27.03.2017 Solution

In [4]:

```
from sklearn.ensemble import GradientBoostingRegressor
from sklearn.model_selection import cross_val_score
import pandas as pd
import xgboost as xgb
```

In [5]:

```
train = pd.read_csv("train.tsv")
test = pd.read_csv("test.tsv")
names = train.columns
print list(names)
```

```
['Num', 'y', 'year', 'week', 'shift', 'item_id', 'f1', 'f2', 'f3', 'f4', 'f5', 'f6', 'f7', 'f8', 'f9', 'f10', 'f11', 'f12', 'f13', 'f14', 'f15', 'f16', 'f17', 'f18', 'f19', 'f20', 'f21', 'f22', 'f23', 'f24', 'f25', 'f26', 'f27', 'f28', 'f29', 'f30', 'f31', 'f32', 'f33', 'f34', 'f35', 'f36', 'f37', 'f38', 'f39', 'f40', 'f41', 'f42', 'f43', 'f44', 'f45', 'f46', 'f47', 'f48', 'f49', 'f50', 'f51', 'f52', 'f53', 'f54', 'f55', 'f56', 'f57', 'f58', 'f59', 'f60']
```

In [6]:

```
frac = 0.2
part_train = train.sample(frac=frac, random_state=42)

X = part_train.drop(['Num','y'], axis=1)
y = part_train['y']
model = GradientBoostingRegressor(n_estimators=500, max_depth=5, random_state=43)
model.fit(X, y)
```

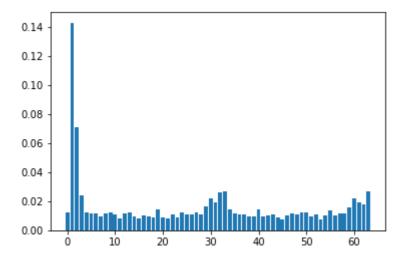
Out[6]:

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In [7]:

```
names = X.columns
print list(X)
print len(list(X))
from matplotlib import pyplot
print model.feature_importances_
print len(model.feature_importances_)
pyplot.bar(range(len(model.feature_importances_)), model.feature_importances_)
pyplot.show()
```

```
['year', 'week', 'shift', 'item_id', 'f1', 'f2', 'f3', 'f4', 'f5', 'f6', 'f7', 'f8', 'f9', 'f10', 'f11', 'f12', 'f13', 'f14', 'f15', 'f16', 'f17', 'f18', 'f19', 'f20', 'f21', 'f22', 'f23', 'f24', 'f25', 'f26', 'f27', 'f28', 'f
29', 'f30', 'f31', 'f32', 'f33', 'f34', 'f35', 'f36', 'f37', 'f38', 'f39',
   'f40', 'f41', 'f42', 'f43', 'f44', 'f45', 'f46', 'f47', 'f48', 'f49', 'f5
0', 'f51', 'f52', 'f53', 'f54', 'f55', 'f56', 'f57', 'f58', 'f59', 'f60']
64
[ 0.01252075  0.14303574  0.07131155  0.02418114  0.01221781  0.01188209
      0.01170501
                                        0.0098745
                                                                           0.01143587
                                                                                                              0.01212424
                                                                                                                                               0.01102528
                                                                                                                                                                                    0.00819361
                                                                           0.0093722
                                                                                                                                               0.01042351
     0.01169803
                                      0.01206517
                                                                                                              0.00832315
                                                                                                                                                                                    0.00956621
      0.00871316 0.01456915
                                                                           0.00849977
                                                                                                              0.00841509 0.01073536
                                                                                                                                                                                    0.00889267
                                        0.01083225
                                                                                                              0.01210024 0.01061919
      0.01256548
                                                                           0.01065334
                                                                                                                                                                                    0.01639528
                                        0.01919249
                                                                                                              0.02667149
                                                                                                                                               0.01454976
     0.02173276
                                                                           0.0262198
                                                                                                                                                                                    0.01128726
     0.01105401 0.0110169
                                                                           0.0095959
                                                                                                              0.00936221 0.01442775
                                                                                                                                                                                    0.00929186
     0.0103459
                                        0.01101017
                                                                           0.00918248 0.00756775 0.01025586
                                                                                                                                                                                    0.01159855
                                                                                                              0.00972417
                                                                                                                                                 0.01092428
     0.01073895
                                        0.01198172
                                                                           0.01202121
                                                                                                                                                                                    0.00751371
     0.01012561 0.01357801
                                                                           0.0100329
                                                                                                              0.01185768 0.01151517
                                                                                                                                                                                    0.01581406
      0.02167937 0.01937214 0.01811179 0.02669947
64
```



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```
In [8]:
y = train['y']
X = train.drop(['Num','y'], axis=1)
X = train[['year', 'week', 'shift', 'item_id', 'f60',
'f59', 'f58', 'f57', 'f53', 'f37', 'f30', 'f29', 'f28',
'f27', 'f26']]
model = xgb.XGBRegressor(max_depth=15,
                          n estimators=2000,
                          silent=False,
                          learning rate=0.25)
#model = GradientBoostingRegressor(n_estimators=500, max_depth=10, random_state=43)#1000 15
model.fit(X, y)
Out[8]:
XGBRegressor(base_score=0.5, colsample_bylevel=1, colsample_bytree=1, gamma=
0,
       learning_rate=0.2, max_delta_step=0, max_depth=15,
       min_child_weight=1, missing=None, n_estimators=1000, nthread=-1,
       objective='reg:linear', reg_alpha=0, reg_lambda=1,
       scale_pos_weight=1, seed=0, silent=False, subsample=1)
In [12]:
def smape(F, A):
    res = 0
    for i in range(len(F)):
        res += float(abs(F[i] - A[i]) / (abs(A[i]) + abs(F[i])))
    res *= float(200) / len(F)
    return res
train_y = model.predict(train[['year', 'week', 'shift', 'item_id', 'f60',
'f59', 'f58', 'f57', 'f53', 'f37', 'f30', 'f29', 'f28',
'f27', 'f26']])
print smape(train['y'], train_y)
preds = model.predict(test[['year', 'week', 'shift', 'item_id', 'f60',
'f59', 'f58', 'f57', 'f53', 'f37', 'f30', 'f29', 'f28',
'f27', 'f26']])
solution = pd.read csv("sample submission.tsv")
solution['y'] = preds
solution.to_csv("solution.tsv", sep=',', index=False)
6.83943974255
```

```
In [ ]:
```

In []: