Technology Fundamentals for Analytics Lab

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Agenda

- 1. Exam Details
- 2. Guest Speaker NICK ELPRIN Founder http://www.dominodatalab.com
- 3. Kaggle Details [Presentation]
- 4. Evaluating Models
- 5. Deploying Models [Cases Chequed.com and iNovum.]

Nick Elprin

http://www.dominodatalab.com

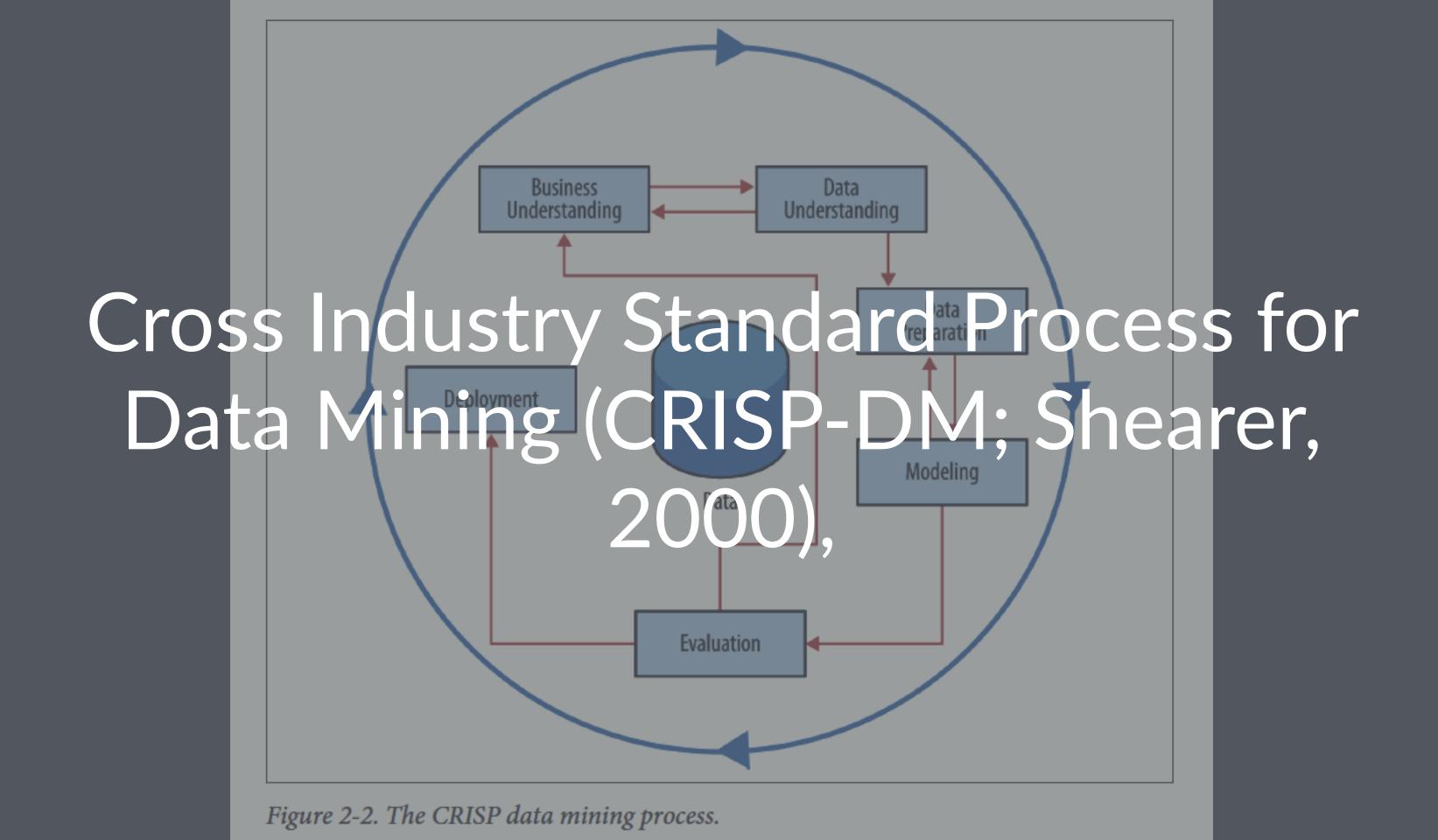
Domino Data Lab, Inc. is a new company that started out with a focus on enabling much easier cloud computation, and doing "version control for data science."

