



Artificial Intelligence

in Stock Exchanges and Financial Markets

» Modality: online

» Duration: 12 months.

» Certificate: TECH Global University

» Accreditation: 90 ECTS

» Schedule: at your own pace

» Exams: online

» Target Group: University graduates who have previously completed any degree in the fields of Social and Legal Sciences, Administrative and Business Sciences, or Design and Artificial Intelligence

Website: www.techtitute.com/us/school-of-business/executive-master-degree/master-artificial-intelligence-stock-exchanges-financial-markets

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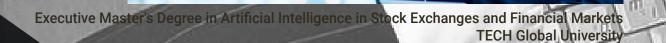
Why Study at TECH? Why Our Program? Objectives Welcome p. 4 p. 6 p. 10 p. 14 06 05 Methodology Skills Structure and Content p. 26 p. 48 p. 20 80 **Course Management** Our Students' Profiles Impact on Your Career p. 56 p. 60 p. 64 Benefits for Your Company Certificate

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01 **Welcome**

Artificial Intelligence (AI)-based tools, such as machine learning algorithms and natural language processing, are optimizing investment decision-making by analyzing large volumes of data in real time. For example, algorithmic trading systems employing AI can identify patterns and perform trades with much greater speed and accuracy than humans, which can generate significant profits and minimize risks. In this context, TECH offers an innovative university program designed to equip graduates with the necessary tools to effectively lead the digital transformation of their companies. In addition, it is conducted entirely online, allowing them to organize their schedules autonomously, with the support of the Relearning methodology.









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At TECH Global University



Innovation

The university offers an online learning model that balances the latest educational technology with the most rigorous teaching methods. A unique method with the highest international recognition that will provide students with the keys to develop in a rapidly-evolving world, where innovation must be every entrepreneur's focus.

"Microsoft Europe Success Story", for integrating the innovative, interactive multi-video system.



The Highest Standards

Admissions criteria at TECH are not economic. Students don't need to make a large investment to study at this university. However, in order to obtain a qualification from TECH, the student's intelligence and ability will be tested to their limits. The institution's academic standards are exceptionally high...

95%

of TECH students successfully complete their studies



Networking

Professionals from countries all over the world attend TECH, allowing students to establish a large network of contacts that may prove useful to them in the future.

+100000

+200

executives prepared each year

different nationalities



Empowerment

Students will grow hand in hand with the best companies and highly regarded and influential professionals. TECH has developed strategic partnerships and a valuable network of contacts with major economic players in 7 continents.

+500

collaborative agreements with leading companies



Talent

This program is a unique initiative to allow students to showcase their talent in the business world. An opportunity that will allow them to voice their concerns and share their business vision.

After completing this program, TECH helps students show the world their talent.



Multicultural Context

While studying at TECH, students will enjoy a unique experience. Study in a multicultural context. In a program with a global vision, through which students can learn about the operating methods in different parts of the world, and gather the latest information that best adapts to their business idea.

TECH students represent more than 200 different nationalities.



TECH strives for excellence and, to this end, boasts a series of characteristics that make this university unique:



Analysis

TECH explores the student's critical side, their ability to question things, their problem-solving skills, as well as their interpersonal skills.



Academic Excellence

TECH offers students the best online learning methodology. The university combines the Relearning method (postgraduate learning methodology with the best international valuation) with the Case Study. Tradition and vanguard in a difficult balance, and in the context of the most demanding educational itinerary.



Economy of Scale

TECH is the world's largest online university. It currently boasts a portfolio of more than 10,000 university postgraduate programs. And in today's new economy, **volume + technology = a ground-breaking price**. This way, TECH ensures that studying is not as expensive for students as it would be at another university.





Learn with the best

In the classroom, TECH's teaching staff discuss how they have achieved success in their companies, working in a real, lively, and dynamic context. Teachers who are fully committed to offering a quality specialization that will allow students to advance in their career and stand out in the business world.

Teachers representing 20 different nationalities.



At TECH, you will have access to the most rigorous and up-to-date case analyses in academia"



professional success in senior business management.

It is a challenge that demands effort and dedication, but it opens the door to a promising future. Students will learn from the best teaching staff and with the most flexible and innovative educational methodology.



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This program will provide you with a multitude of professional and personal advantages, among which we highlight the following:



A Strong Boost to Your Career

By studying at TECH, students will be able to take control of their future and develop their full potential. By completing this program, students will acquire the skills required to make a positive change in their career in a short period of time.

70% of students achieve positive career development in less than 2 years.



Develop a strategic and global vision of the company

TECH offers an in-depth overview of general management to understand how each decision affects each of the company's different functional fields.

Our global vision of companies will improve your strategic vision.



Consolidate the student's senior management skills

Studying at TECH means opening the doors to a wide range of professional opportunities for students to position themselves as senior executives, with a broad vision of the international environment.

You will work on more than 100 real senior management cases.



You will take on new responsibilities

The program will cover the latest trends, advances and strategies, so that students can carry out their professional work in a changing environment.

45% of graduates are promoted internally.



Access to a powerful network of contacts

TECH connects its students to maximize opportunities. Students with the same concerns and desire to grow. Therefore, partnerships, customers or suppliers can be shared.

You will find a network of contacts that will be instrumental for professional development.



Thoroughly develop business projects.

Students will acquire a deep strategic vision that will help them develop their own project, taking into account the different fields in companies.

20% of our students develop their own business idea.



Improve soft skills and management skills

TECH helps students apply and develop the knowledge they have acquired, while improving their interpersonal skills in order to become leaders who make a difference.

Improve your communication and leadership skills and enhance your career.



You will be part of an exclusive community

Students will be part of a community of elite executives, large companies, renowned institutions, and qualified teachers from the most prestigious universities in the world: the TECH Global University community.

We give you the opportunity to study with a team of world-renowned teachers.





