



Problem definition

Uncaught errors in models cost real \$\$

Q:

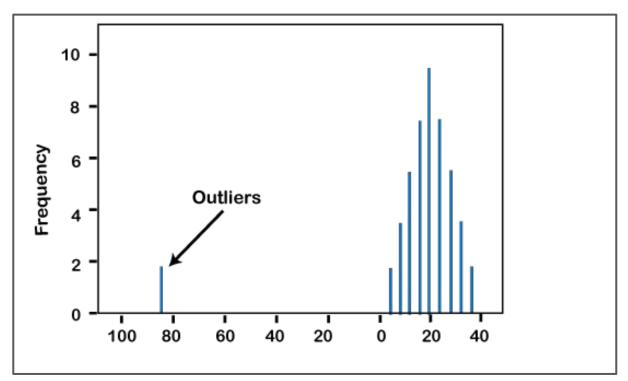
How do you identify when some feature is likely to be an error?

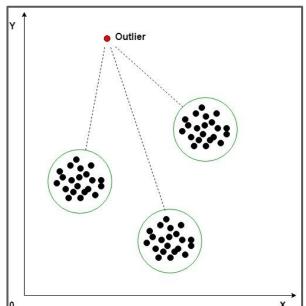
(without hard-coding a huge set of rules)

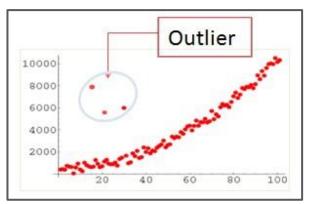
What is an outlier?

In statistics, an **outlier** is a data point that differs significantly from other observations.^{[1][2]} An outlier may be due to variability in the measurement or it may indicate experimental error; the latter are sometimes excluded from the data set.^[3] An outlier can cause serious problems in statistical analyses.

What is an outlier?

