

#### **Frameworks**

MinoTauro Accounts
Connecting to MinoTauro
MinoTauro jobs and params

- Many alternatives
  - Caffe2 (by Berkeley)
  - Microsoft Cognitive Toolkit (aka CNTK)
  - MXNet (by Apache)
  - PyTorch (by Facebook)
  - TensorFlow (by Google)
  - PaddlePaddle (by Baidu)
  - Keras (on top of TF, Theano, CNTK)
  - •



#### **Frameworks**

**MinoTauro Accounts** 

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- One account per registered student
- Each student is responsible of its own account
- BSC clusters sometimes have down time
  - We will inform you promptly
  - Deadlines will be addapted if needed
- Cluster priorities
  - Based on resources and time
  - Don't leave everything for the last 3 days



Frameworks
MinoTauro Accounts

**Connecting to MinoTauro** 

**MinoTauro jobs and params** 

- Connect through SSH
  - ssh username@mt1.bsc.es
- Work on your home directory
  - /home/nct01/username
- To change your password, use the dt node
  - ssh username@dt01.bsc.es
  - passwd
- Use dt also for moving data

```
(DW) scp username@dt01.bsc.es:gpfs_path local_path (UP) scp local_path username@dt01.bsc.es:gpfs_path
```

- MT has no outside access!
- Check the MinoTauro user manual online



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**MinoTauro jobs and params** 

- To execute you must submit a job to a queue
  - bsc\_cs: max 48h
  - debug: max 1h (faster to get in)

```
#SBATCH--partition=debug
```

- Time to get enqueued depends on resources and time asked, and quota
- Giving a maximum execution time (hard limit!)

```
#SBATCH--time=HH:MM:SS
```

Initial execution path

```
#SBATCH--workdir=pathname
```

Error and log file

```
#SBATCH--error=file_name_%j.err
#SBATCH--output=file name %j.out
```



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- Define resources to use
  - Just one CPU

```
#SBATCH --ntasks=1
```

Just one GPU (unless you parallelize)

```
#SBATCH --gres gpu:1
```

- Define software needed (modules to load)
  - This includes Python 3.6 + Keras + TF

```
module purge; module load K80/default impi/2018.1 mkl/2018.1 cuda/8.0 CUDNN/7.0.3 python/3.6.3 ML
```



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#### Example of launcher file

```
#!/bin/bash
#SBATCH --job-name="test job"
#SBATCH --workdir=.
#SBATCH --output=mnist %j.out
#SBATCH --error=mnist %j.err
#SBATCH --ntasks=1
#SBATCH --gres gpu:1
#SBATCH --time=00:02:00
module purge; module load K80/default impi/2018.1
mkl/2018.1 cuda/8.0 CUDNN/7.0.3 python/3.6.3 ML
python some code.py
```



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Launching jobs

```
sbatch launcher_file.cmd
```

Checking the status of jobs

```
squeue
```

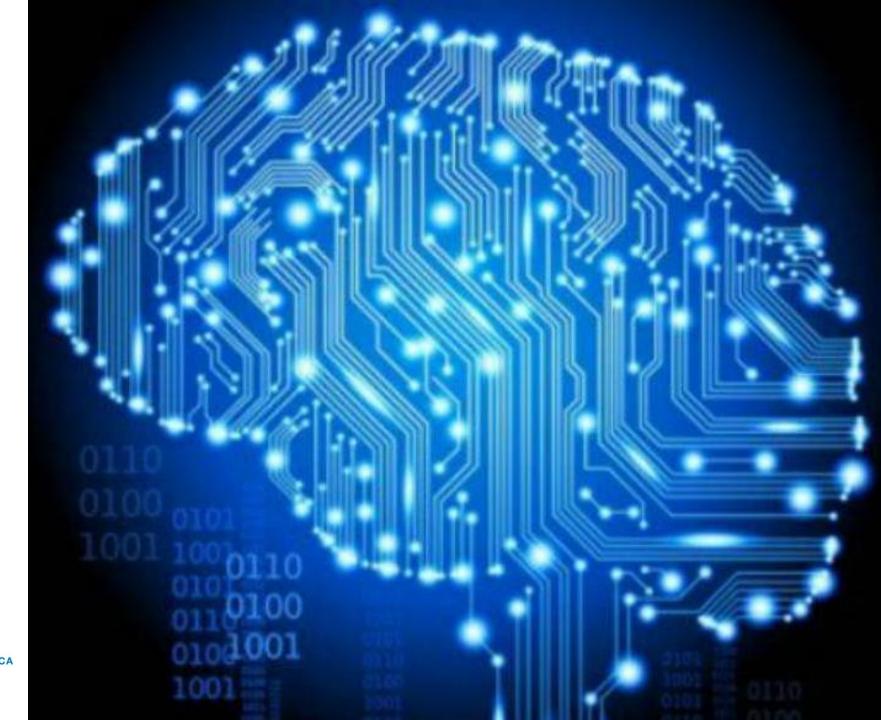
Killing jobs

```
scancel jobId
```

Checking a job "live"
 squeue (get computing node)
 ssh node (from within login node)



