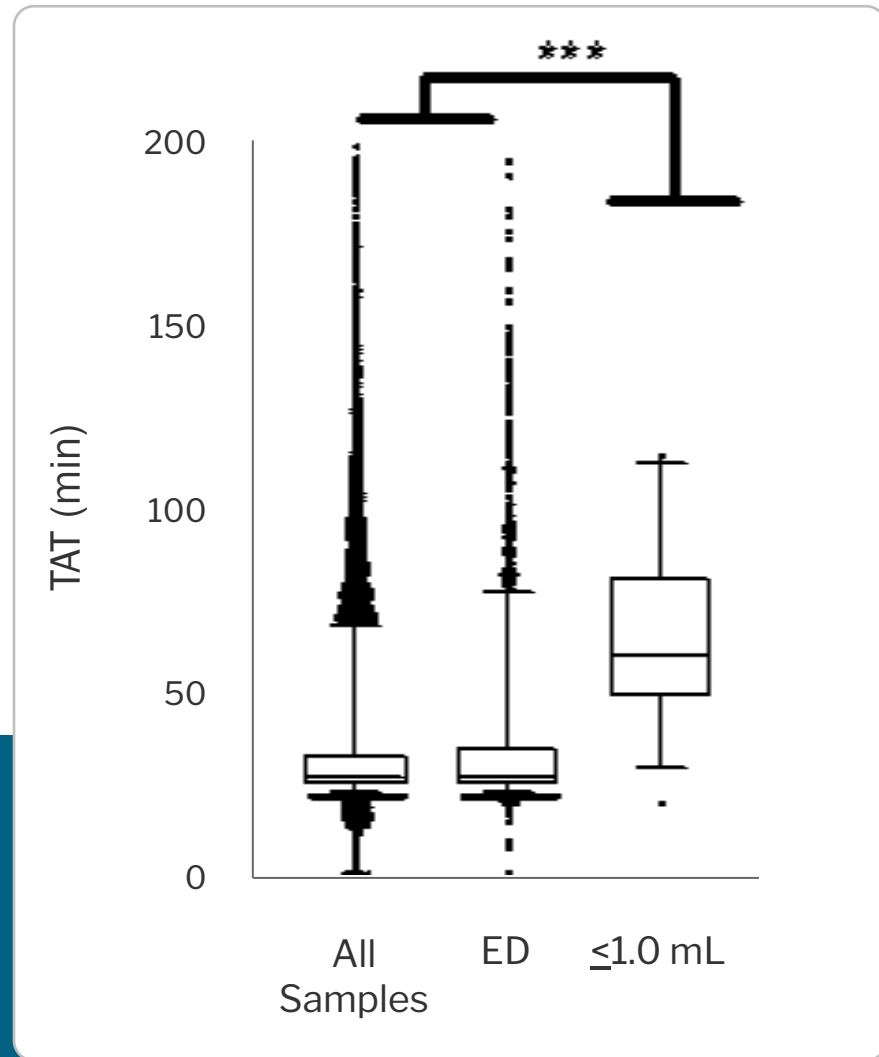
cwfarnsworth@wustl.  
edu



# Testing methods in modern laboratories have changed



# QNS samples = longer turnaround time (TAT)



Normal sample TAT:

30 mins

Short sample TAT:

62 mins

Canceled Sample TAT:

