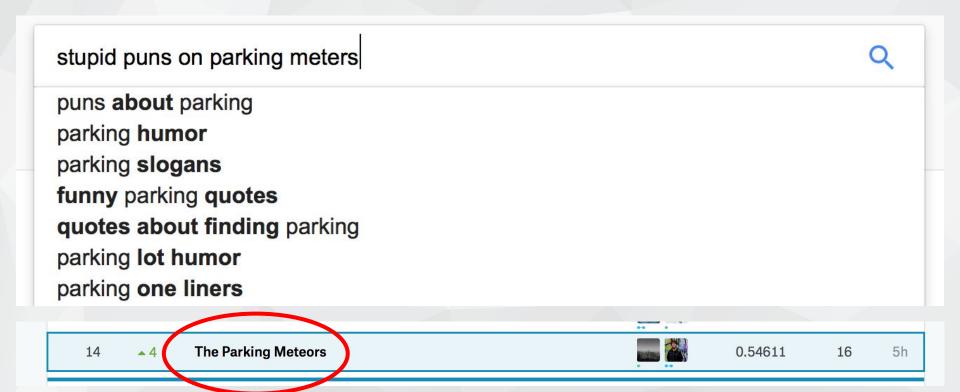


### Background Research



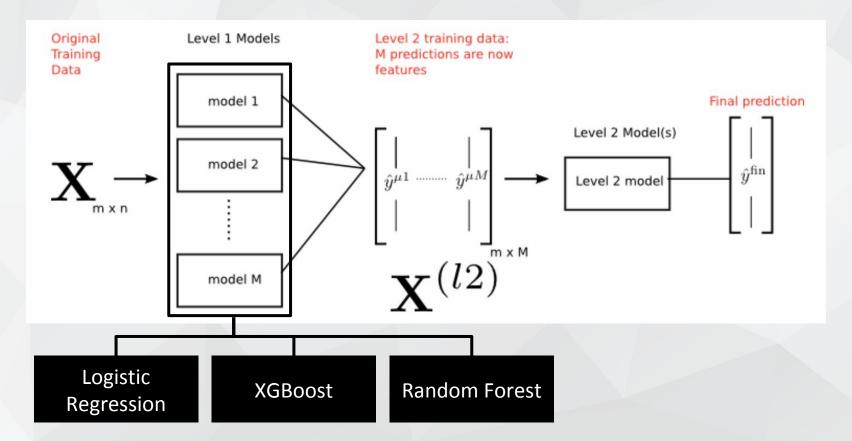
# Feature Engineering (1/2)

- Most important features: simple target encoding
- Datetime
  - o 'day\_part': morning/afternoon/evening/night
    - Target encoded features based on day\_part
  - 'minutes\_since\_midnight': high importance for RF + XGBoost
- Parkingrecords.csv
  - Mapped meters to street segments in training data
  - Several mean encoded features using transactions per street segment

# Feature Engineering (2/2)



## Machine Learning



#### Mlxtend: Concise Code for Stacking Grid Search

```
clf1 = LogisticRegression()
clf2 = RandomForestClassifier(n estimators=500, max features='sqrt')
clf3 = XGBClassifier()
lr = LogisticRegression()
sclf = StackingClassifier(classifiers=[clf1, clf2, clf3], meta classifier=lr)
params = { 'logisticregression C': [1, 10],
          'randomforestclassifier min samples leaf':[10, 25],
          'xqbclassifier n estimators': [20, 40, 60],
          'xgbclassifier learning rate': [0.001, 0.01, 0.1],
          'xgbclassifier subsample': [0.5, 0.75],
          'xgbclassifier max depth': [3, 5, 10],
          'xqbclassifier colsample bytree': [0.3, 0.5],
          'meta-logisticregression C': [10, 100]}
grid = GridSearchCV(estimator=sclf, param grid=params, cv=5, scoring=fhalf scorer, refit=True, n jobs=-1, verbose=1)
grid.fit(X, y)
```



#### Results & Conclusions

F0.5 Score	Precision	Recall
0.58	0.58	0.58



- 1. Starting with simple models and available data
- 2. Importance of background research
- 3. Benefits of a well-designed workflow