# Malicious Input Detection for Deep Neural Networks

Emilio R. Balda





## **Outline**

- Personal Background
  - home country
  - bachelor studies
- The Ilmenau Experience
  - life as a MSCSP student in Ilmenau
  - a tensor-based master thesis
- From Ilmenau to Aachen
  - the Institute for Theoretical Information Technology
  - our research fields
- Technical Talk
  - introduction to neural networks
  - malicious input detection
  - results
- Personal advice





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## **Personal Background**

#### **Home Country**

- My full name is Emilio Rafael Balda Cañizares
- I was born in the city of Guayaquil, located on the coast of Ecuador



#### **Ecuador**

Language: Spanish
Population: 16 Million
Area: 283,560 km²

#### Guayaquil

Elevation: 0 mts

Temperature: 25 to 30 °C Humidity: 65 to 95%

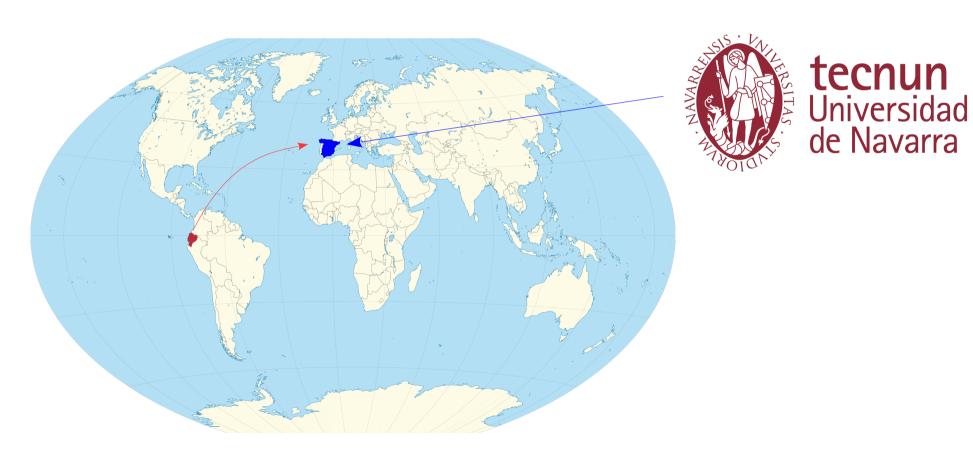




## **Personal Background**

#### **Bachelor Studies**

- Bachelor in Telecommunication Systems Engineering
  - at University of Navarra, Spain
  - in the Engineering faculty known as TECNUN







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#### Life as a MSCSP Student in Ilmenau

- Personal impression of the MSCSP program
  - strongly research oriented, the perfect choice for future Ph.D. candidates
  - highly specialized on communication networks and signal processing
  - provides several opportunities to explore and be creative on different research topics





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  - large variety of activities and clubs
  - Its easy to adopt the comfortable Ilmenau way of life
  - a 2 years stay allows you to experience many of the activities that Ilmenau offers







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- I personally can say that I really enjoyed my studies in Ilmenau





#### A Tensor-based Master Thesis

 Master Thesis Title: "Perturbation analysis of tensor-based algorithms" - Advisor: Prof. Dr.-Ing. Martin Haardt





#### A Tensor-based Master Thesis

- Master Thesis Title: "Perturbation analysis of tensor-based algorithms" - Advisor: Prof. Dr.-Ing. Martin Haardt
  - Conducted a theoretical perturbation analysis of:
    - the truncated Higher-Order Singular Value Decomposition (HOSVD), mainly used for dimensionality reduction [1]
    - the Joint Eigen-Value Decomposition (JEVD), mainly used for data analysis [2]
    - the Canonical Polyadic Decomposition (CPD), mainly used for data analysis (conference version currently being written)

[1]	E. R. Balda, S. A. Cheema, J. Steinwandt, M. Haardt, A. Weiss, and A. Yeredor, "First-order perturbation analysis of low-rank tensor approximations based on the truncated HOSVD," in <i>Proceedings of ASILOMAR</i> , Nov. 2016
[2]	E. R. Balda, S. A. Cheema, A. Weiss, M. Haardt, and A. Yeredor, "Perturbation Analysis of Joint Eigenvalue Decomposition Algorithms," in <i>Proceesings of ICASSP</i> , March. 2017





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- is led by Prof. Dr. Rudolf Mathar
  - Head of the Institute
  - Pro-Rector of Research and Structure at RWTH Aachen University





- The Institute for Theoretical Information Technology (TI)
  - is located in Aachen, Germany







## The Institute for Theoretical Information Technology

- The Institute for Theoretical Information Technology (TI)
  - is located in Aachen, Germany
  - on the 3<sup>rd</sup> floor of the ICT Cubes of the RWTH Aachen Univeristy [BA+11]



[BA+11]

Böcherer, G., Altenbach, F., Malsbender, M., & Mathar, R. "Writing on the facade of RWTH ICT Cubes: Cost constrained geometric Huffman coding." in Proceedings of *IEEE Wireless Communication Systems (ISWCS)*, 2011.



