# Profit Maximizing Machine Learning

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# ML is evaluated on:





Accuracy, Precision, Recall, Specificity, F1 score, Area Under the Curve, ...



## Businesses care about:



#### Maximum LIKELIHOOD credit scoring

- ✗ Binary goal: will applicant DEFAULT or not?
  → Classification
- ✗ Basel II/III requirements: explainable model
  → Logistic Regression, Trees, ...
- X Can't use some features such as: race, gender, ...



#### Maximum PROFIT credit scoring

#### <u>Combine</u>

**X** Logistic Regression: explainable, easy

$$P(Y = ext{default} \mid \mathbf{x}_i) = rac{e^{eta_0 + eta^T x_i}}{1 + e^{eta_0 + eta^T x_i}}$$

**X** Expected Maximum Profit (EMP) measure: profit! [Verbraken et al.], asymmetric cost problems

EMP = empCreditScoring(predictions, true\_classes, p0=0.55, p1=0.1, ROI=0.2644)

#### **X** Genetic Algorithm: (lazy) optimization

lr\_weights = GA(lr\_weights, maximize=EMP)



#### 2,427,683 \$ +31.2 %, AUC .78 MLE Logistic Regression



EMP Logistic Regression

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# THANKS!

### Any questions?

You can find me at

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Reference: *"Profit Maximizing Logistic Regression Modeling for Credit Scoring"*, Devos et al. IEEE DSW 2018

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