127 mins

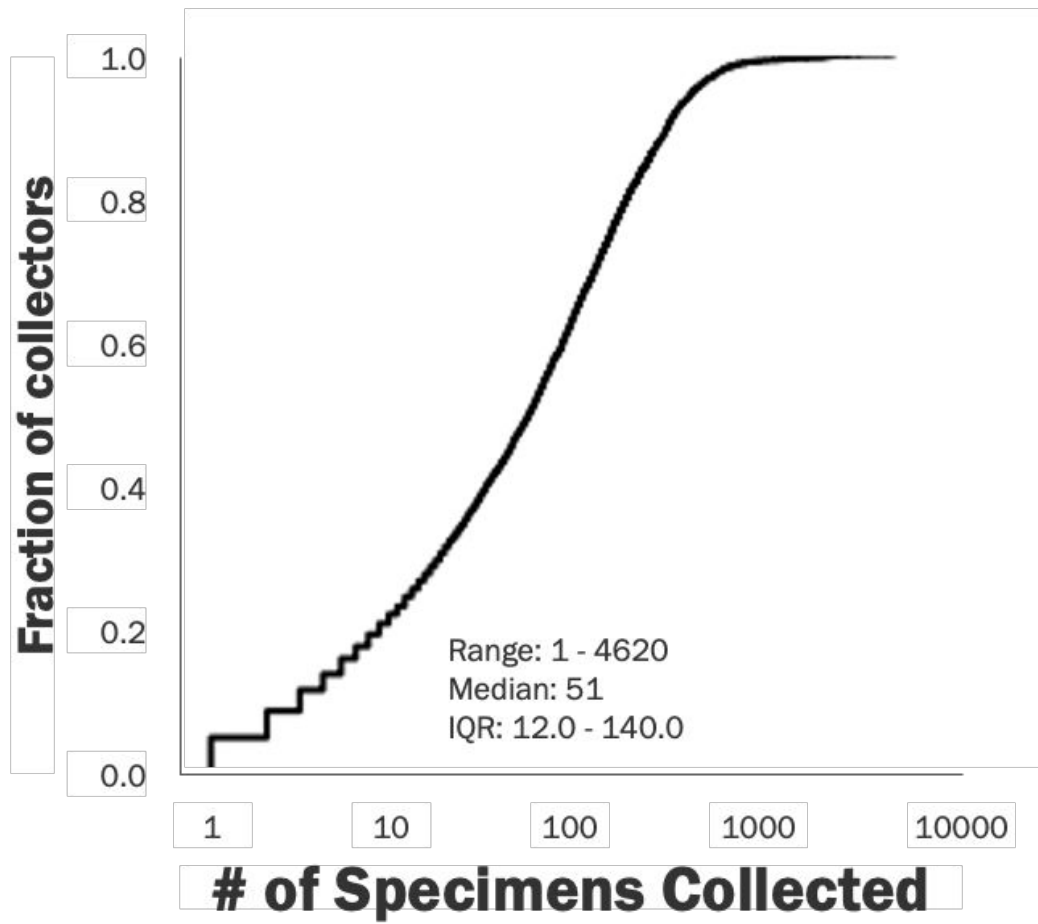


Big problem  
in ED's

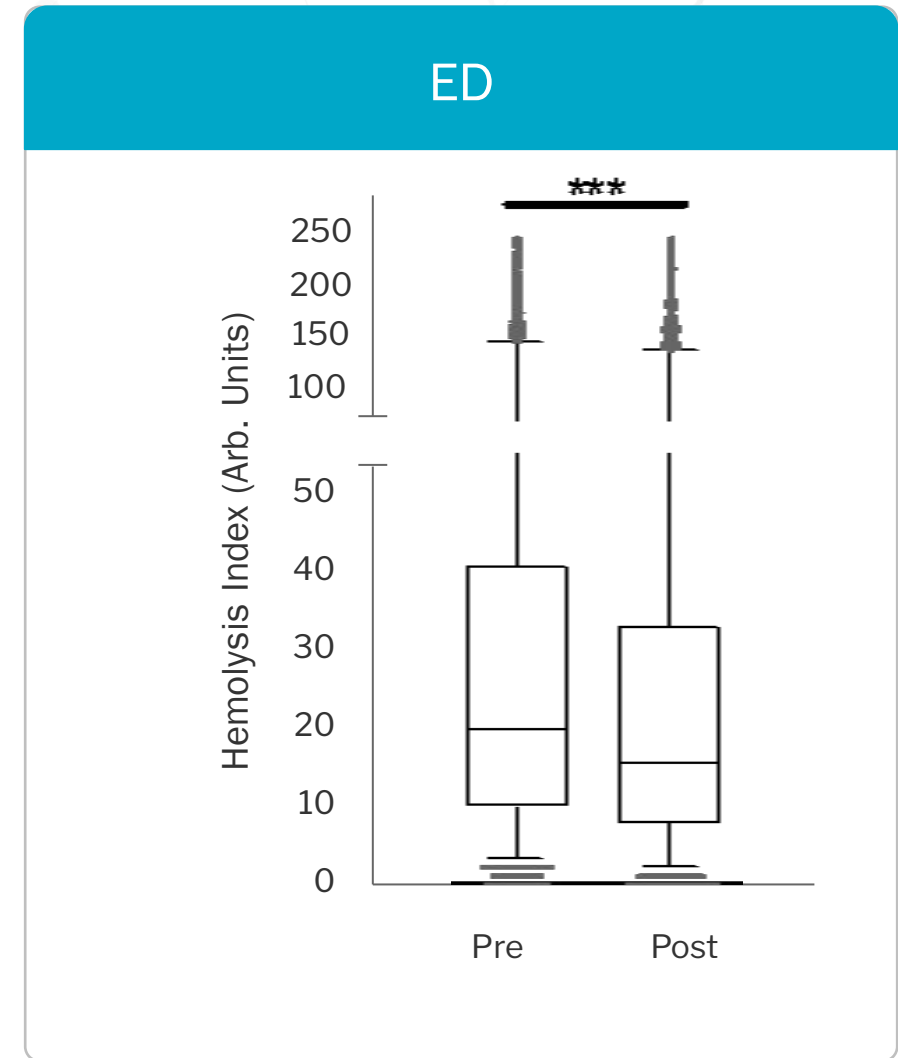