- Staff:
  - Professor Dr. Rudolf Mathar is head of the institute
  - Prof. Dr-Ing. Anke Schmeink is head of the research group Information Theory and Systematic Design of Communication Systems









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- 2 + 22 academic staff (April 2017)
- 4 non-academic staff (April 2017)
- 5 student assistants (April 2017)





#### Our Research Fields

- Information Theory & Communication Theory
- Network Design, Control & Optimization
- OFDM Systems
- Compressed Sensing & Signal Classification
- Planning, Simulation, Evaluation for Energy Grids





#### Our Research Fields

- Information Theory & Communication Theory
- Network Design, Control & Optimization
- OFDM Systems
- Compressed Sensing & Signal Classification
- Planning, Simulation, Evaluation for Energy Grids
- Data Analysis and Deep Learning
  - Theoretical limits for Deep Learning architectures
  - Applying tensor algebra to machine learning problems (\*)
  - Machine learning for signal processing and communication applications





#### **Outline**

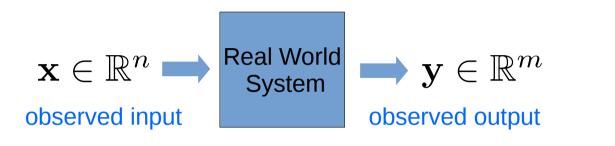
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#### Introduction to Neural Networks

What the world gives us



$$\mathbf{y} = f(\mathbf{x})$$

- How do we approximate  $f(\cdot)$  ?
  - Naive approach: approximate it with a polynomial and "train" the polynomial weights. Example for n=5 and m=1

• e.g. 
$$\hat{y} = w_1(x_1)^3 + w_2(x_1)^2 x_2 + w_3(x_1)^2 x_3 + \dots + w_N(x_5)^3$$

estimation

these weights are "trained" by minimizing

$$\min_{\mathbf{w}} \sum_{\mathbf{x} \in \mathcal{S}} (y - \hat{y})^2$$

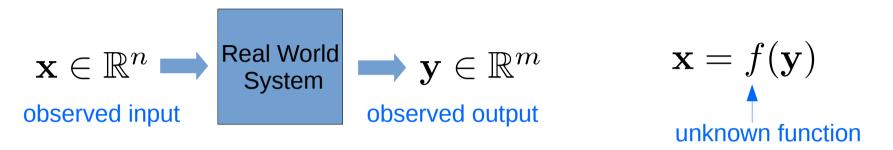
large set of observations





#### Introduction to Neural Networks

What the world gives us



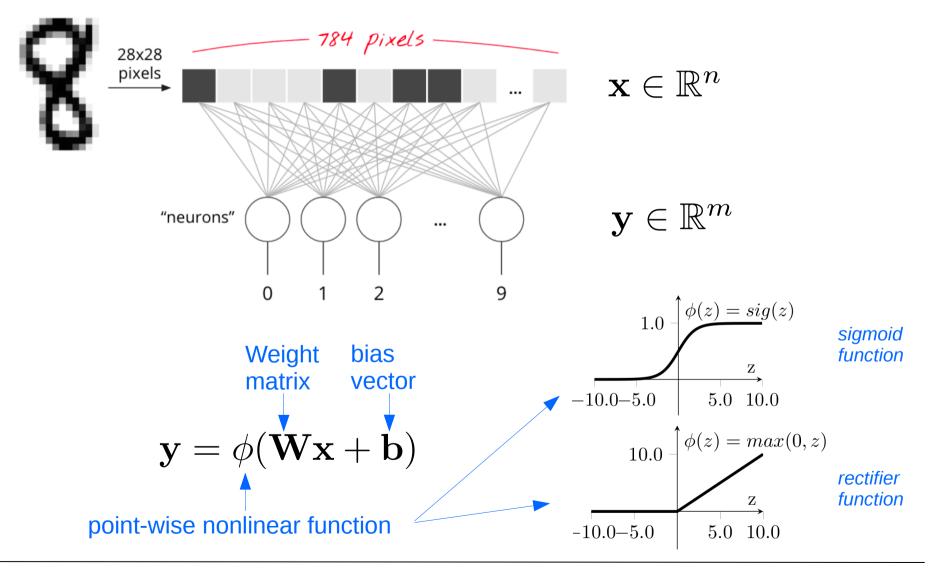
- How do we approximate  $f(\cdot)$  ?
  - Better approach: Use a "neural network" composed of linear and non-linear functions and train its parameters
    - a neural network with finite amount of parameters can approximate a wide variety of functions, see the universal approximation theorem
    - cost functions as  $\min_{\mathbf{w}} \sum_{\mathbf{x} \in \mathcal{S}} \|\mathbf{y} \hat{\mathbf{y}}\|_2^2$  are also used here