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TECH makes the goals of their students their own goals too Working together to achieve them

The Executive Master's Degree in Artificial Intelligence in Stock Exchanges and Financial Markets will enable students to:



Understand the historical evolution and key developments in Artificial Intelligence (AI)



Analyze and apply data collection, integration, and storage techniques, including data warehouses



Study the principles and applications of neural networks, genetic algorithms, and data processing techniques in Al





Understand the fundamental concepts of statistics, data classification and the data life cycle



Develop skills in data science, transformation of data into information, and efficient management of datasets



Apply techniques of statistical inference, exploratory analysis and data pre-processing for data mining



Explore concepts of intelligent agents, knowledge representation, and the semantic web in intelligent systems



09

Introduce and apply machine learning techniques, including decision trees, neural networks, Bayesian methods, and clustering techniques



Study and apply various optimization algorithms and techniques, including sorting algorithms, trees, graphs, and backtracking techniques



Understand and apply the fundamentals of Deep Learning and convolutional neural networks to Deep Learning



Develop and optimize deep neural networks, using techniques such as Transfer Learning and Data Augmentation



Develop skills in Natural Language Processing (NLP) with RNNs, attention mechanisms and Transformers models



Customize and train Deep Learning models with TensorFlow, using advanced tools and techniques of the platform





Apply convolutional neural networks in Deep Computer Vision, including the use of pre-trained models and object classification and detection techniques



Explore Autoencoders, GANs and Diffusion Models for data representation and generation



Introduce and apply concepts of bio-inspired computation and evolutionary models for optimization



Optimize technical and fundamental analysis of financial markets using AI, including algorithmic trading techniques and performance analysis





Master Big Data technologies for large-scale, real-time processing of financial data, ensuring security and privacy



Develop AI strategies in financial services and other industries, considering specific risks and applications



Explore ethical and regulatory aspects of Al in finance, promoting responsible practices and regulatory compliance



Skills Entrepreneurs will acquire skills in technical and fundamental analysis of markets using Artificial Intelligence, allowing them to optimize trading strategies and make more informed decisions. They will also be trained in the implementation and management of algorithmic trading systems, analysis of large volumes of data with Big Data tools, and application of Machine Learning and natural language processing techniques. In addition, professionals will be able to address ethical and regulatory challenges associated with AI, ensuring solutions implemented are both innovative and compliant with industry regulations.





Apply AI techniques and strategies to improve efficiency in the retail sector



Effectively create training data sets for natural language processing (NLP) tasks



Delve into understanding and application of genetic algorithms

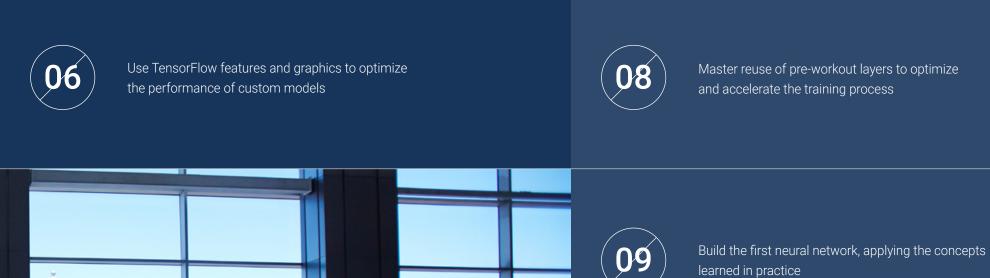




Implement noise removal techniques using automatic encoders



Run grouping layers and their use in Deep Computer Vision models with Keras





Optimize the development and application of chatbots and virtual assistants, understanding their operation and potential applications



Activate Multilayer Perceptron (MLP) using the Keras library



Apply data scanning and preprocessing techniques, identifying and preparing data for effective use in machine learning models



Develop data cleaning techniques to ensure the quality and accuracy of the information used in subsequent analyses



Implement effective strategies for handling missing values in datasets, applying imputation or elimination methods according to context





Investigate languages and software for the creation of ontologies, using specific tools for the development of semantic models



Analyze financial statements with Natural Language Processing (NLP) to extract valuable insights and perform accurate valuations of companies





Apply Explainable Artificial Intelligence (XAI) techniques to ensure transparency and understanding of models used in finance



Comply with ethical and regulatory standards in the implementation of AI in the financial sector, ensuring responsible and legally compliant practices



Develop and evaluate High Frequency Trading (HFT) strategies, optimizing speed and accuracy in order execution



Visualize financial data in an advanced way with tools, such as Plotly and Dash, facilitating informed decision making





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Syllabus

The curriculum will provide comprehensive training in technical and fundamental analysis of financial markets, using Artificial Intelligence to enhance indicator visualization, pattern recognition and trading automation. In this way, entrepreneurs will be able to implement advanced techniques, such as convolutional neural networks, to identify investment opportunities and use Reinforcement Learning to develop effective algorithmic trading strategies.

Crucial aspects of fundamental analysis and large-scale financial data processing will also be covered, using Big Data tools, such as Hadoop and Spark, to handle large volumes of information efficiently and securely. Machine Learning and NLP techniques for modeling financial performance, fraud detection and accurate valuations will also be examined. At the same time, it will focus on the design of algorithmic trading strategies and associated risk management.

In this way, TECH has developed a complete university program in a completely online format, which allows graduates to access educational materials from any device with an Internet connection. This eliminates the need to move to a physical location and adhere to fixed schedules. Additionally, it employs the revolutionary Relearning methodology, which focuses on the repetition of fundamental concepts to ensure a deep understanding of the content.

This Executive Master's Degree takes place over 12 months and is divided into 20 modules:

Module 1	Fundamentals of Artificial Intelligence
Module 2	Data Types and Data Life Cycle
Module 3	Data in Artificial Intelligence
Module 4	Data Mining: Selection, Pre-Processing and Transformation
Module 5	Algorithm and Complexity in Artificial Intelligence
Module 6	Intelligent Systems
Module 7	Machine Learning and Data Mining
Module 8	Neural Networks, the Basis of Deep Learning
Module 9	Deep Neural Networks Training
Module 10	Model Customization and Training with TensorFlow

Module 11	Deep Computer Vision with Convolutional Neural Networks
Module 12	Natural Language Processing (NLP) with Recurrent Neural Networks (RNN) and Attention
Module 13	Autoencoders, GANs and Diffusion Models
Module 14	Bio-Inspired Computing
Module 15	Artificial Intelligence: Strategies and Applications
Module 16	Technical Analysis of Financial Markets with Al
Module 17	Fundamental Analysis of Financial Markets with Al
Module 18	Large Scale Financial Data Processing
Module 19	Algorithmic Trading Strategies
Module 20	Ethical and Regulatory Aspects of Al in Finance

Where, When and How is it Taught?