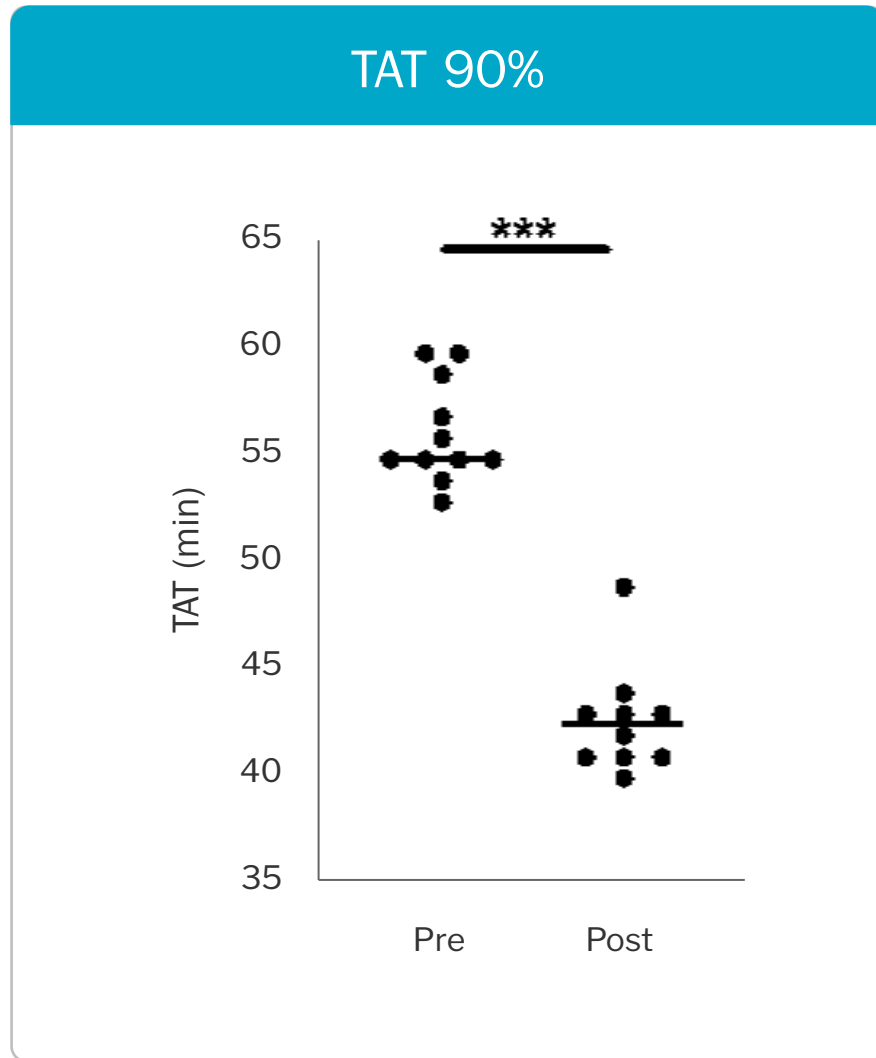
- Prolonged length of stay
- Low throughput
- Long boarding times