#### Introduction to Neural Networks

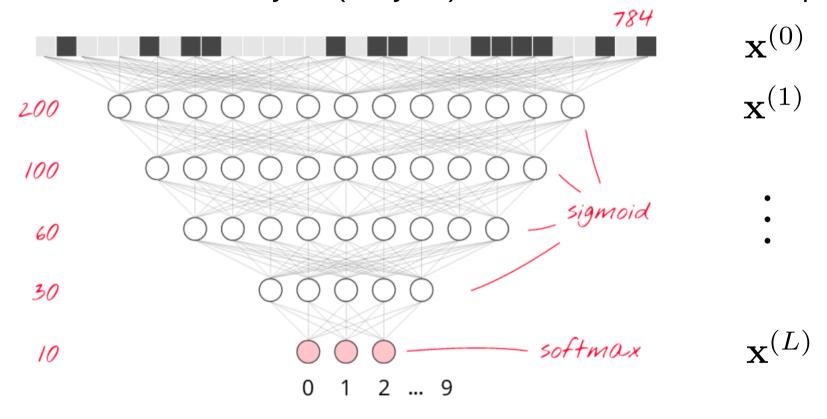
Lets start with a neural network of 1 layer





#### Introduction to Neural Networks

Now lets add more layers (L layers) to make this network "deep"



$$\mathbf{x}^{(l)} = \phi^{(l)} \left( \mathbf{W}^{(l)} \mathbf{x}^{(l-1)} + \mathbf{b}^{(l)} \right) \quad \forall l = 1, 2, \dots, L$$



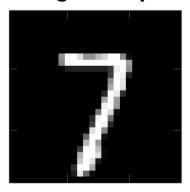


## **Malicious Input Detection**

 Malicious inputs are intentionally designed to "fool" a given neural network. These inputs can be "adversarial" or "rubbish" examples.

#### **Adversarial Input**

**Original Input** 



Prediction: 7 Confidence: 99.9%

"small" perturbation



Prediction: 2 Confidence: 89.6%







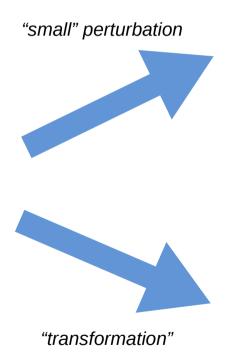
## **Malicious Input Detection**

 Malicious inputs are intentionally designed to "fool" a given neural network. These inputs can be "adversarial" or "rubbish" examples.

#### **Adversarial Input**

# Original Input

Prediction: 7 Confidence: 99.9%



7

Prediction: 2 Confidence: 89.6%

#### **Rubbish Input**



Prediction: 3

Confidence: 99.9%





#### **Malicious Input Detection**

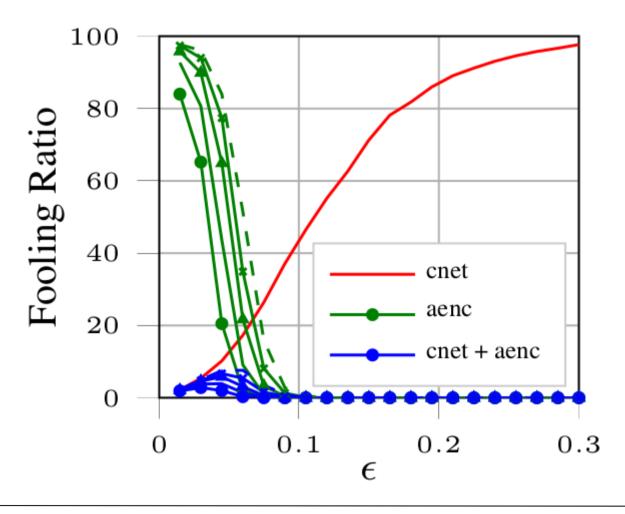
- Traditional approaches
  - include adversarial inputs during the training phase
  - denoise the input, using an autoencoder, before feeding it to the classifier
- All traditional approaches require adversarial and noisy data during training
- Proposed approach:
  - Use a hypothesis testing setup to detect malicious inputs
  - We make use of the hypothesis testing tools for designing an "optimal" test for detection





#### Results

- On the MNIST dataset, we are able to detect 100% of the rubbish inputs
- The detection ratio of adversarial inputs depends on the allowed perturbation norm (controlled by the parameter  $\epsilon$  )



- Larger perturbations are easier to detect by the autoencoder (aenc)
- Smaller perturbations are less likely to fool the classifier network (cnet)





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#### **Personal Advice**

- If you intend to pursue a Ph.D. after your master studies, try to get 1
  publication before graduating
  - the BRP, ARP, and master thesis are perfect for this purpose
  - is not common for Ph.D. applicants to already have some publications
  - you will have more opportunities to select from
  - is better to spend 2 extra months for publishing a paper if you later save 3 or more months of Ph.D. search





# Thank you