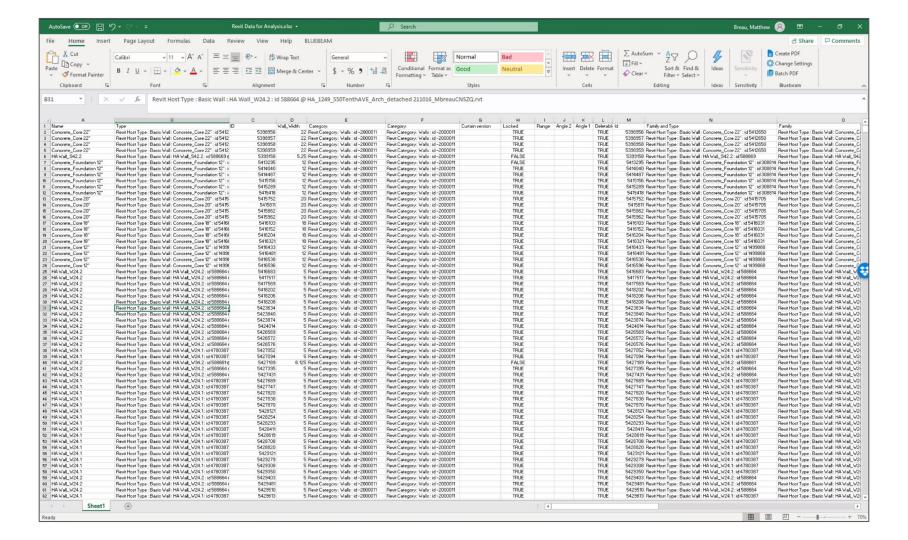




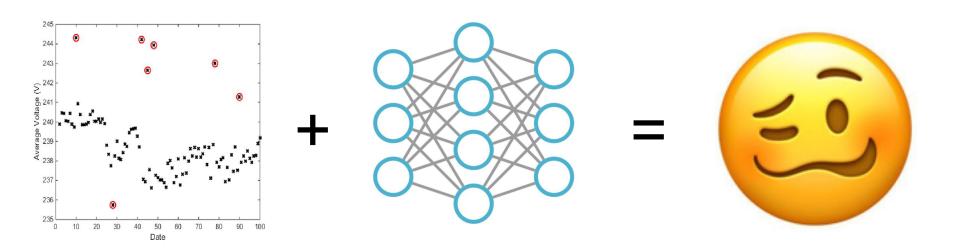


Outliers can be hard to spot

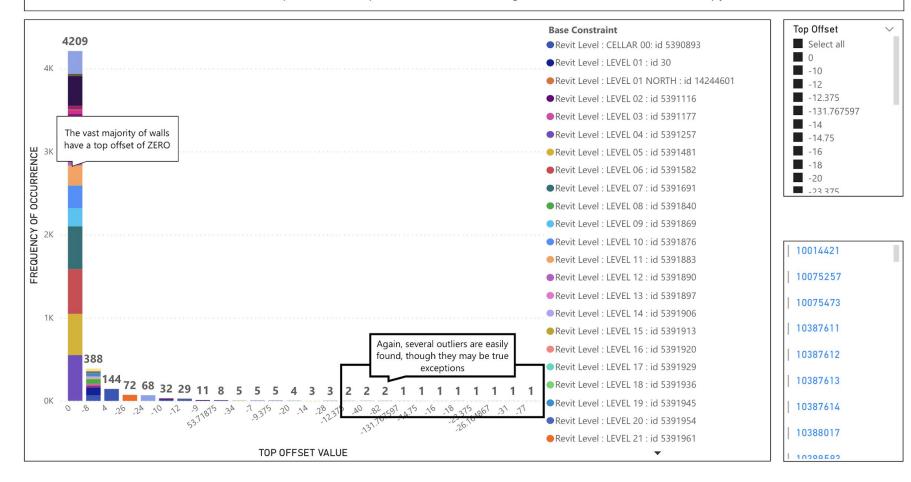




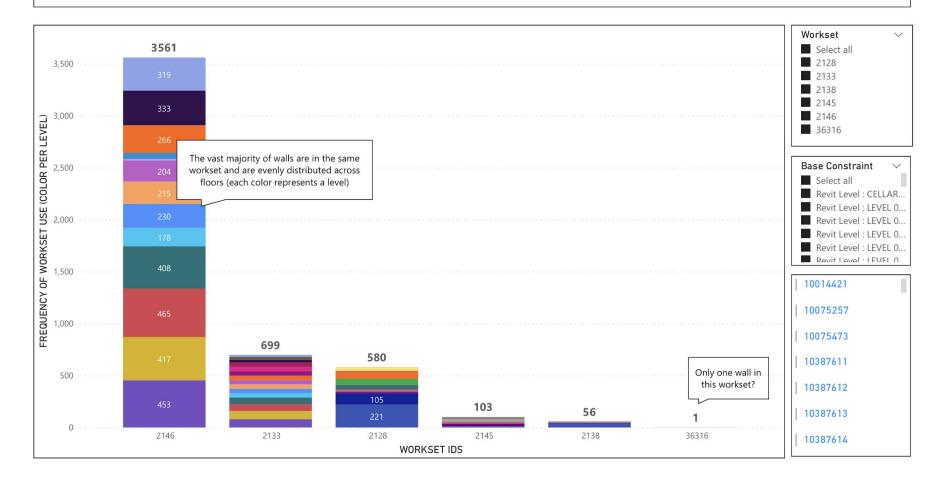
Outliers are bad for Machine Learning



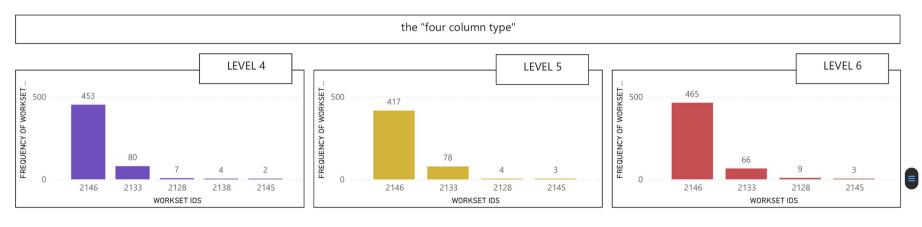
Outliers: TOP OFFSET (each column represents a top offset value, the height of the column shows the qty of walls with that value)