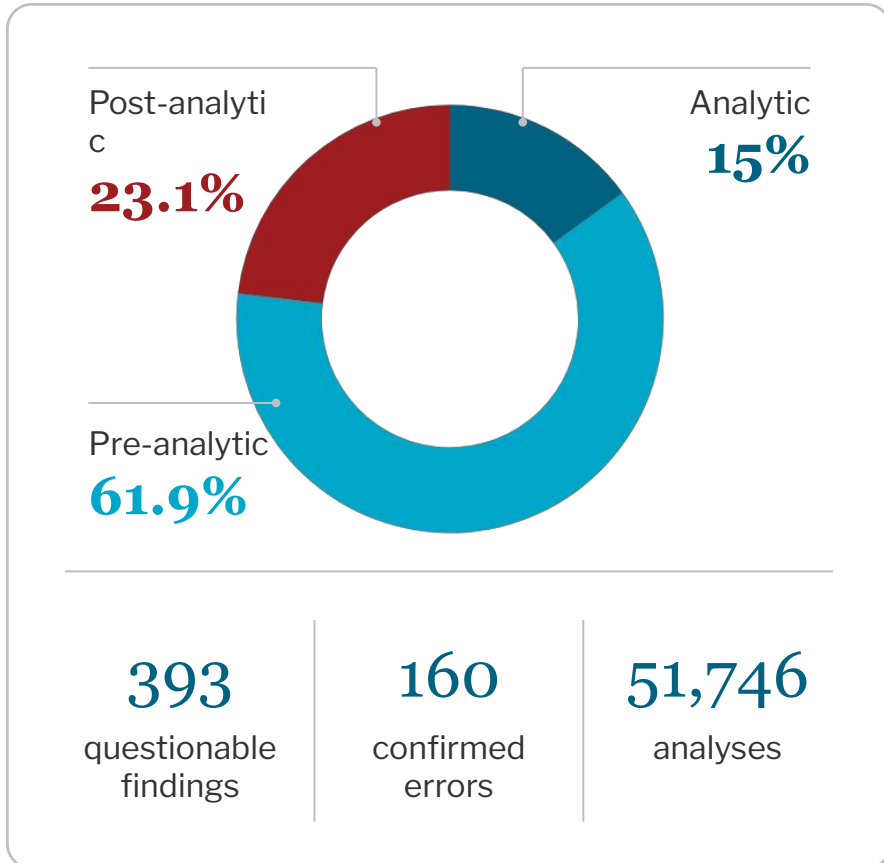
# Substantial variability in # of collections and hemolysis frequency



# QNS Policy improves in-lab TAT and reduces hemolysis



# Errors can occur at any point in the lab testing process



Physicians and nurses told to pay attention to test results



Suspected laboratory error was recorded



Daily, lab physician visited and appraised for errors

# Assessing causes of error in the BJH/ Wash U Laboratory

## Preanalytical Category

- > Collection errors
- > IV contamination
- > Specimens too old
- > Improperly Labeled

## Analytical

- > QC out of Range (assay drift)
- > Instrument problems
- > Reagent Issues  
(bad reagent pack)

## Post-analytical Category

- Comment errors
- Result entry errors
- Dilution errors

## Data Sources

Errors captured by querying LIS or by LIS flagging.

Daily report manually curated by a trained medical laboratory scientist.

# Lab error occurs frequently and are mostly preanalytic

87,317 Errors

60,748 Were hemolysis errors!

