# GluGo2.0

Assessing the Merits of the current GluGo Algorithm

### New T1D Paradigm on the Horizon



Dexcom G6 Sensor



### GluGo's Algorithm

- Structural model (no machine learning involved)
- Jingxian Liu (17') and Jihong Jin (16')

$$UpwardSlope(/min) = \frac{\frac{Sens}{CarbRatio} * Carb}{4*60}$$

$$DownwardSlope(/min) = -\frac{Insulin*Sens - Basal*4*Sens}{4*60}$$

# Initial Hypotheses, Questions, and Concerns

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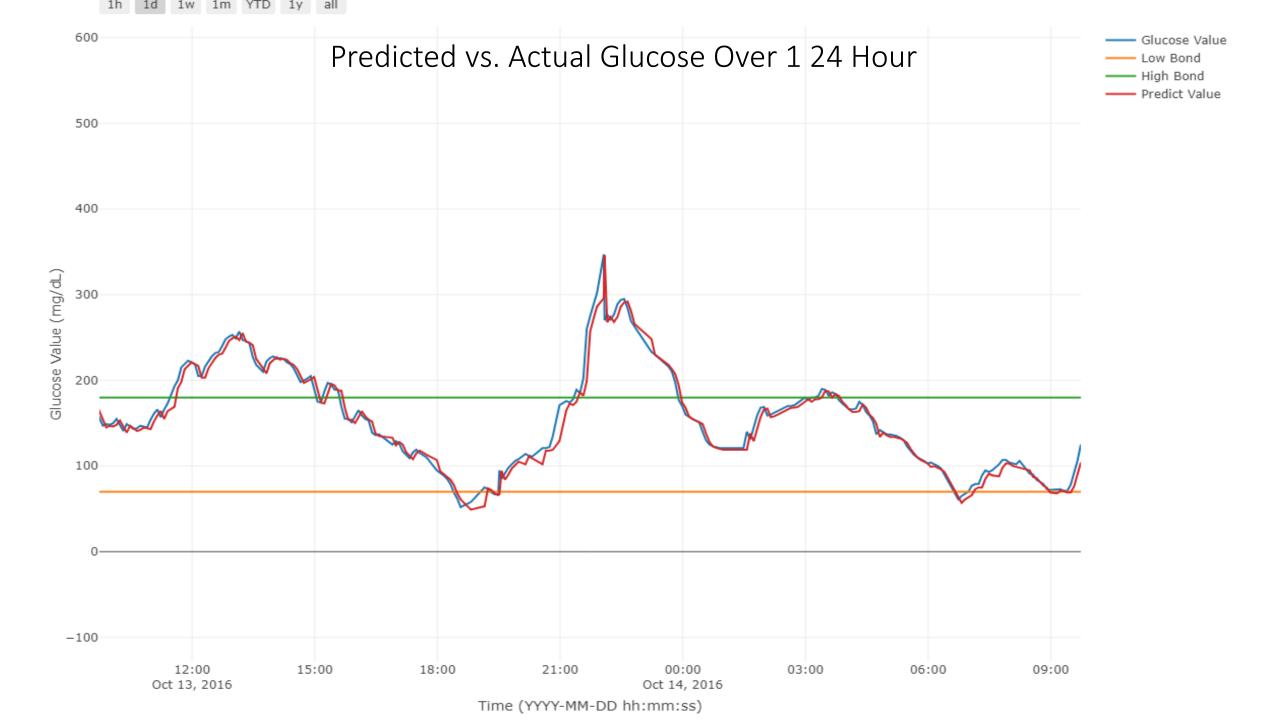
### Questions

- Are the predictions biased?
- Why is it less accurate for high glucose

# Initial Hypotheses, Questions, and Concerns

#### Concerns

- The algorithm is based solely on theory, and makes *many* simplifying assumptions
- Doesn't "learn" about the patient (no ML) not personalized



