# **Real-time Detection of Infusion Set Failures in a Closed-Loop Artificial Pancreas**

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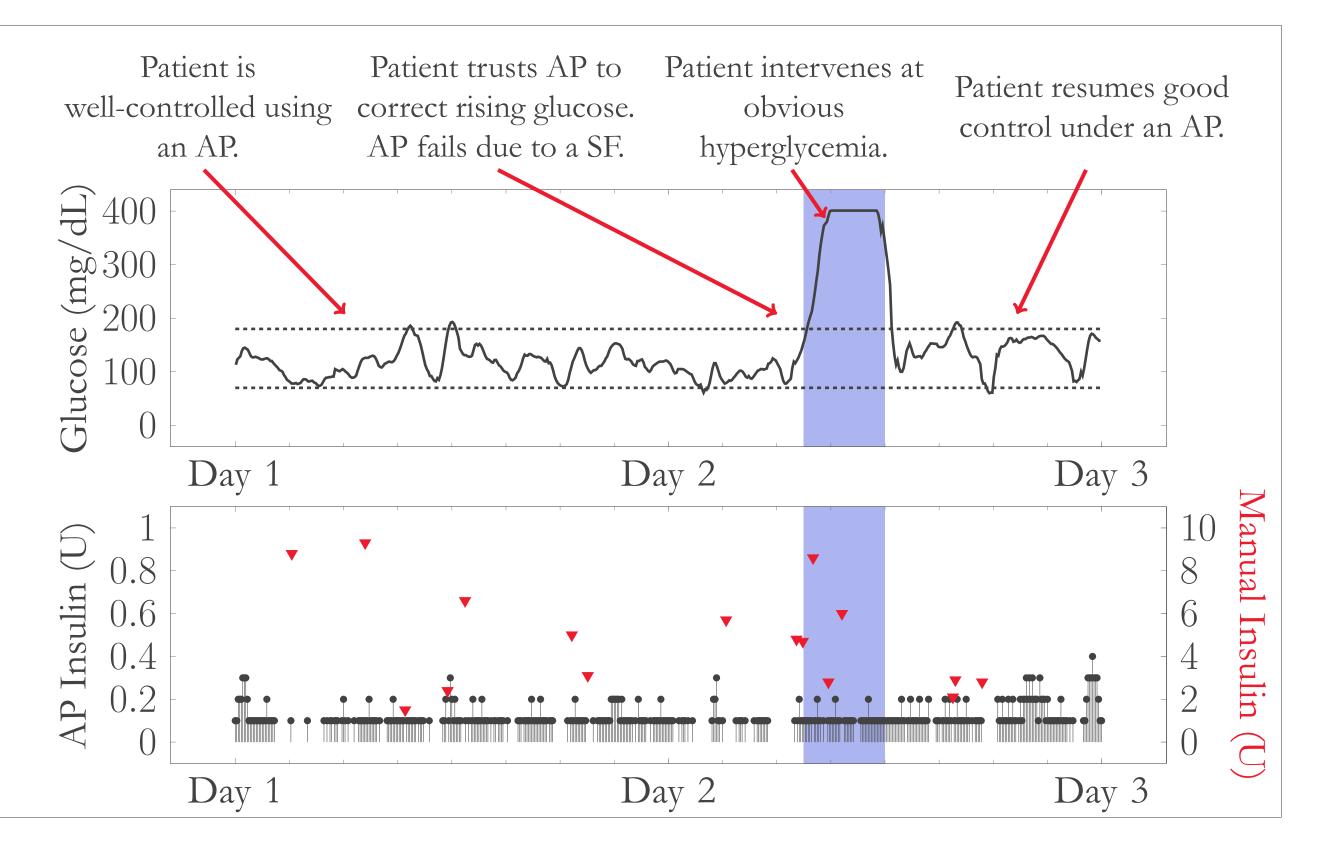
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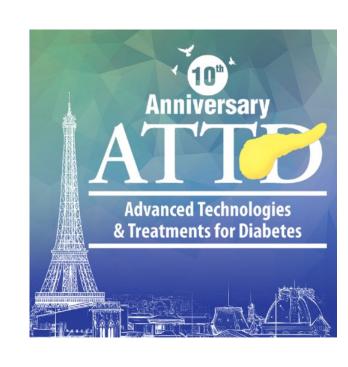
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## Background