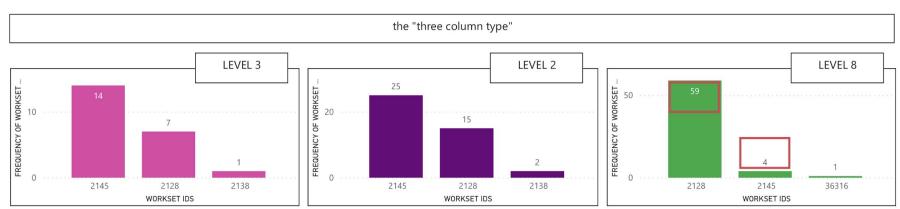


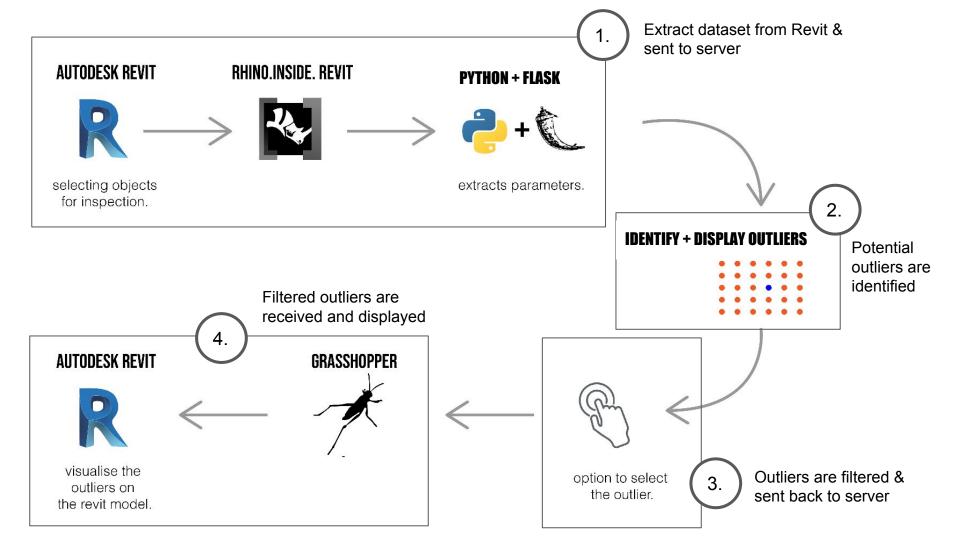
Outliers: WORKSET USAGE (each column represents a workset, height shows number of walls in that workset)



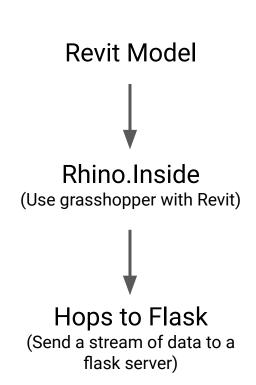
Outliers: NUMBER OF WORKSETS IN USE BY LEVEL (each column represents a workset)

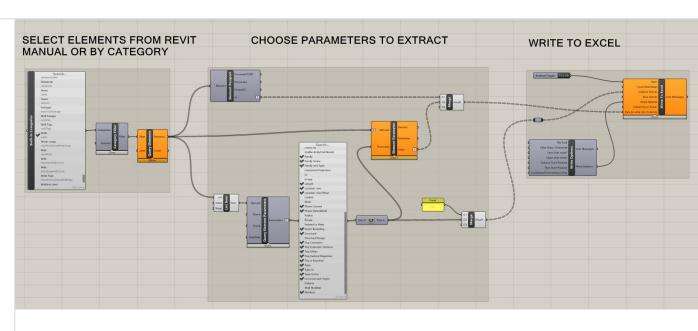






1. Extract data from Revit





2. Identify possible outliers

Flask Server (Call server to access JSON)



SK Learn

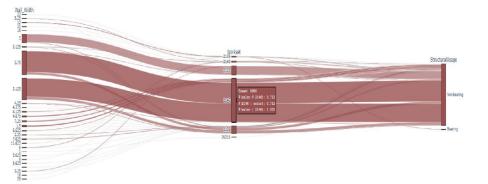
(Univariate outliers, Mahalanobis, Box Plot, Scatter Plot)



Flask Server

(Send list of ids to server)