Exam Details

- You should be able to go from statement to R code. You can use a 1 page both sides cheat sheet.
- You should be able to go from R code to a statement describing what it does.
- You should have conceptual knowledge of Hadoop and Mapreduce [not specific syntax].
- You should be able to understand deployment of models for cost benefit analysis.

Kaggle

Cross Industry Standard Process for Data Mining (CRISP-DM; Shearer, 2000),



Stages of Model Development

Pay attention we will use this as a framework

- 1. Data understanding
- 2. Data preparation
- 3. Modeling
- 4. Evaluation < Focus today
- 5. Deployment (DDD) <- Focus today
- 5. Business Understanding

What is a good model?

What is a good model?

- 1. Predicts on data that the model has not been trained on. (This indicates that overfitting is not likely an issue).
- 2. Minimizes errors (classification or prediction)
- 3. (maybe) Understandible/parsimonious

Evaluation

What is a good model? Classification

Accuracy = Number of correct decisions/Total number of decisions

Prediction (Continuous DV)

- Root-mean-square error (RMSE)
- Root Mean Squared Logarithmic Error (RMSLE)

Classification Accuracy Confusion Matrix

		Predicted class		
		Cat	Dog	Rabbit
ass	Cat	5	3	0
Actual class	Dog	2	3	1
	Rabbit	0	2	11

Prediction (Continuous DV)

Root-mean-square error (RMSE)

$$RMSE = \sqrt{\frac{1}{N} \sum_{i=1}^{N} (y_i - x_i)^2}$$

Prediction (Continuous DV)

Root Mean Squared Logarithmic Error (RMSLE)

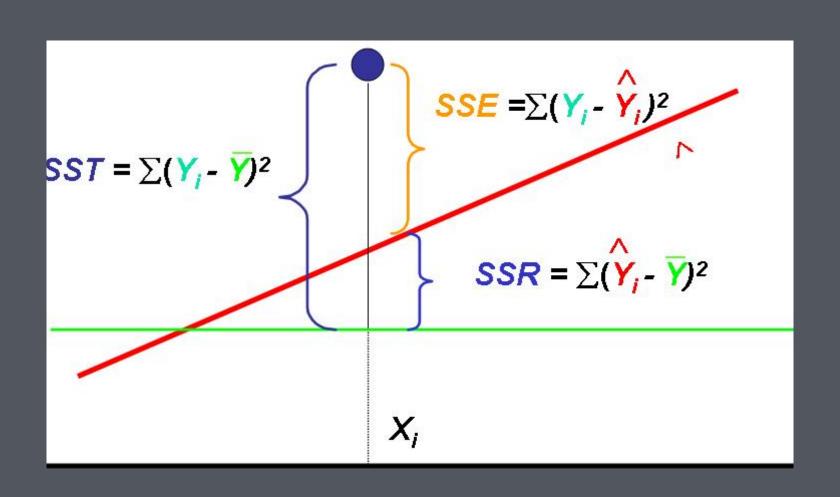
Submissions are evaluated one the Root Mean Squared Logarithmic Error (RMSLE). The RMSLE is calculated as

$$\sqrt{\frac{1}{n} \sum_{i=1}^{n} (\log(p_i + 1) - \log(a_i + 1))^2}$$

Where:

- n is the number of hours in the test set
- *p_i* is your predicted count
- a_i is the actual count
- log(x) is the natural logarithm

Prediction (Continuous DV) Coefficient of Determination (R2)



Coefficient of Deternination
$$ightarrow$$
 $R^2 = \frac{SSR}{SST} = 1 - \frac{SSE}{SST}$

Sum of Squares Total $ightarrow$ $SST = \sum (y - \bar{y})^2$

Sum of Squares Regression $ightarrow$ $SSR = \sum (y' - \bar{y'})^2$

Sum of Squares Error $ightarrow$ $SSE = \sum (y - y')^2$

What if one class is prevalent?

• Example, Fraud occurs in less than 1/100.

What if there is Unequal Costs and Benefits

Difference between a false positive and a false negative.



HR Analytics Background - Survey Based methods

- 1. Multiple measures per construct
- 2. Data reduction
- 3. Regression
- 4. Deploy to predict job success

1. Multiple measures per construct.

Example

Neuroticism is a fundamental personality trait in the study of psychology characterized by anxiety, fear, moodiness, worry, envy, frustration, jealousy, and loneliness.

1. Multiple measures per construct.

Construct = Neuroticism Part of Big 5

I am not a worrier.

I am seldom sad or depressed.

I often feel helpless and want someone else to solve my problems.

At times I have been so ashamed I just wanted to hide.

I often feel inferior to others.

When I'm under a great deal of stress, sometimes I feel like I'm going to pieces.

I rarely feel lonely and blue.

I often feel tense and jittery.

1. Multiple measures per construct.

Survey on Likert Scale.

- 1. Strongly disagree
- 2. Disagree
- 3. Neither agree nor disagree
- 4. Agree
- 5. Strongly agree

2. Average multiple measures from construct, reverse coding where necessary

- I rarely feel fearful or anxious. [6-Value]
- Too often, when things go wrong, I get discouraged and want to give up. [Value]

3. Regression

• Regression using multiple constructs and a dependent variable likely to be context specific.

4. Deploy to predict job success

Variables are weighted in relationship to significance

HR Analytics DV - Ranking of salespeople

- We take our existing sales
- How might we evaluate the model?

HR Analytics DV - Ranking of salespeople

- 1. R2 [Easy because is outcome of regression and fits with IO psychology
- 2. RMSE based on rank
- 3. Classify top 20%. Either individual is in top 20% or not.
- 4. Classify in 5 different sections (0-20, 21-40, etc.)

Deploying Model

Deploying Model

How do you use the HR analytics?

Chequed fit score from 1-5

https://www.chequed.com/pre-employment-assessment-test/

Take 5 minutes and find an example of a model deployed in the wild.

Examples

Deploy - Classification

Consider

• For this case study, let's consider a real example of targeted marketing: targeting the best prospects for a charity mailing.

[It costs \$1 to print and send a brochure. How might I improve the likelihood of a positive outcome?]

Expected Value

Expected Benefit = Pr(x) * Vr + [1 - Pr(x)] * Vnotr

What business problem are we solving.

p(x)\$99 - [1-p(x)]\$1 > 0.

What other factors should be considered?

What factors should be considered?

- Some people might give \$1 others \$1000.
- There could be a negative cost of predicting wrong that differs depending

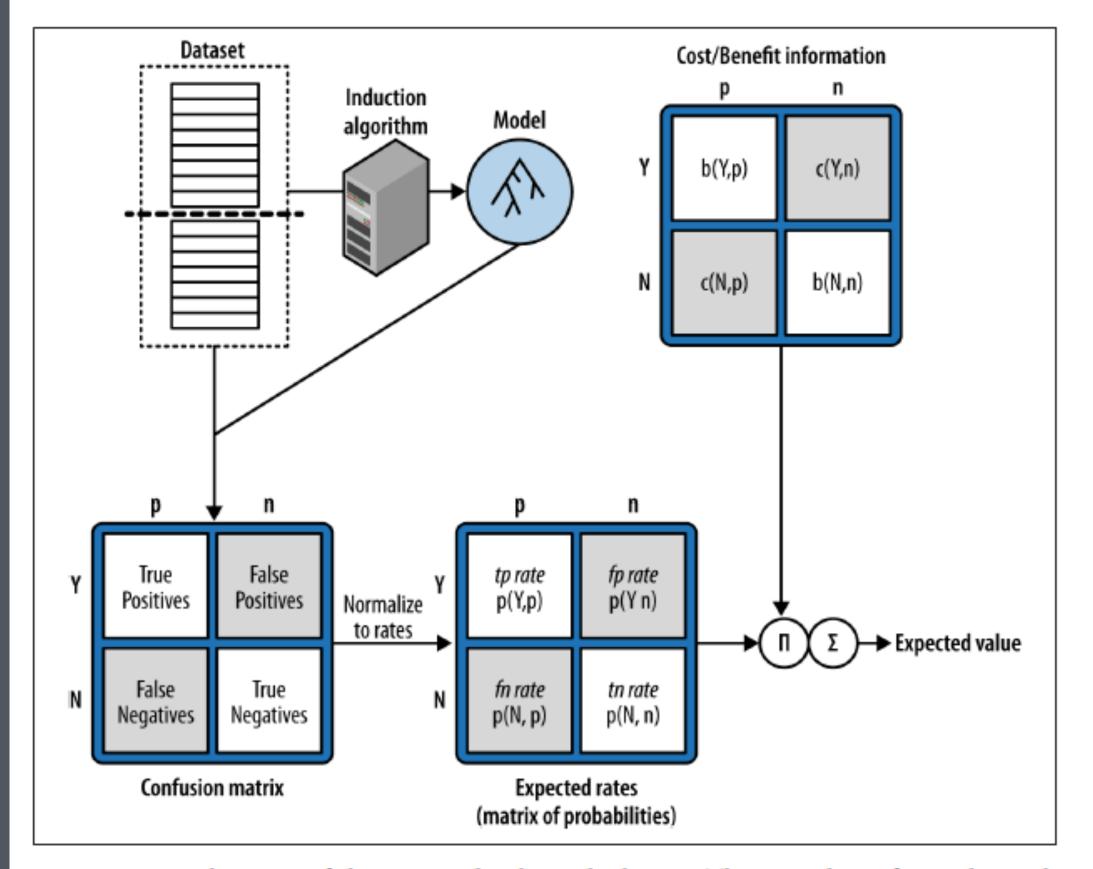


Figure 7-2. A diagram of the expected value calculation. The Π and Σ refer to the multiplication and summation in the expected value calculation.

