General approaches for metrics optimization

Overview

- Loss vs metric
- Approaches to metrics optimization in general



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Synonyms: loss, cost, objective



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Custom loss for XGBoost

Define an 'objective':

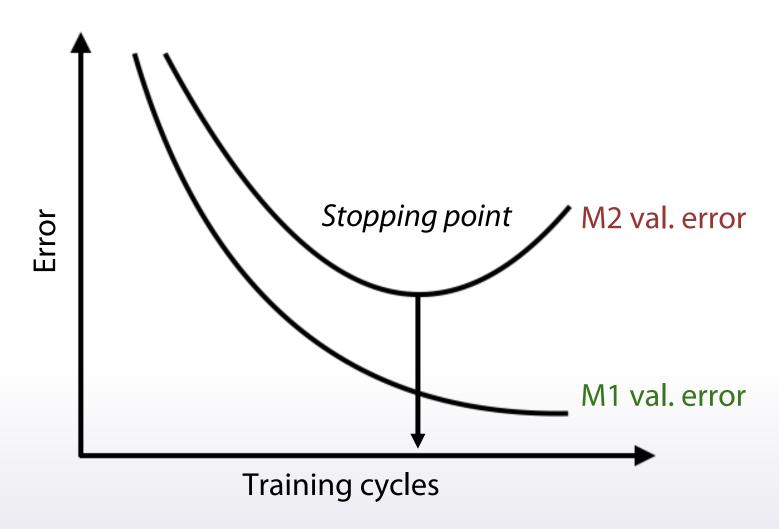
 function that computes first and second order derivatives w.r.t. predictions.

```
def logregobj(preds, dtrain):
labels = dtrain.get_label()
preds = 1.0 / (1.0 + np.exp(-preds))
grad = preds - labels
hess = preds * (1.0-preds)
return grad, hess
```

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 - Any

Early stopping

- Optimize metric M1, monitor metric M2
 - Stop when M2 score is the best



Conclusion

- Loss vs metric
- Approaches in general:
 - Just run the right model
 - Preprocess train and optimize another metric
 - Optimize another metric, postprocess predictions
 - Write a custom loss function
 - Optimize another metric, use early stopping