# MNIST example



# MNIST example

### Lets get the data

```
wget https://s3.amazonaws.com/img-
datasets/mnist.npz
```

#### Upload it to the cluster

```
scp mnist.npz
nct01058@dt01.bsc.es:/home/nct01/nct0
10XX/.keras/datasets
```

(you will need to create that directory first)



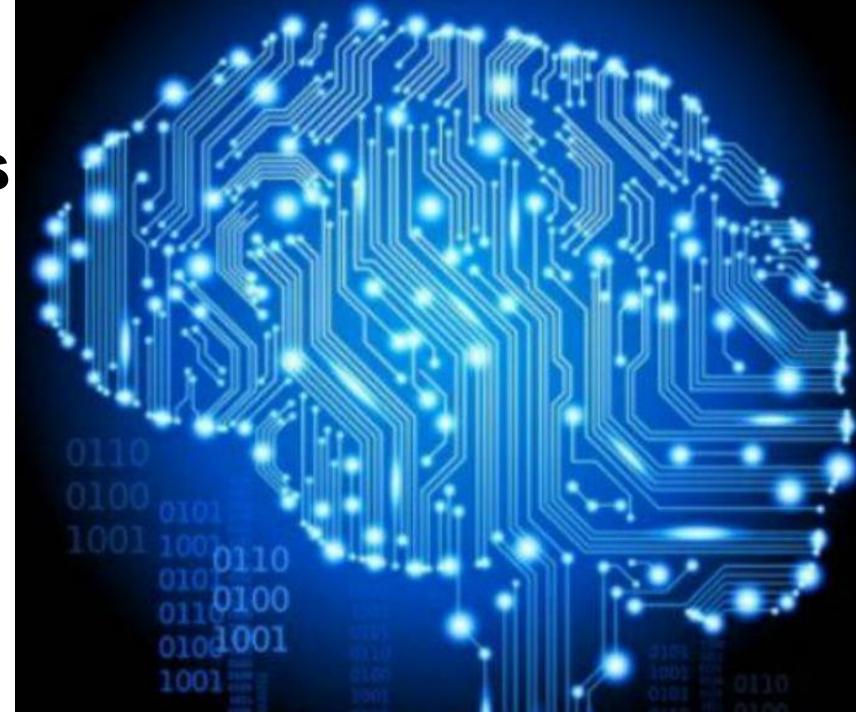
# MNIST example

https://github.com/UPC-MAI-DL/UPC-MAI-DL.github.io/blob/master/\_codes/1.FNN-CNN/mnist\_fnn\_example.py

https://github.com/UPC-MAI-DL/UPC-MAI-DL.github.io/blob/master/\_codes/1.FNN-CNN/mnist\_cnn\_example.py



# More experiments



# More experiments

 You can start with CIFAR10. Its relatively hard but fast to compute. Also, in color (channels!).

```
wget http://www.cs.toronto.edu/~kriz/cifar-10-
python.tar.gz
```

- You may need to rename the file, once uploaded to .keras/datasets to cifar-10-batches-py.tar.gz
- If you have problems, run the command locally, to get the files, and then upload them to MT.



