

# Practical Lessons from Predicting Clicks on Ads at Facebook

GBDT+LR

# GBDT

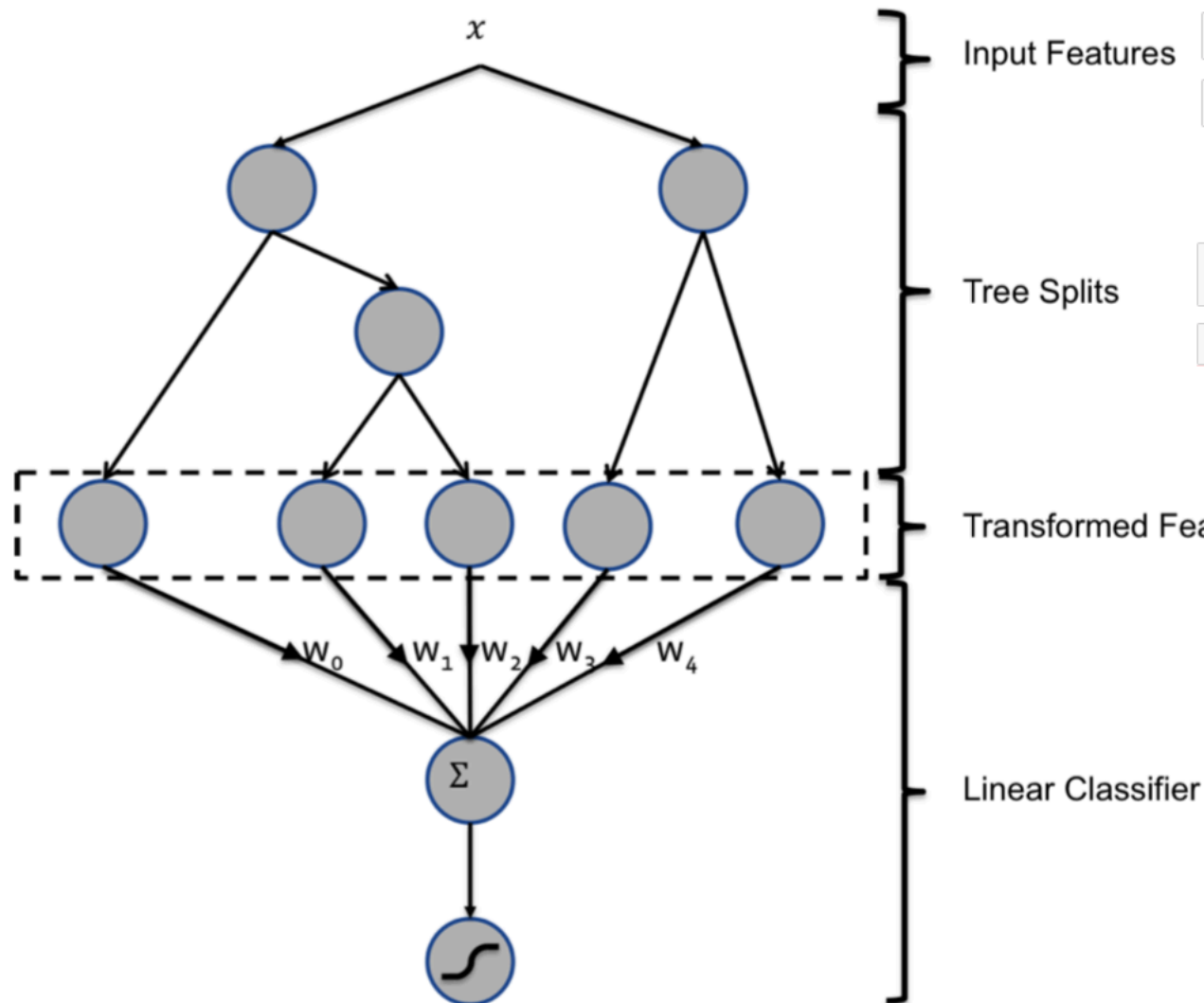
Gradient Boosting Decision Tree

$$\hat{y}_i = \sum_{k=1}^K f_k(x_i), f_k \in F$$

# NE

## Normalized Entropy

$$NE = \frac{-\frac{1}{N} \sum_{i=1}^n \left( \frac{1+y_i}{2} \log(p_i) + \frac{1-y_i}{2} \log(1-p_i) \right)}{-(p * \log(p) + (1-p) * \log(1-p))}$$



```
1 X_train = sparse.csc_matrix((data, (row, cols)), shape=[data_size, data_feature_num])
```

```
1 Y_train = np.asarray(y).astype(np.int64)
```

```
1 Y_train = np.expand_dims(Y_train, axis=-1)
```

```
1 gbm1 = GradientBoostingClassifier(n_estimators=feature_num, random_state=10, subsample=0.6, max_depth=3,
2 min_samples_split=900)
```

```
1 gbm1.fit(X_train, Y_train)
```

```
1 train_new_feature = gbm1.apply(X_train)
2 train_new_feature = train_new_feature.reshape(-1, feature_num)
```

```
1 enc = OneHotEncoder()
2 enc.fit(train_new_feature)
```

```
1 train_new_feature2 = np.array(enc.transform(train_new_feature).toarray())
```

```
1 reg = LinearRegression().fit(train_new_feature2, Y_train)
```