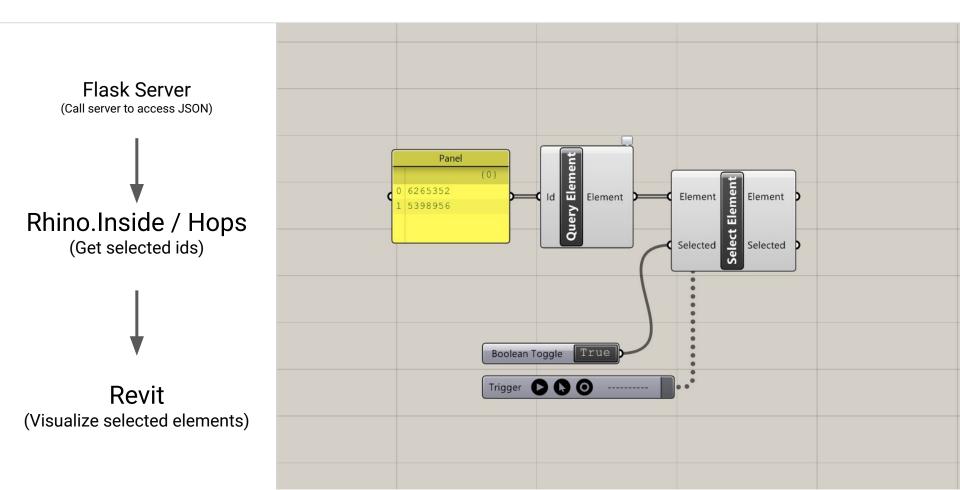
```
#Outlier Functions...
#Detecting Mutivariate Outliers
                                                                                                           #Detecting Univariate Outliers
#each parameter at the input should be numerical
                                                                                                          def univariate_outlier(df,variable,q1_percent,q3_percent):
def robust mahalanobis method(df):
                                                                                                              q1 = df[variable].quantile(q1_percent)
    #Minimum covariance determinant
                                                                                                              q3 = df[variable].quantile(q3 percent)
    rng = np.random.RandomState(0)
                                                                                                              iqr = q3-q1
    real cov = np.cov(df.values.T)
                                                                                                              inner fence = 1.5*igr
   X = rng.multivariate normal(mean=np.mean(df, axis=0), cov=real cov, size=506)
                                                                                                              outer_fence = 3*iqr
    cov = MinCovDet(random_state=0).fit(X)
   mcd = cov.covariance #robust covariance metric
                                                                                                              #inner fence lower and upper end
                                                                                                              inner_fence_le = q1-inner_fence
    robust mean = cov.location #robust mean
                                                                                                              inner_fence_ue = q3+inner_fence
    inv covmat = sp.linalg.inv(mcd) #inverse covariance metric
                                                                                                              #outer fence lower and upper end
    #Robust M-Distance
                                                                                                              outer_fence_le = q1-outer_fence
    x minus mu = df - robust mean
                                                                                                              outer_fence_ue = q3+outer_fence
   left term = np.dot(x minus mu, inv covmat)
    mahal = np.dot(left term, x minus mu.T)
                                                                                                              outliers prob = []
    md = np.sqrt(mahal.diagonal())
                                                                                                              outliers_poss = []
                                                                                                              for i in range(len(df[variable])):
    #Flag as outlier
                                                                                                                x = df[variable].values[i]
    outlier = []
                                                                                                                index = df[variable].take([i]).index[0]
   C = np.sqrt(chi2.ppf((1-0.00001), df=df.shape[1]))#degrees of freedom = number of variables
                                                                                                                if x < outer fence le or x > outer fence ue:
    for index, value in enumerate(md):
                                                                                                                  outliers prob.append(index)
        if value > C:
                                                                                                              for i in range(len(df[variable])):
            outlier.append(index)
                                                                                                                x = df[variable].values[i]
        else:
                                                                                                                index = df[variable].take([0]).index[0]
            continue
                                                                                                                if x <= inner fence le or x >= inner fence ue:
    outlier id = df.index[outlier]
                                                                                                                    outliers poss.append(index)
    return outlier id
                                                                                                              return outliers_prob, outliers_poss
```



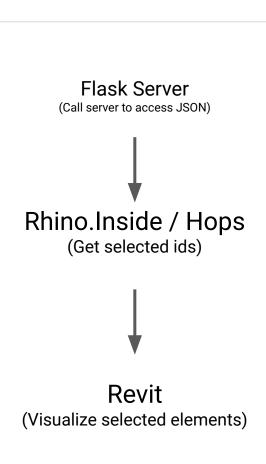
3. Visualize, Explore, and Select Outliers

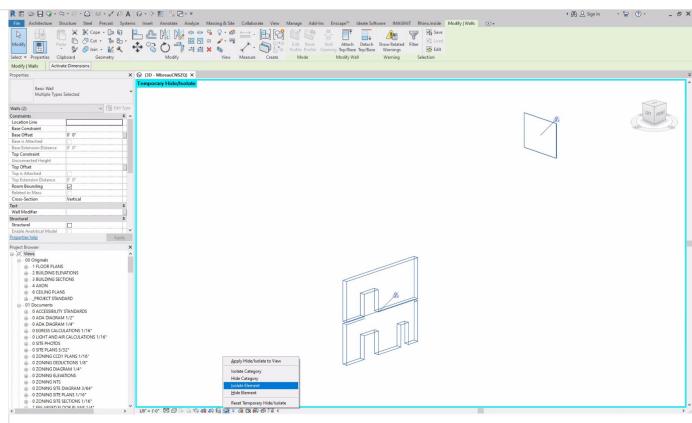
Flask Server (Call server to access JSON) Web Dashboard (React / Typescript / Plotly / Material-UI) Flask Server (Send list of ids to server)

4. View outliers in Revit



4. View outliers in Revit





Future Development

- Use Hops instead of Excel for data export
 Automate export/ analysis procedure
- Automatic highlighting in Revit
 Hops should return IDs into Grasshopper so the user does not need to copy & paste
- "Remember" User's Input
 Once the user has confirmed that a particular parameter on a particular element its not an outlier, that element/ parameter would not be flagged as an outlier again
- Recipes
 Learn from experience which parameters are likely to be worthwhile to cross-reference, and prioritize these for new projects.
- Automatic recognition of meta-patterns
 Implement multi-variable regressions



Thanks!

https://github.com/jperaino/cuckoo