TECH offers the possibility to develop this Executive Master's Degree in Artificial Intelligence in Stock Exchanges and Financial Markets completely online. Throughout the 12 months of the educational program, the students will be able to access all the contents of this program at any time, allowing them to self-manage their study time.

A unique, key, and decisive educational experience to boost your professional development and make the definitive leap.

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Mod	ule 1. Fundamentals of Artificial Intellig	ence				
1.1. 1.1.1. 1.1.2. 1.1.3. 1.1.4.	History of Artificial Intelligence When Do We Start Talking About Artificial Intelligence? References in Film Importance of Artificial Intelligence Technologies that Enable and Support Artificial Intelligence	1.2. 1.2.1. 1.2.2. 1.2.3.	Artificial Intelligence in Games Game Theory Minimax and Alpha-Beta Pruning Simulation: Monte Carlo	1.3. 1.3.1. 1.3.2. 1.3.3. 1.3.4. 1.3.5.	Neural Networks Biological Fundamentals Computational Model Supervised and Unsupervised Neural Networks Simple Perceptron Multilayer Perceptron	Genetic Algorithms History Biological Basis Problem Coding Generation of the Initial Population Main Algorithm and Genetic Operators Evaluation of Individuals: Fitness
1.5. 1.5.1. 1.5.2. 1.5.3. 1.5.4. 1.5.5.	Thesauri, Vocabularies, Taxonomies Vocabulary Taxonomy Thesauri Ontologies Knowledge Representation: Semantic Web	1.6. 1.6.1. 1.6.2. 1.6.3.	Semantic Web Specifications RDF, RDFS and OWL Inference/ Reasoning Linked Data	1.7. 1.7.1. 1.7.2.	Expert Systems and DSS Expert Systems Decision Support Systems	Chatbots and Virtual Assistants Types of Assistants: Voice and Text Assistants Fundamental Parts for the Development of an Assistant: Intents, Entities and Dialog Flow Integrations: Web, Slack, Whatsapp, Facebook Assistant Development Tools: Dialog Flow, Watson Assistant
1.9.	Al Implementation Strategy	1.10.1. 1.10.2 1.10.3	Future of Artificial Intelligence Understand How to Detect Emotions Using Algorithms Creating a Personality: Language, Expressions and Content Trends of Artificial Intelligence Reflections			

Module 2. Data Types and Data Life Cycle			
 2.1. Statistics 2.1.1. Statistics: Descriptive Statistics, Statistical Inferences 2.1.2. Population, Sample, Individual 2.1.3. Variables: Definition, Measurement Scales 	 2.2. Types of Data Statistics 2.2.1. According to Type 2.2.1.1. Quantitative: Continuous Data and Discrete Data 2.2.1.2. Qualitative: Binomial Data, Nominal Data and Ordinal Data 	 2.2.2. According to Its Shape 2.2.2.1. Numeric 2.2.2.2. Text: 2.2.2.3. Logical 2.2.3. According to Its Source 2.2.3.1. Primary 2.2.3.2. Secondary 	2.3. Life Cycle of Data2.3.1. Stages of the Cycle2.3.2. Milestones of the Cycle2.3.2. FAIR Principles
 2.4. Initial Stages of the Cycle 2.4.1. Definition of Goals 2.4.2. Determination of Resource Requirements 2.4.3. Gantt Chart 2.4.4. Data Structure 	2.5. Data Collection2.5.1. Methodology of Data Collection2.5.2. Data Collection Tools2.5.3. Data Collection Channels	2.6. Data Cleaning2.6.1. Phases of Data Cleansing2.6.2. Data Quality2.6.3. Data Manipulation (with R)	 2.7. Data Analysis, Interpretation and Result Evaluation 2.7.1. Statistical Measures 2.7.2. Relationship Indexes 2.7.3. Data Mining
2.8. Datawarehouse2.8.1. Elements that Comprise It2.8.2. Design2.8.3. Aspects to Consider	2.9. Data Availability2.9.1. Access2.9.2. Uses2.9.3. Security	2.10. Regulatory Framework 2.10.1. Data Protection Law 2.10.2. Good Practices 2.10.3. Other Regulatory Aspects	
Module 3. Data in Artificial Intelligence			
3.1. Data Science3.1.1. Data Science3.1.2. Advanced Tools for Data Scientists	3.2. Data, Information and Knowledge3.2.1. Data, Information and Knowledge3.2.2. Types of Data3.2.3. Data Sources	3.3. From Data to Information3.3.1. Data Analysis3.3.2. Types of Analysis3.3.3. Extraction of Information from a Dataset	 3.4. Extraction of Information Through Visualization 3.4.1. Visualization as an Analysis Tool 3.4.2. Visualization Methods 3.4.3. Visualization of a Data Set
3.5. Data Quality3.5.1. Quality Data3.5.2. Data Cleaning3.5.3. Basic Data Pre-Processing	3.6. Dataset3.6.1. Dataset Enrichment3.6.2. The Curse of Dimensionality3.6.3. Modification of Our Data Set	3.7. Unbalance3.7.1. Classes of Unbalance3.7.2. Unbalance Mitigation Techniques3.7.3. Balancing a Dataset	3.8. Unsupervised Models3.8.1. Unsupervised Model3.8.2. Methods3.8.3. Classification with Unsupervised Models
3.9. Supervised Models	3.10. Tools and Good Practices		