As evidence emerges that artificial pancreas systems may improve clinical outcomes for patients with type 1 diabetes and decrease the burden of disease management, there is the risk that

### decrease in burden could cause patients to be less observant of when devices stop functioning appropriately.





See the following related poster for additional study information: ATTD7-0151: Zone-MPC Performance

Right is an example of an infusion set failure (SF) that occurred under closed-loop as well as the patient's thought process in detecting and addressing this fault.

## Methods

### **SF** Detection Algorithm

The SF detection algorithm was developed previously [1]. Briefly, patient-specific baselines of glucose control and insulin requirements are adapted over time via averages on sliding windows. These baselines are compared to current values and alarms occur when thresholds are exceeded.

**Zone Model Predictive Control (Zone-MPC)** 

Zone-MPC [2] is an AP algorithm that controls to a desired zone, rather than to a single setpoint.

### **Study Design**

RCT with cross-over:

SAP	VS.	Zone-MPC
		+ SF det.

Algorithm + Patient Patient Detection Detection

SFs were determined by failed correction doses ketones > 0.6(11)

### **Primary Clinical Outcome**

The SF detection algorithm attempts to reduce SF-associated hyperglycemia. Therefore, the primary clinical outcome sought to reduce

the time > 250 mg/dL in the 4 hours prior to a SF detection

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Patient	Demograph	ics

Subjects:	19
Sex:	11 F, 8 M
Age (years):	$28.0 \pm 10.8$
Years with T1D:	$12.7 \pm 7.1$
Insulin (U/day):	56.3 $\pm$ 18.4
Weight (kg):	86.1 $\pm$ 22.8
Insulin/BW	$0.65 \pm 0.22$
(U/day/kg):	

Reported as mean  $\pm$  SD

(iii)insulin leaking at insertion pump-occlusion alarms (iv)

between Zone-MPC + SF detection and SAP arms.

### Results

#### **Primary Clinical Outcome** 240Time > 250 mg/dL (min)180 120 ••• 60 •• SAP Zone-MPC + SF detection

Zone-MPC SAP + SF det. Failure to Correct 16 11 Insulin Leaking at Site 3 Infection Pump-Occlusion Unknown ()16 19 Total

**Classification of Failures** 

#### **Algorithm Performance**

The SF detection algorithm attained 88% 0.22 Sensitivity FP/day (72% sensitivity and 0.28 FP/day predicted from previous analysis.)

Mean time > 250 mg/dL: 101 vs. 43 minutes (p < 0.01) Total number of SFs not statistically significant between intervention groups (Chi-square test of independence, p = 0.49)

This is the first study in the literature that investigates online detection of SFs.

## Discussion

## Conclusion

False positives: Fewer FPs than indicated by previous retrospective analysis. Likely reducible further by announcing meals.

False negatives: 3 SFs remained undetected. One overzealous patient had already experienced three SFs in the seven day period, making her quicker to act than the algorithm. The other two developed very slowly and simulations indicate that they can be easily detected by increasing the length of the long window.

Future Work: The main contribution of this work is in evaluating the derived features for the detection of SFs. Other classification techniques (e.g., discriminant analysis) that incorporate these features could further improve performance.

As patient burden is reduced by each generation of advanced diabetes technology, fault detection algorithms will help ensure that patients are alerted when they need to manually intervene.

### References

- Howsmon DP et al. Continuous glucose monitoring enables detection of Losses in Infusion Set Actuation (LISAs). Sensors. 2017;17(1):161.
- Gondhalekar R et al. Periodic zone-MPC with asymmetric costs for outpatient-ready safety of an artificial pancreas to treat type 1 diabetes. Automatica. 2016;71:237-46.





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