Category	N=	Freq. (%)
Hemolyzed reported	41,047	48.2
Hemolyzed masked	19,701	23.1
Quantity not sufficient	8,068	9.5
Clotted samples	5,840	6.9
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IV contamination	1,122	1.3
Too old to test	550	0.6
Sample integrity	92	0.1
Requisition errors	62	0.1
Total	85,133	100

# Lab error occurs frequently and are mostly preanalytic

87,317 Errors

60,748 Were hemolysis errors!

= 94% Of all error!  
Without hemolysis, preanalytical error

24,385 of 25,808 errors

Category	N=	Freq. (%)
Hemolyzed reported	41,047	48.2
Hemolyzed masked	19,701	23.1
Quantity not sufficient	8,068	9.5
Clotted samples	5,840	6.9
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Too old to test	550	0.6
Other- specimen integrity	92	0.1
Requisition errors	62	0.1
Total	85,133	100

# CAP requires feedback to collectors for quality

GEN. 40499 Specimen Collection Feedback

Phase I

There is a mechanism to provide feedback to the collectors of specimens on issues relating to specimen quality and labeling.

Note: The accuracy of an analytic result depends upon the initial quality of the specimen. Proper collection techniques are essential.

---

Evidence of Compliance:

- ✓ Written procedure defining methods for providing feedback to specimen collectors AND
- ✓ Records of communication of specimen collections issues, such as QM reports, staff meeting minutes OR records of employee counseling

Problem: Who do you provide feedback to??