3.10.1. Good Practices for Data Scientists

3.10.2. The Best Model

3.10.3. Useful Tools

3.9.1. Supervised Model

3.9.3. Classification with Supervised Models

3.9.2. Methods

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Mod	ule 4. Data Mining: Selection, Pre-Proce	ssing a	and Transformation				
4.1.1. 4.1.2.	Statistical Inference Descriptive Statistics vs. Statistical Inference Parametric Procedures Non-Parametric Procedures	4.2.1. 4.2.2.	Exploratory Analysis Descriptive Analysis Visualization Data Preparation	4.3.1. 4.3.2.	Data Preparation Integration and Data Cleaning Normalization of Data Transforming Attributes	4.4.1. 4.4.2.	Missing Values Treatment of Missing Values Maximum Likelihood Imputation Methods Missing Value Imputation Using Machine Learni
4.5.1. 4.5.2. 4.5.3. 4.9. 4.9.1. 4.9.2.	Noise in the Data Noise Classes and Attributes Noise Filtering The Effect of Noise Instance Selection Methods for Instance Selection Prototype Selection Advanced Methods for Instance Selection	4.6.1. 4.6.2. 4.6.3.	The Curse of Dimensionality Oversampling Undersampling Multidimensional Data Reduction Data Pre-processing in Big Data Environments	4.7.1.	From Continuous to Discrete Attributes Continuous Data Vs. Discreet Data Discretization Process	4.8.1. 4.8.2.	The Data Data Selection Prospects and Selection Criteria Selection Methods
	ule 5. Algorithm and Complexity in Artif		telligence Efficiency and Analysis of Algorithms	526	Mathematical Analysis Criteria for Non-Recursive	5.3	Sorting Algorithms
5.1.1. 5.1.2.	Strategies Recursion Divide and Conquer Other Strategies	5.2.1. 5.2.2. 5.2.3. 5.2.4.	Efficiency Measures Measuring the Size of the Input Measuring Execution Time Worst, Best and Average Case Asymptotic Notation	5.2.7.	Algorithms Mathematical Analysis of Recursive Algorithms Empirical Analysis of Algorithms	5.3.1. 5.3.2. 5.3.3. 5.3.4. 5.3.5.	Concept of Sorting Bubble Sorting Sorting by Selection Sorting by Insertion Merge Sort Quick Sort
5.4.1. 5.4.2. 5.4.3. 5.4.4. 5.4.5.	Algorithms with Trees Tree Concept Binary Trees Tree Paths Representing Expressions Ordered Binary Trees Balanced Binary Trees	5.5.1. 5.5.2.	Algorithms Using Heaps Heaps The Heapsort Algorithm Priority Queues	5.6.1. 5.6.2. 5.6.3.	Graph Algorithms Representation Traversal in Width Depth Travel Topological Sorting	5.7.1. 5.7.2. 5.7.3. 5.7.4.	Greedy Algorithms Greedy Strategy Elements of the Greedy Strategy Currency Exchange Traveler's Problem Backpack Problem
5.8.1. 5.8.2.	Minimal Path Finding The Minimum Path Problem Negative Arcs and Cycles Dijkstra's Algorithm	5.9.1. 5.9.2. 5.9.3.	Greedy Algorithms on Graphs The Minimum Covering Tree Prim's Algorithm Kruskal's Algorithm Complexity Analysis	5.10.1	Backtracking Backtracking Alternative Techniques		

Mod	lule 6. Intelligent Systems					
6.1. 6.1.1. 6.1.2. 6.1.3. 6.1.4.	3	6.2. 6.2.1. 6.2.2. 6.2.3. 6.2.4. 6.2.5.	Agent Architectures The Reasoning Process of an Agent Reactive Agents Deductive Agents Hybrid Agents Comparison	 6.3. Information and Knowledge 6.3.1. Difference between Data, Information and Knowledge 6.3.2. Data Quality Assessment 6.3.3. Data Collection Methods 6.3.4. Information Acquisition Methods 6.3.5. Knowledge Acquisition Methods 	6.4. 6.4.1. 6.4.2. 6.4.3.	Definition of Knowledge Representation According to Roles
6.5. 6.5.1. 6.5.2. 6.5.3. 6.5.4. 6.5.5.	Ontologies Introduction to Metadata Philosophical Concept of Ontology Computing Concept of Ontology Domain Ontologies and Higher-Level Ontologies How to Build an Ontology	6.6.3.	Ontology Languages and Ontology Creation Software Triple RDF, Turtle and N RDF Schema OWL SPARQL	6.6.5. Introduction to Ontology Creation Tools 6.6.6. Installing and Using Protégé	6.7. 6.7.1. 6.7.2.	Semantic Web Current and Future Status of the Semantic Web Semantic Web Applications
6.8.1. 6.8.2. 6.8.3. 6.8.4. 6.8.5. 6.8.6. 6.8.7.	Global Visíon Taxonomy Thesauri Folksonomy	6.9.1. 6.9.2. 6.9.3. 6.9.4. 6.9.5.	Knowledge Representation Assessment and Integration Zero-Order Logic First-Order Logic Descriptive Logic Relationship between Different Types of Logic Prolog: Programming Based on First-Order Logic	6.10. Semantic Reasoners, Knowledge-Based Systems and Expert Systems 6.10.1. Concept of Reasoner 6.10.2. Reasoner Applications 6.10.3. Knowledge-Based Systems 6.10.4. MYCIN: History of Expert Systems 6.10.5. Expert Systems Elements and Architecture 6.10.6. Creating Expert Systems		