### Summary Statistics on Residuals and Absolute Residuals

### **Residuals:**

Mean	Median	Standard Deviation	Minimum	Maximum
-2.92	-2	9.11	-285	296

Sample Size
n = 28,078

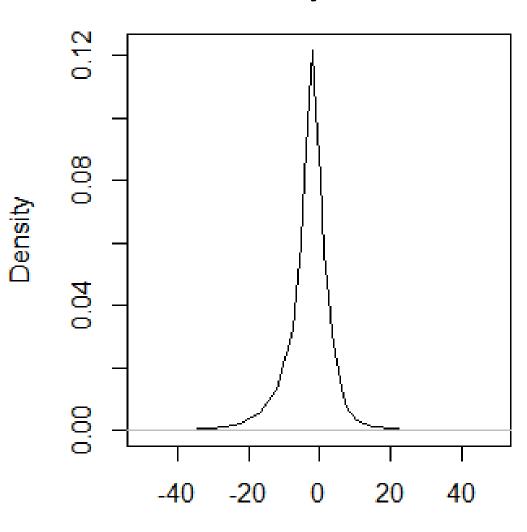
### **Absolute Residuals:**

Mean	Median	Standard Deviation	Minimum	Maximum	Average % Residual
5.16	3	8.05	0	296	4.01%

# Residuals (Predicted – Actual) by Actual Glucose Value

### 300 0 200 100 Residuals -100 0 -300100 200 300 400 Actual Glucose Values

### **Residual Density Plot**



Residual Value (in range -50 to 50)

### Upon Closer Inspection...

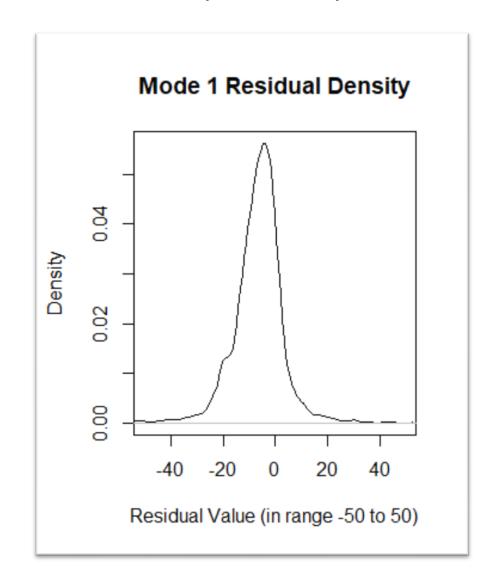
- The algorithm predicts at most 15 minutes into the future
- Usually only *5 minutes*

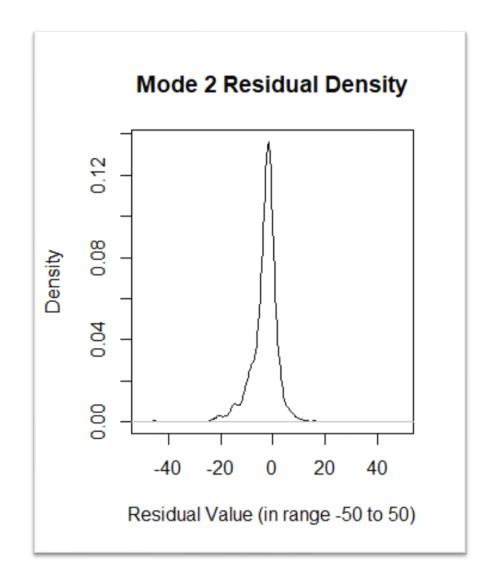
- The test data used by the previous group is not a random sample
- Only a fraction of the total sample includes carb intake
- Algorithm functions in two "modes"
- Mode 1: After Carbohydrate Intake
- *Mode 2*: No Carbohydrate Intake

# Mode 1 (Carbs) vs. Mode 2 (No Carbs)

	Mode 1	Mode 2		
Mean Residual	-5.53	-3.25		
Mean Abs. Residual	9.74	4.23		
Average % Difference	7.37%	3.73%		
Sample Size	n = 3,016	n = 1,714		

### Mode 1 (Carbs) vs. Mode 2 (No Carbs)





# Conclusions about Algorithm

1. Old algorithm in it's current form is not optimal

1. Prior Assessments of the model's accuracy are inflated due to a biased sampling procedure

# Looking Forward: Building New Test Datasets

### Compiling a new dataset

Α	В	С	D	E	F	G
Timestamp	Glucose (ml/dL)	Basal Insulin (U/hr)	Insulin Sensitivity (mg/dL/U)	Carb Ratio (g/U)	Bolus Insulin (U)	Carb Intake (g)