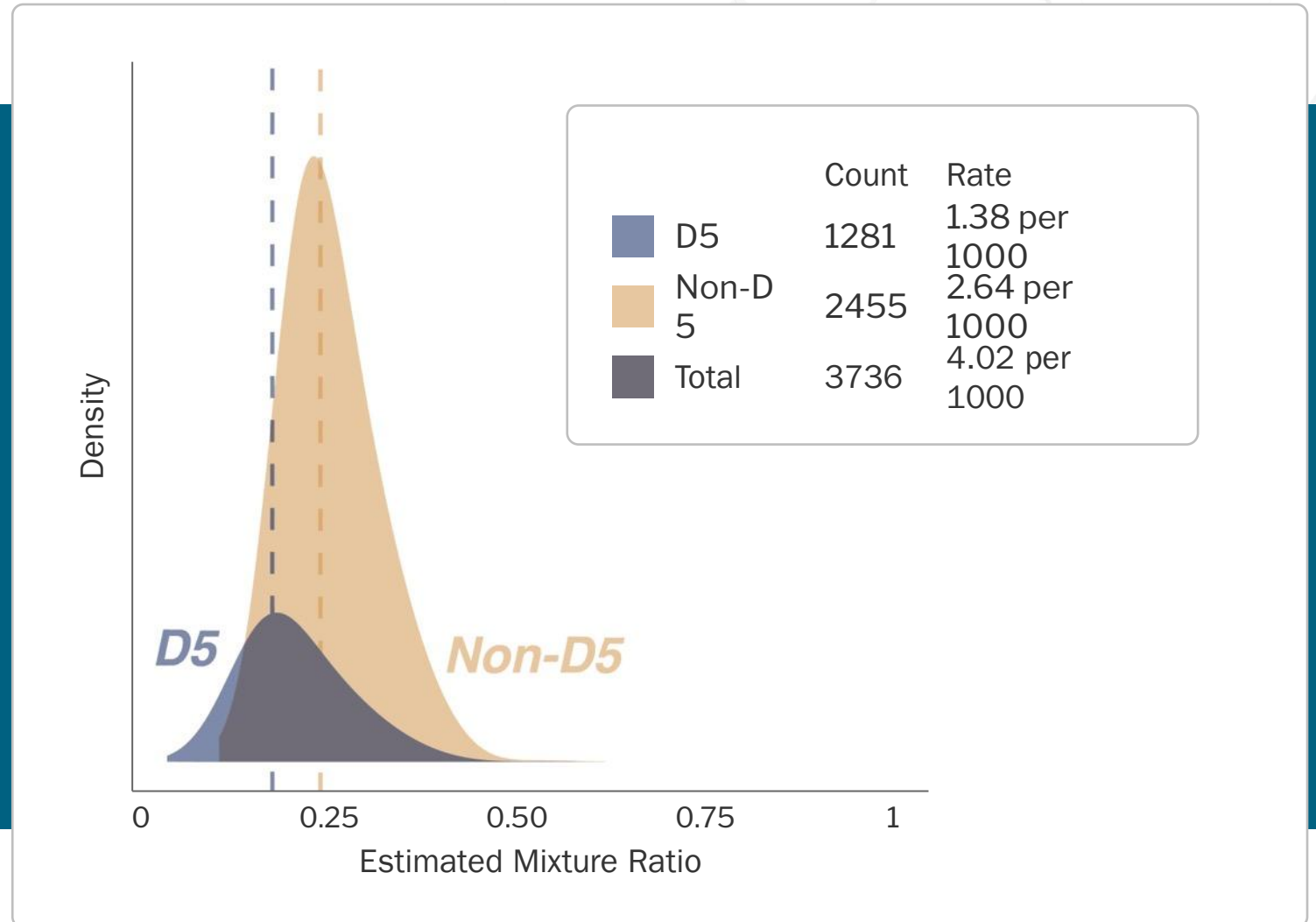


# Identifying underperformers using your laboratory data

- Collectors across all units are captured using PPID
- Collector associated with the hemolysis index for each specimen
- Assess the frequency of hemolyzed samples

# Humans are bad at detecting IVF contamination

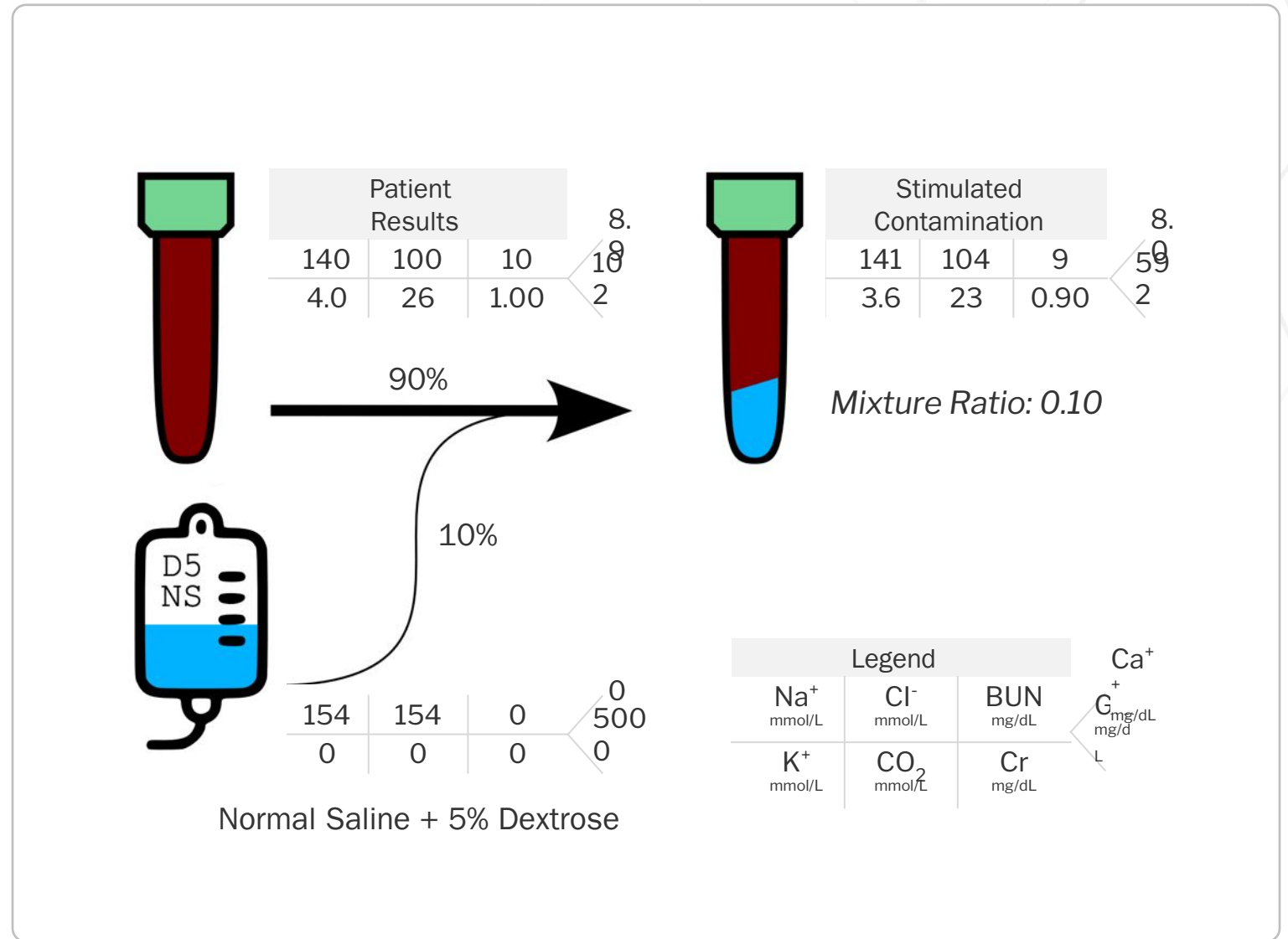
- BMP results as contaminated (specimen redrawn within 4 h)
- 18% (IQR 9-27%) for dextrose containing fluids
- 24% (IQR 16-38%) for non-dextrose fluids



# How much contamination is too much?

928,742  
BMP results

- ▶ Simulated mixing study
- ▶ Assessed # results exceeded
  - Total Allowable Error (TEa)



# Total allowable error (TEa) exceeded at ~10% normal saline (NS)

Minimum Significant Mixtures  
 50% of results contaminated at this ratio will exceed TEa thresholds.

	Sodium	Chloride	BUN	Calcium	Potassium	CO2	Creatinine	Glucose
Normal Saline (NS)	30%	7%	12%	12%	14%	10%	29%	6%
CLIA TEa	4mmol/L	3mmol/L	2mg/dL	1mg/dL	0.5mmol/L	2mmol/L	0.3mg/dL	6 mg/dL

Spies NC & Farnsworth CW, Jour of Laboratory Medicine. 2024;48:29-36.  
 Spies NC et al. Clinical Chemistry 2024;70:444-52.  
 Spies NC et al. Clinical Chemistry 2024;70:444-52.

# Harnessing your own data to establish single analyte delta checks

Most predictive single-analyte delta check and cutoff value

Decreased calcium;  
 $\Delta\text{Ca}\% \leq -24\%$

Sensitivity (95% CI) at the cutoff

76.4% (70.7%–82.0%)

Specificity (95% CI) at the cutoff

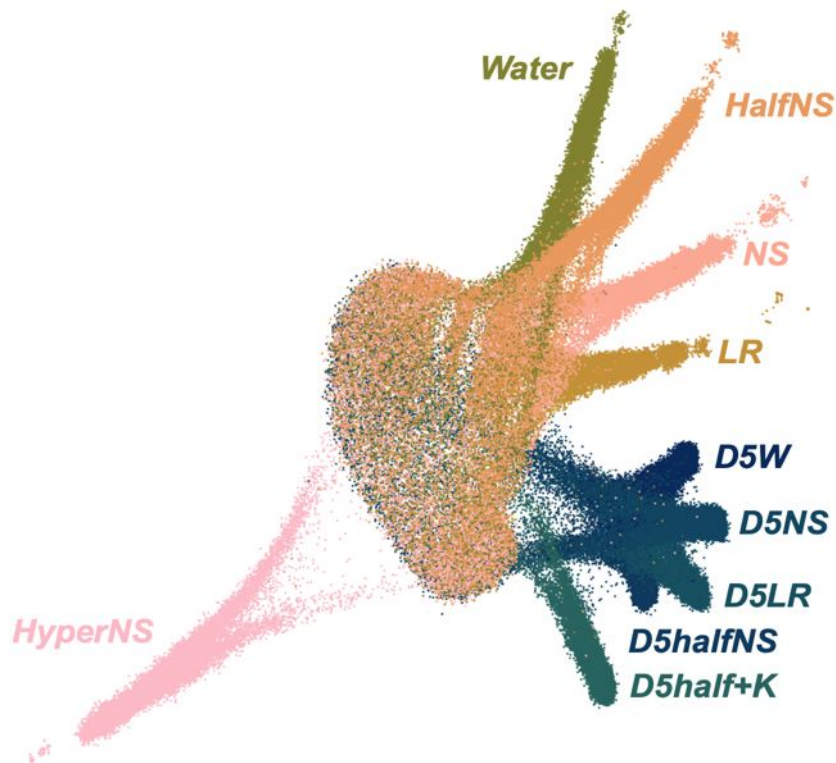
99.2% (98.7%–99.8%)