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Module 7. Machine Learning and Data Mini	ng		
 7.1. Introduction to Knowledge Discovery Processes and Basic Concepts of Machine Learning 7.1.1. Key Concepts of Knowledge Discovery Processes 7.1.2. Historical Perspective of Knowledge Discovery Processes 7.1.3. Stages of the Knowledge Discovery Processes 7.1.4. Techniques Used in Knowledge Discovery Processes 7.1.5. Characteristics of Good Machine Learning Models 	7.1.6. Types of Machine Learning Information7.1.7. Basic Learning Concepts7.1.8. Basic Concepts of Unsupervised Learning	 7.2. Data Exploration and Pre-Processing 7.2.1. Data Processing 7.2.2. Data Processing in the Data Analysis Flow 7.2.3. Types of Data 7.2.4. Data Transformations 7.2.5. Visualization and Exploration of Continuous Variables 7.2.6. Visualization and Exploration of Categorical Variables 7.2.7. Correlation Measures 	7.2.8. Most Common Graphic Representations 7.2.9. Introduction to Multivariate Analysis and Dimensionality Reduction
7.3. Decision Trees7.3.1. ID Algorithm7.3.2. Algorithm C7.3.3. Overtraining and Pruning7.3.4. Result Analysis	 7.4. Evaluation of Classifiers 7.4.1. Confusion Matrixes 7.4.2. Numerical Evaluation Matrixes 7.4.3. Kappa Statistic 7.4.4. ROC Curves 	 7.5. Classification Rules 7.5.1. Rule Evaluation Measures 7.5.2. Introduction to Graphic Representation 7.5.3. Sequential Overlay Algorithm 	 7.6. Neural Networks 7.6.1. Basic Concepts 7.6.2. Simple Neural Networks 7.6.3. Backpropagation Algorithm 7.6.4. Introduction to Recurrent Neural Networks
7.7. Bayesian Methods 7.7.1. Basic Probability Concepts 7.7.2. Bayes' Theorem 7.7.3. Naive Bayes 7.7.4. Introduction to Bayesian Networks	 7.8. Regression and Continuous Response Models 7.8.1. Simple Linear Regression 7.8.2. Multiple Linear Regression 7.8.3. Logistic Regression 7.8.4. Regression Trees 7.8.5. Introduction to Support Vector Machines (SVM) 7.8.6. Goodness-of-Fit Measures 	7.9. Clustering 7.9.1. Basic Concepts 7.9.2. Hierarchical Clustering 7.9.3. Probabilistic Methods 7.9.4. EM Algorithm 7.9.5. B-Cubed Method 7.9.6. Implicit Methods	7.10 Text Mining and Natural Language Processing (NLP) 7.10.1. Basic Concepts 7.10.2. Corpus Creation 7.10.3. Descriptive Analysis 7.10.4. Introduction to Feelings Analysis

8.1. 8.1.1. 8.1.2. 8.1.3.	Deep Learning Types of Deep Learning Applications of Deep Learning Advantages and Disadvantages of Deep Learning	8.2. 8.2.1. 8.2.2. 8.2.3.	Surgery Sum Product Transfer	8.3.1. 8.3.2.	Layers Input Layer Hidden Layer Output Layer	8.4. 8.4.1. 8.4.2. 8.4.3.	Architecture Design Connection between Layers
8.5.1. 8.5.2. 8.5.3.	Construction of the First Neural Network Network Design Establish the Weights Network Training	8.6. 8.6.1. 8.6.2. 8.6.3.	Trainer and Optimizer Optimizer Selection Establishment of a Loss Function Establishing a Metric	8.7.1 8.7.2 8.7.3.	Application of the Principles of Neural Networks Activation Functions Backward Propagation Parameter Adjustment	8.8 F 8.8.1. 8.8.2. 8.8.3.	3 3
8.9.	Implementation of MLP (Multilayer Perceptron) with Keras	8.10.	Fine Tuning Hyperparameters of Neural Networks				
8.9.1. 8.9.2. 8.9.3.	Definition of the Network Structure Model Compilation	8.10.2	. Selection of the Activation Function . Set the Learning Rate . Adjustment of Weights				
Mod	lule 9. Deep Neural Networks Training						
9.1. 9.1.1. 9.1.2. 9.1.3.	Gradient Problems Gradient Optimization Techniques Stochastic Gradients Weight Initialization Techniques	9.2. 9.2.1. 9.2.2. 9.2.3.	Reuse of Pre-Trained Layers Transfer Learning Training Feature Extraction Deep Learning	9.3. 9.3.1. 9.3.2. 9.3.3.		9.4. 9.4.1. 9.4.2. 9.4.3.	9

9.5. Overfitting

9.5.1. Cross Validation 9.5.2. Regularization

9.5.3. Evaluation Metrics

9.9. Practical Application of Transfer Learning

9.9.1. Transfer Learning Training

9.9.2. Feature Extraction 9.9.3. Deep Learning

9.6. Practical Guidelines

9.6.1. Model Design

9.6.2. Selection of Metrics and Evaluation Parameters 9.7.2. Feature Extraction

9.6.3. Hypothesis Testing

9.7. Transfer Learning

9.7.1. Transfer Learning Training

9.7.3. Deep Learning

9.8. Data Augmentation

9.8.1. Image Transformations

9.8.2. Synthetic Data Generation

9.8.3. Text Transformation

9.10. Regularization

9.10.1. L and L

9.10.2. Regularization by Maximum Entropy

9.10.3. Dropout

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Module 10. Model Customization and Training with TensorFlow 10.1. TensorFlow 10.2. TensorFlow and NumPy 10.3. Model Customization and Training 10.4. TensorFlow Features and Graphs **Algorithms** 10.2.1. NumPv Computing Environment for TensorFlow 10.4.1. Functions with TensorFlow 10.1.1. Use of the TensorFlow Library 10.1.2. Model Training with TensorFlow 10.2.2. Using NumPy Arrays with TensorFlow 10.4.2. Use of Graphs for Model Training 10.3.1. Building Custom Models with TensorFlow 10.1.3. Operations with Graphs in TensorFlow 10.2.3. NumPy Operations for TensorFlow Graphs 10.4.3. Grap Optimization with TensorFlow Operations 10.3.2. Management of Training Parameters 10.3.3. Use of Optimization Techniques for Training 10.5. Loading and Preprocessing Data 10.6. The API tfdata 10.8. Keras Preprocessing Layers 10.7. The TFRecord Format with TensorFlow 10.6.1. Using the tfdataAPI for Data Processing 10.7.1. Using the TFRecord API for Data Serialization 10.8.1. Using the Keras Preprocessing API 10.6.2. Construction of Data Streams with tfdata 10.7.2. Loading TFRecord Files with TensorFlow 10.8.2. Construction of Pre-Processing Pipelined 10.5.1. Loading Data Sets with TensorFlow 10.6.3. Using the tfdata API for Model Training 10.7.3. Using TFRecord Files for Training Models with Keras 10.5.2. Pre-Processing Data with TensorFlow 10.8.3. Using the Keras Pre-processing API for 10.5.3. Using TensorFlow Tools for Data Manipulation Model Training 10.9. The TensorFlow Datasets Project 10.10. Building a Deep Learning application with TensorFlow 10.9.1. Using TensorFlow Datasets for Data Loading 10.9.2. Data Pre-Processing with TensorFlow Datasets 10.10.1. Practical Application 10.9.3. Using TensorFlow Datasets for Model Training 10.10.2. Building a Deep Learning Application with TensorFlow 10.10.3. Training a model with TensorFlow

10.10.4. Use of the Application for the Prediction of

Results

Module 11. Deep Computer Vision with Convolutional Neural Networks						
11.1. The Cortex Visual Architecture 11.1.1. Functions of the Visual Cortex 11.1.2. Theories of Computational Vision 11.1.3. Models of Image Processing	11.2. Convolutional Layers 11.2.1 Reuse of Weights in Convolution 11.2.2. Convolution D 11.2.3. Activation Functions	11.3. Grouping Layers and Implementation of Grouping Layers with Keras 11.3.1. Pooling and Striding 11.3.2. Flattening 11.3.3. Types of Pooling	11.4. CNN Architecture 11.4.1. VGG Architecture 11.4.2. AlexNet Architecture 11.4.3. ResNet Architecture			
11.5. Implementing a CNN ResNet - Using Keras 11.5.1. Weight Initialization 11.5.2. Input Layer Definition 11.5.3. Output Definition	11.6. Use of Pre-Trained Keras Models 11.6.1. Characteristics of Pre-Trained Models 11.6.2. Uses of Pre-Trained Models 11.6.3. Advantages of Pre-Trained Models	11.7. Pre-Trained Models for Transfer Learning 11.7.1. Transfer Learning 11.7.2. Transfer Learning Process 11.7.3. Advantages of Transfer Learning	 11.8. Classification and Localization in Deep Computer Vision 11.8.1. Image Classification 11.8.2. Localization of Objects in Images 11.8.3. Object Detection 			
11.9. Object Detection and Object Tracking 11.9.1. Object Detection Methods 11.9.2. Object Tracking Algorithms 11.9.3. Tracking and Localization Techniques	 11.10. Semantic Segmentation 11.10.1. Deep Learning for Semantic Segmentation 11.10.2. Edge Detection 11.10.3. Rule-Based Segmentation Methods 					

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Module 12. Natural Language Processing (NLP) with Recurrent Neural Networks (RNN) and Attention

12.1. Text Generation using RNN

- 12.1.1. Training an RNN for Text Generation
- 12.1.2. Natural Language Generation with RNN
- 12.1.3. Text Generation Applications with RNN

12.2. Training Data Set Creation

- 12.2.1. Preparation of the Data for Training an RNN
- 12.2.2. Storage of the Training Dataset
- 12.2.3. Data Cleaning and Transformation
- 12.2.4. Sentiment Analysis

12.3. Classification of Opinions with RNN

- 12.3.1. Detection of Themes in Comments
- 12.3.2. Sentiment Analysis with Deep Learning Algorithms

12.4. Encoder-Decoder Network for **Neural Machine Translation**

- 12.4.1. Training an RNN for Machine Translation
- 12.4.2. Use of an encoder-decoder network for machine translation
- 12.4.3. Improving the Accuracy of Machine Translation with RNNs

12.5. Attention Mechanisms

- 12.5.1. Application of Care Mechanisms in RNN
- 12.5.2. Use of Care Mechanisms to Improve the Accuracy of the Models
- 12.5.3. Advantages of Attention Mechanisms in Neural Networks

12.6. Transformer models

- 12.6.1. Using Transformers Models for Natural Language Processing
- 12.6.2. Application of Transformers Models for Vision
- 12.6.3. Advantages of Transformers Models

12.7. Transformers for vision

- 12.7.1. Use of Transformers Models for Vision
- 12.7.2. Image Data Preprocessing
- 12.7.3. Training a Transformers Model for Vision

12.8. Hugging Face Transformer Library

- 12.8.1. Using Hugging Face's Transformers Library
- 12.8.3. Advantages of Hugging Face's Transformers Library

12.9. Other Transformers Libraries Comparison

- 12.9.1. Comparison Between Different Transformers Libraries
- 12.9.2. Use of the other Transformers libraries
- 12.9.3. Advantages of the Other Transformers Libraries

12.10. Development of an NLP Application with RNN and Attention. Practical Application

- 12.10.1. Development of a Natural Language Processing Application with RNN and Attention
- 12.10.2. Use of RNN, Attention Mechanisms and Transformers Models in the Application
- 12.10.3. Evaluation of the Practical Application

- 12.8.2. Hugging Face's Transformers Library Application

13.1. Representation of Efficient Data 13.1.1. Dimensionality Reduction 13.1.2. Deep Learning 13.1.3. Compact Representations	 13.2. PCA Realization with an Incomplete Linear Automatic Encoder 13.2.1. Training Process 13.2.2. Implementation in Python 13.2.3. Use of Test Data 	13.3. Stacked Automatic Encoders13.3.1. Deep Neural Networks13.3.2. Construction of Coding Architectures13.3.3. Use of Regularization	13.4. Convolutional Autoencoders 13.4.1. Design of Convolutional Models 13.4.2. Convolutional Model Training 13.4.3. Results Evaluation
13.5. Noise Suppression of Automatic Encoders 13.5.1. Filter Application 13.5.2. Design of Coding Models 13.5.3. Use of Regularization Techniques	13.6. Sparse Automatic Encoders 13.6.1. Increasing Coding Efficiency 13.6.2. Minimizing the Number of Parameters 13.6.3. Using Regularization Techniques	13.7. Variational Automatic Encoders 13.7.1. Use of Variational Optimization 13.7.2. Unsupervised Deep Learning 13.7.3. Deep Latent Representations	13.8. Trendy MNIST Image Generation 13.8.1. Pattern Recognition 13.8.2. Image Generation 13.8.3. Deep Neural Networks Training
13.9. Generative Adversarial Networks and Diffusion Models 13.9.1. Content Generation from Images 13.9.2. Modeling of Data Distributions 13.9.3. Use of Adversarial Networks	13.10 Model Implementation 13.10.1. Practical Application 13.10.2. Implementation of the Models 13.10.3. Use of Real Data 13.10.4. Results Evaluation		