# Looking Forward: Building New Test Datasets

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Α	В	С	D	E	F	G
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Testing the new dataset

```
from dataAccess import dataAccess as DB
import datetime as DT
import csv
dbpath = "lab.db"
katDB = DB("Kate",dbpath)
   dat = patient.getData()
gluPredict = []
while i < len(dat):
   if dat[i][9] == "INRANGE":
      if dat[i][3] != None:
          bolis = dat[i][3]
          carb = dat[i][4]
          carbRate = dat[i][15]
          basalRate = dat[i][16]
          sens = dat[i][17]
          qluPredict.append(dat[i-1][2])
          currentTime = dat[i][1]
          upwardSlope = ((sens/carbRate) * carb) / (4*60.0)
          downwardSlope = (- bolis * sens - basalRate * 4 * sens) / (4*60.0)
              upValue3 = round(upValue2 + ((dat[i+2][1] - dat[i+1][1]).total\_seconds() / 60.0) * upwardSlope) 
              gluPredict.append(upValue1)
              gluPredict.append(upValue2)
              gluPredict.append(upValue3)
              while dat[i][3] == None and i < len(dat) -4 and dat[i+1][9] == "INRANGE" and (dat[i][1] - currentTime).total_seconds() < 15300
                 newValue = round(dat[i-1][2] + ((dat[i][1] - dat[i-1][1]).total_seconds() / 60.0) * downwardSlope)
                 qluPredict.append(newValue)
```

# Moving from a database to directly reading a csv and using pandas library

```
import pandas
data = pandas.read_csv('mlk.csv')
length_csv = len(data)
gluPredict = []
while i < length_csv:
    if data["Bolus Insulin (U)"][i] != None: # assuming that if Bolus insulin has a value then, it will be a
        bolis = data["Bolus Insulin (U)"][i]
         carb = data["Carb Intake (g)"][i]
         carbRate = data["Carb Ratio (g/U)"][i]
         basalRate = data["Basal Insulin (mg/dL)"][i]
         sens = data["Insulin Sensitivity (mg/dL/U)"][i]
         gluPredict.append(data["Glucose (ml/dL)"][i])
        print("glucose level: ")
print(data["Glucose (ml/dL)"][i]) #
        currentTime = data["Timestamp"][i] #
         upwardSlope = ((sens/carbRate) * carb) / (4*60.0)
         downwardSlope = (-bolis * sens - basalRate * 4 * sens) / (4*60.0)
        if data["Bolus Insulin (U)"][i].isnull():
                 upValue1 = round(data["Glucose (ml/dL)"][i-1] + ((data["Timestamp"][i] - data["Timestamp"][i-1]).total_seconds() / 60.0) * upwardSlope)
upValue2 = round(upValue1 + ((data["Timestamp"][i+1] - data["Timestamp"][i]).total_seconds() / 60.0) * upwardSlope)
upValue3 = round(upValue2 + ((data["Timestamp"][i+2] - data["Timestamp"][i+1]).total_seconds() / 60.0) * upwardSlope)
                  gluPredict.append(upValue1)
                   gluPredict.append(upValue2)
                   gluPredict.append(upValue3)
                   while data["Bolus Insulin (U)"][i] == None and i < length_csv -4 and (data["Timestamp"][i] - currentTime).total_seconds() < 15300 :
                      newValue = round (data["Glucose (ml/dL)"][i-1] + ((data["Timestamp"][i] - (data["Timestamp"][i-1]).total_seconds() / 60.0) * downwardSlope
                       print("i am here")
                       gluPredict.append(newValue)
         newWalue = round(data["Glucose (ml/dL)"][i-1] - ((data["Timestamp"][i] - data["Timestamp"][i-1]).total_seconds() / 60.0) * (data["Basal Insulin"][i] *
```

### Looking Forward: Better Algorithms

SVR Model with Physiological Features:

