

**clusterAI 2020**  
**ciencia de datos en ingeniería**  
**industrial**  
**UTN BA**  
**curso I5521**

## **clase\_04: ML Strategy**

**Presentación: Matías Callara**

# agenda clase04: ML Strategy

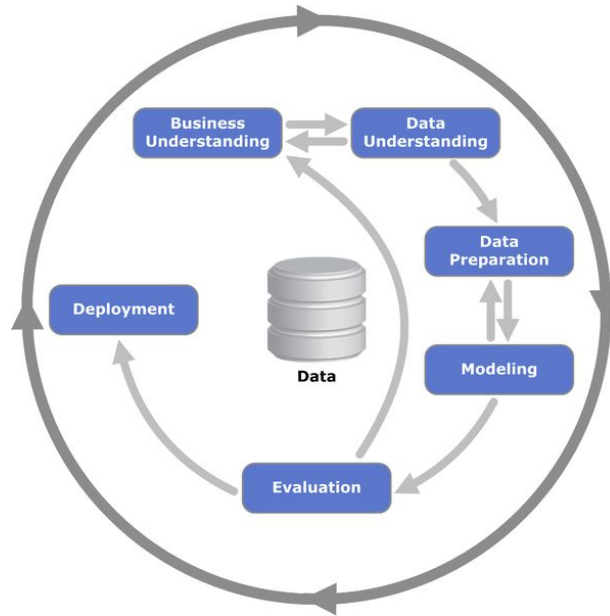
- This class is a summary of [Andrew Ng's great class on ML Strategy](#). He is a much better communicator than me on this topic so I suggest you go and join his course and enjoy a great teacher, teaching material he knows and is passionate about. 🙌

# agenda clase04: ML Strategy

- Data Science Project Phases
- Orthogonalization
- Evaluation metric

# Solving problems with ML

## Data Science Project Phases



[Kenneth Jensen / CC BY-SA \(https://creativecommons.org/licenses/by-sa/3.0\)](https://creativecommons.org/licenses/by-sa/3.0/) - [CRISP-DM Process Diagram](#)

# Orthogonalization

## Reaching the target

Use sliders  $a$  &  $b$  to reach target  $T$ .

- First, try to reach  $T$  with point  $NO$  and then with point  $O$ . Which one was easier?

From Linear Algebra. If  $x_1, x_2$  are orthogonal, I can solve a system of equations  $Ax = b$  by solving one equation at a time  $a_{11}x_1 + a_{12}x_2 = b_1$ ,  $a_{21}x_1 + a_{22}x_2 = b_2$ .

# Working as a Data Scientist

Get ready to compete!



**(With a lot of effort) I manage to get very small error  
on the training set. Am I done?**



**(With a lot of effort) I manage to get very small error on the training set. Am I done?**

Yes

No



Tc

0



**(With a lot of effort) I manage to get very small error on the training set. Am I done?**

Yes

No

✓ 0%



# Leaderboard



**(With a lot of effort) I manage to get very small error  
on the validation set. Am I done?**



**(With a lot of effort) I manage to get very small error on the validation set. Am I done?**

Yes

No



Tc

0

**(With a lot of effort) I manage to get very small error on the validation set. Am I done?**

Yes

No

✓ 0%



# Leaderboard



**(With a lot of effort) I manage to get very small error  
on the test set. Am I done?**



**(With a lot of effort) I manage to get very small error on the test set. Am I done?**

Yes

No



Tc

0



**(With a lot of effort) I manage to get very small error on the test set. Am I done?**

Yes

No

✓ 0%



# Leaderboard



**(With a lot of effort) I manage to get very small error  
in the real world. Am I done?**



**(With a lot of effort) I manage to get very small error in the real world. Am I done?**

Yes

No

Tc



0

**(With a lot of effort) I manage to get very small error in the real world. Am I done?**

Yes

No

✓ 0%



# Leaderboard



# Assumptions

- Getting better training error
  - More flexible model (underfitted)
  - Optimization procedure
- Getting better dev (validation) error
  - Regularization (overfitted)
  - More training data
- Getting better test error
  - More dev set
- Getting better Real Life error
  - Change dev set
  - Change cost function

# Assumptions

We need to do well in:

- Train
  - More flexible model (underfitted)
  - Optimization procedure
- Dev
  - Regularization (overfitted)
  - More training data
- Test
  - More dev set
- Real Life
  - Change dev set
  - Change cost function



# Evaluation: Evaluation Metric

Evaluation vs Loss function

Example:

Binary Classification:

Loss function: Cross Entropy

Evaluation Metric: F1

# Single Evaluation Metric

Having a single real number evaluation metric will help you compare the performance of different models.

Combining different metrics

# Satisficing and Optimizing Metrics

In the problem understanding phase we need to identify which are objectives we need to optimize for and which are conditions that we need to satisfy.

Another cat classification example



Classifier	Accuracy	Running time
A	90%	80ms
B	92%	95ms
C	95%	1,500ms

# Let's get (a bit) philosophical

The way we learn, do, learn, do, learn, do ... (DS as a tool)

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# Let's get (a bit) philosophical

In Mathematics, we try to come up with abstractions (structures, patterns) that will help us deal with complex concepts, apply them to other situations (context) or compare them with other perspective (abstractions) of the same phenomena.

Compare  $Ax = b$  vs  $a_{11} x_1 + a_{12} x_2 = b_1, a_{21} x_1 + a_{22} x_2 = b_2$ .

How does this relate to ML?

# For a next session...

- . Relación entre funciones y vectores.
- . Funciones como vectores infinitos.
- . Matrices como funciones lineales.
- . Operadores como “funciones” para funciones.
- Least squares