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Module 14. Bio-Inspired Computing			
14.1. Introduction to Bio-Inspired Computing 14.1.1. Introduction to Bio-Inspired Computing	14.2. Social Adaptation Algorithms 14.2.1. Bio-Inspired Computing Based on Ant Colonies 14.2.2. Variants of Ant Colony Algorithms 14.2.3. Particle Cloud Computing	14.3. Genetic Algorithms 14.3.1. General Structure 14.3.2. Implementations of the Major Operators	14.4. Space Exploration-Exploitation Strategies for Genetic Algorithms14.4.1. CHC Algorithm14.4.2. Multimodal Problems
14.5. Evolutionary Computation Models (I) 14.5.1. Evolutionary Strategies 14.5.2. Evolutionary Programming 14.5.3. Algorithms Based on Differential Evolution	14.6. Evolutionary Computation Models (II) 14.6.1. Evolutionary Models Based on Estimation of Distributions (EDA) 14.6.2. Genetic Programming	14.7. Evolutionary Programming Applied to Learning Problems14.7.1. Rules-Based Learning14.7.2. Evolutionary Methods in Instance Selection Problems	14.8. Multi-Objective Problems14.8.1. Concept of Dominance14.8.2. Application of Evolutionary Algorithms to Multi-Objective Problems
14.9. Neural Networks (I) 14.9.1. Introduction to Neural Networks 14.9.2. Practical Example with Neural Networks	 14.10. Neural Networks (II) 14.10.1. Use Cases of Neural Networks in Medical Research 14.10.2. Use Cases of Neural Networks in Economics 14.10.3. Use Cases of Neural Networks in Artificial Vision 		

Module 15. Artificial Intelligence: Strategies and Applications

15.1. Financial Services

- 15.1.1. The Implications of Artificial Intelligence (AI) in Financial Services Opportunities and Challenges
- 15.1.2. Case Uses
- 15.1.3. Potential Risks Related to the Use of Al
- 15.1.4. Potential Future Developments/Uses of Al

15.2. Implications of Artificial Intelligence in Healthcare Service

- 15.2.1. Implications of AI in the Healthcare Sector Opportunities and Challenges
- 15.2.2. Case Uses

15.3. Risks Related to the Use of AI in Healthcare Service

- 15.3.1. Potential Risks Related to the Use of Al
- 15.3.2. Potential Future Developments/Uses of Al

15.4. Retail

- 15.4.1. Implications of AI in Retail. Opportunities and Challenges
- 15.4.2. Case Uses
- 15.4.3. Potential Risks Related to the Use of Al
- 15.4.4. Potential Future Developments/Uses of Al

15.5. Industry

- 15.5.1. Implications of AI in Industry Opportunities and Challenges
- 15.5.2. Case Uses

15.6. Potential Risks Related to the Use of AI in Industry

- 15.6.1. Case Uses
- 15.6.2. Potential Risks Related to the Use of Al
- 15.6.3. Potential Future Developments/Uses of Al

15.7. Public Administration

- 15.7.1. Al Implications for Public Administration Opportunities and Challenges
- 15.7.2. Case Uses
- 15.7.3. Potential Risks Related to the Use of Al-
- 15.7.4. Potential Future Developments/Uses of Al

15.8. Educational

- 15.8.1. Al Implications for Education Opportunities and Challenges
- 15.8.2. Case Uses
- 15.8.3. Potential Risks Related to the Use of Al
- 15.8.4. Potential Future Developments/Uses of Al

15.9. Forestry and Agriculture

- 15.9.1. Implications of AI in Forestry and Agriculture.
 Opportunities and Challenges
- 15.9.2. Case Uses
- 15.9.3. Potential Risks Related to the Use of Al
- 15.9.4. Potential Future Developments/Uses of Al

15.10 Human Resources

- 15.10.1. Implications of AI for Human Resources Opportunities and Challenges
- 15.10.2. Case Uses
- 15.10.3. Potential Risks Related to the Use of Al
- 15.10.4. Potential Future Developments/Uses of Al

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Module 16. Technical Analysis of Financial Markets with Al 16.3. Financial Pattern Recognition 16.4. Quantitative Trading Strategies 16.1. Analysis and Visualization of 16.2. Optimization and Automation of Technical Indicators with Plotly Technical Indicators with Scikit-learn with CNN with QuantConnect and Dash 16.2.1. Automation of Indicators with Scikit-learn 16.4.1. Building Algorithmic Trading Systems with 16.3.1. Using CNN in TensorFlow to Identify Patterns 16.2.2. Optimization of Technical Indicators in Charts QuantConnect 16.1.1. Implementation of Interactive Charts with Plotly 16.2.3. Creating Personalized Indicators with Keras 16.3.2. Improving Recognition Models with Transfer 16.4.2. Backtesting Strategies with QuantConnect 16.1.2. Advanced Visualization of Time Series Learning Techniques 16.4.3. Integrating Machine Learning into Trading with Matplotlib 16.3.3. Validation of Recognition Models in Real-Strategies with OuantConnect 16.1.3. Creating Real-Time Dynamic Dashboards Time Markets with Dash 16.6. Time Series Modeling with LSTM in 16.5. Algorithmic Trading with Reinforcement 16.7. Application of Explainable Artificial 16.8. High-Frequency Trading (HFT) Optimized Intelligence (XAI) in Finance with Machine Learning Models Learning Using TensorFlow Keras for Price Forecasting 16.5.1. Reinforcement Learning for Trading 16.6.1. Applying LSTM to Price Forecasting 16.7.1. Applicability of XAI in Finances 16.8.1. Developing ML Models for HFT 16.5.2. Creating Trading Agents with TensorFlow 16.6.2. Implementing LSTM Models in Keras for 16.7.2. Applying LIME to Trading Models 16.8.2. Implementing HFT Strategies with TensorFlow Reinforcement Learning Financial Time Series 16.7.3. Using SHAP for Feature Contribution Analysis 16.8.3. Simulation and Evaluation of HFT in in Al Decisions 16.5.3. Simulating and Tuning Agents in OpenAl Gym 16.6.3. Optimization and Parameter Fitting in Time Controlled Environments Series Models 16.9. Volatility Analysis Using Machine Learning 16.10. Portfolio Optimization with Genetic Algorithms 16.9.1. Applying Intelligent Models to Predict Volatility 16.9.2. Implementing Volatility Models with PyTorch 16.10.1. Fundamentals of Genetic Algorithms for 16.9.3. Integrating Volatility Analysis into Portfolio Investment Optimization in Markets Risk Management 16.10.2. Implementing Genetic Algorithms for

Portfolio Selection

Evaluation of Portfolio Optimization Strategies

16.10.3.