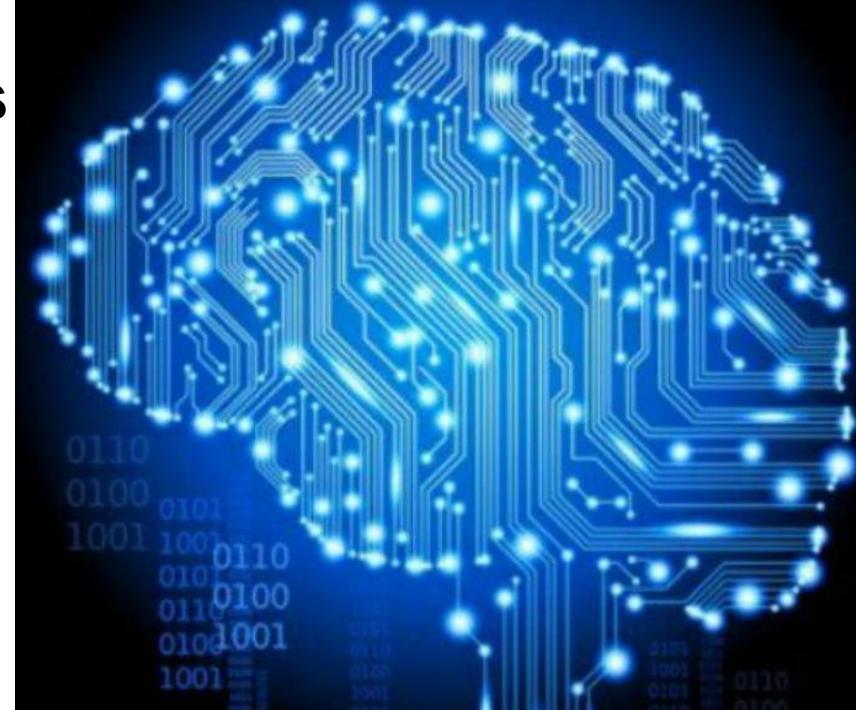
# More experiments

- MNIST and CIFAR datasets come preprocessed for Keras (from keras.datasets import cifar10). This is rarely the case.
- Start by looking for a new dataset, and figuring out how to load it into keras (e.g., [1])

[1]

https://blog.keras.io/ building-powerfulimage-classificationmodels-using-verylittle-data.html





- Next week: Time for your autonomous lab.
- You are supposed to work on your reports. We will be around to answer questions.
- Delivery dates on the web.
- PDF document, 8 pag. (weak restriction).
   Delivered through Raco.
- The student must use coherently deep learning components, making proper analysis of the results obtained, and taking reasonable decisions.



- Goal: The student must make coherent use of deep learning techniques, making proper analysis of the results obtained, and taking reasoned decisions.
- The topic of the report is open to your interests and curiosity.
- Simple options:
  - Choose a problem and train a network to solve it. Use methods coherently.
  - Choose a technique and analyze its impact and behavior in depth.



- What will be valued
  - Results obtained are properly assessed, leading to well argued experiments
    - No random experiments!
  - Report is written in a clear, scientific manner
    - Support your claims!
  - Show plots to illustrate your findings
    - Accuracy/Loss at least!
    - Captions, legends, ...
  - Originality
    - NO MNIST please



#### What will NOT be valued

- Too many experiments on too many aspects without getting into any depth
  - No mindless experiments!
- Questions left open or unanswered, to perform experiments on a different unrelated method.
- Excuses regarding the cluster.
- Reports well over the page limit.
- Repeating stuff said in class.
- Introducing basic concepts.



#### TIPS

- Introduce the data you work with.
   Understand its particularities (size, resolution, sample distribution, variance, etc)
- Define the data splits
- Always train your model until overfit.
- After generating a plot, stare at it for 5 min.
   Make yourself questions and act in consequence. Show this process in the report.
- You don't need to plan ahead. Do an experiment and see what questions it raises. Follow those.



### An very incomplete list of datasets

- http://web.mit.edu/torralba/www/indoor.html (MIT67)
- http://www.vision.caltech.edu/visipedia/CUB-200.html (CUB200)
- http://www.robots.ox.ac.uk/~vgg/data/flowers/102/ (FLOWERS102)
- https://upc-mai-dl.github.io/mlp-convnets-labautonomous/vision.stanford.edu/aditya86/StanfordDogs/ (SDOGS)
- https://upc-mai-dl.github.io/mlp-convnets-labautonomous/www.vision.caltech.edu/lmage\_Datasets/Caltech\_ 101/ (CALTECH101)
- https://upc-mai-dl.github.io/mlp-convnets-labautonomous/www.vision.ee.ethz.ch/datasets\_extra/food-101 (FOOD101)
- <a href="https://www.robots.ox.ac.uk/~vgg/data/dtd/">https://www.robots.ox.ac.uk/~vgg/data/dtd/</a> (TEXTURES)
- <a href="https://upc-mai-dl.github.io/mlp-convnets-lab-autonomous/www.ee.oulu.fi/~olli/Projects/Lumber.Grading.html">https://upc-mai-dl.github.io/mlp-convnets-lab-autonomous/www.ee.oulu.fi/~olli/Projects/Lumber.Grading.html</a> (WOOD)