- 326,103 BMP/CMPs
- Used changes in the RCV to identify potential contaminants
- Chart review performed on 1,489
- Use logistic regression to identify ideal delta thresholds

Data sets	Sample size	Parameters	Logistic regression models	Single-analyte delta checks
Labeled training data set	1489	Sensitivity	77.2% (95% CI: 73.5%–80.9%)	70.3% (95% CI: 66.3%–74.4%)
		Specificity	98.7% (95% CI: 98.1%–99.4%)	97.2% (95% CI: 96.2%–98.2%)
		PPV	91.8% (95% CI: 89.2%–94.4%)	82.3% (95% CI: 78.7%–86.0%)

# Simulated results are outliers from the main embedding

A. Simulated Contamination  
Fluid Type

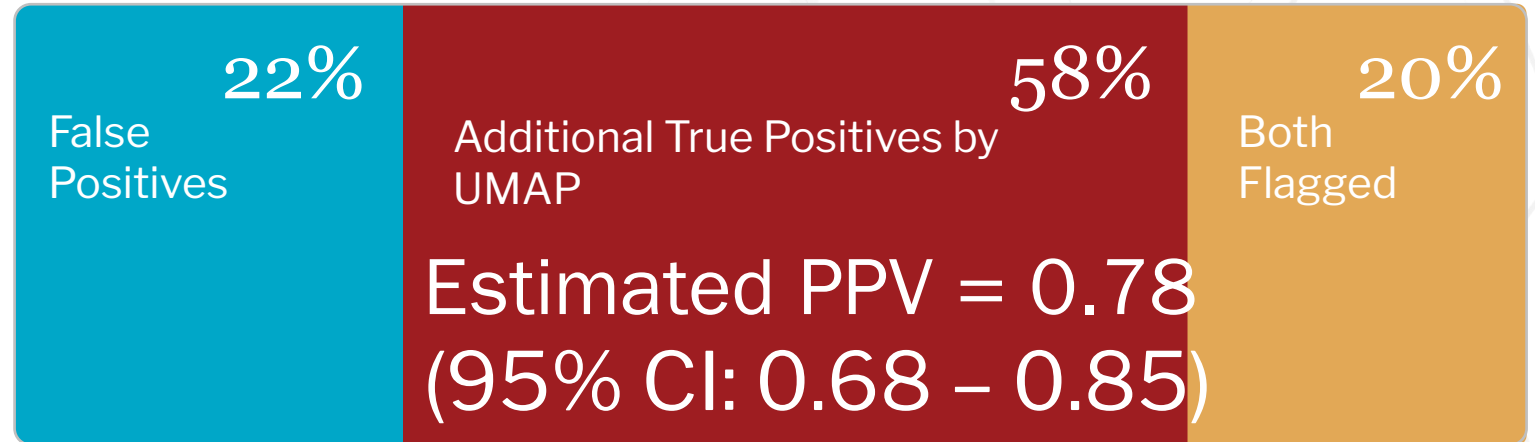


B. Simulated Contamination  
Mixture Ratio



# Unsupervised machine learning to detect IVF contamination

100 Consecutive UMAP Flags



## Future Directions:

01

Can we create better models

02

Does it generalize to other hospitals

03

How do you implement?