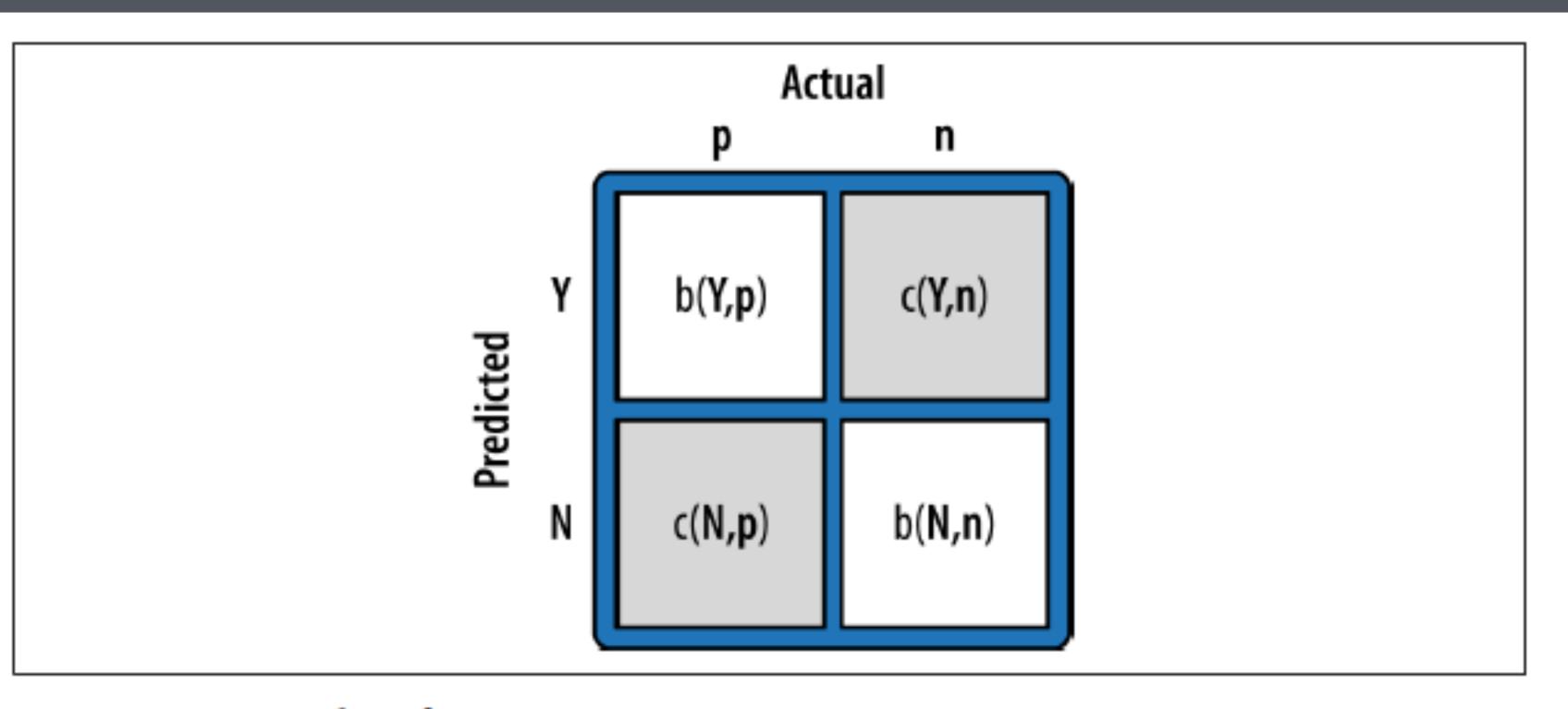


Figure 7-3. A cost-benefit matrix.

While probability can be estimated from the data, cost benefit is a separate process.

b(Y,p) and b(N, n) are the benefits of correct prediction.

c(N,p) and c(Y, n) are the costs of an incorrect prediction.

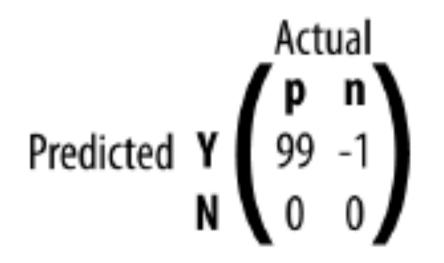


Figure 7-4. A cost-benefit matrix for the targeted marketing example.

Given a matrix of costs and benefits, these are multiplied cell-wise against the matrix of probabilities, then summed into a final value representing the total expected profit. The result is:

Expected profit =
$$p(\mathbf{Y}, \mathbf{p}) \cdot b(\mathbf{Y}, \mathbf{p}) + p(\mathbf{N}, \mathbf{p}) \cdot b(\mathbf{N}, \mathbf{p}) + p(\mathbf{N}, \mathbf{p}) \cdot b(\mathbf{N}, \mathbf{p}) + p(\mathbf{N}, \mathbf{n}) \cdot b(\mathbf{N}, \mathbf{n}) \cdot b(\mathbf{N}, \mathbf{n}) + p(\mathbf{Y}, \mathbf{n}) \cdot b(\mathbf{Y}, \mathbf{n})$$

Share Examples on Progress with Kaggle 2

Interesting Book Resources to Check Out Machine Learning for Hackers

https://github.com/johnmyleswhite/MLforHackers