Module 17. Fundamental Analysis of Financial Markets with Al

17.1. Predictive Financial Performance Modeling with Scikit-Learn

- 17.1.1. Linear and Logistic Regression for Financial Forecasting with Scikit-Learn
- 17.1.2. Using Neural Networks with TensorFlow to Forecast Revenues and Earnings
- 17.1.3. Validating Predictive Models with Cross-Validation Using Scikit-Learn

17.2. Valuation of Companies with Deep Learning

- 17.2.1. Automating the Discounted Cash Flows (DCF) Model with TensorFlow
- 17.2.2. Advanced Valuation Models Using PyTorch
- 17.2.3. Integration and Analysis of Multiple Valuation Models with Pandas

17.3. Analysis of Financial Statements with NLP Using ChatGPT

- 17.3.1. Extracting Key Information from Annual Reports with ChatGPT
- 17.3.2. Sentiment Analysis of Analyst Reports and Financial News with ChatGPT
- 17.3.3. Implementing NLP Models with Chat GPT for Interpreting Financial Texts

17.4. Risk and Credit Analysis with Machine Learning

- 17.4.1. Credit Scoring Models Using SVM and Decision Trees in Scikit-Learn
- 17.4.2. Credit Risk Analysis in Corporations and Bonds with TensorFlow
- 17.4.3. Visualization of Risk Data with Tableau

17.5. Credit Analysis with Scikit-Learn

- 17.5.1. Implementing Credit Scoring Models
- 17.5.2. Credit Risk Analysis with RandomForest in Scikit-Learn
- 17.5.3. Advanced Visualization of Credit Scoring Results with Tableau

17.6. ESG Sustainability Assessment with **Data Mining Techniques**

- 17.6.1. ESG Data Mining Methods
- 17.6.2. ESG Impact Modeling with Regression Techniques
- 17.6.3. Applications of ESG Analysis in Investment Decisions

17.7. Sector Benchmarking with Artificial Intelligence Using TensorFlow and Power BI

- 17.7.1. Comparative Analysis of Companies Using Al
- 17.7.2. Predictive Modeling of Sector Performance with TensorFlow
- 17.7.3. Implementing Industry Dashboards with Power BI

17.8. Portfolio Management with Al Optimization

- 17.8.1. Portfolio Optimization
- 17.8.3. Implementing and Evaluating the Effectiveness of Algorithms in Portfolio Management

17.9. Financial Fraud Detection with Al Using TensorFlow and Keras

- 17.9.1. Basic Concepts and Techniques of Fraud Detection with Al
- 17.9.2. Constructing Neural Network Detection Models in TensorFlow
- 17.9.3. Practical Implementation of Fraud Detection Systems in Financial Transactions

17.10. Analysis and Modeling in Mergers and Acquisitions with AI

- Using Predictive Al Models to Evaluate Mergers and Acquisitions
- 17.10.2. Simulating Post-Merger Scenarios Using Machine Learning Techniques
- 17.10.3. Evaluating the Financial Impact of M&A with Intelligent Models

- 17.8.2. Use of Machine Learning Techniques for Portfolio Optimization with Scikit-Optimize

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Module 18. Large Scale Financial Data Processing

18.1. Big Data in the Financial Context

- 18.1.1. Key Characteristics of Big Data in Finance
- 18.1.2. Importance of the 5 Vs (Volume, Velocity, Variety, Veracity, Value) in Financial Data
- 18.1.3. Use Cases of Big Data in Risk Analysis and Compliance

18.2. Technologies for Storage and Management of Financial Big Data

- 18.2.1. NoSOL Database Systems for Financial Warehousing
- 18.2.2. Using Data Warehouses and Data Lakes in the Financial Sector
- 18.2.3. Comparison between On-Premises and Cloud-Based Solutions

18.3. Real-Time Processing Tools for **Financial Data**

- 18.3.1. Introduction to Tools such as Apache Kafka and Apache Storm
- 18.3.2. Real-Time Processing Applications for Fraud Detection
- 18.3.3. Benefits of Real-Time Processing in Algorithmic Trading

18.4. Integration and Data Cleaning in Finance

- 18.4.1. Methods and Tools for Integrating Data from Multiple Sources
- 18.4.2. Data Cleaning Techniques to Ensure Data Quality and Accuracy
- 18.4.3. Challenges in the Standardization of Financial Data

18.5. Data Mining Techniques Applied to The Financial Markets

- 18.5.1. Classification and Prediction Algorithms in Market Data
- 18.5.2. Sentiment Analysis in Social Networks for Predicting Market Movements
- 18.5.3. Data Mining to Identify Trading Patterns and Investor Behavior

18.6. Advanced Data Visualization for **Financial Analysis**

- 18.6.1. Visualization Tools and Software for Financial Data
- 18.6.2. Design of Interactive Dashboards for Market Monitoring
- 18.6.3. The Role of Visualization in Risk Analysis Communication

18.7. Use of Hadoop and Related Ecosystems in Finance

- 18.7.1. Key Components of the Hadoop Ecosystem and Their Application in Finance
- 18.7.2. Hadoop Use Cases for Large Transaction Volume Analysis
- 18.7.3. Advantages and Challenges of Integrating Hadoop into Existing Financial Infrastructures

18.8. Spark Applications in Financial Analytics

- 18.8.1. Spark for Real-Time and Batch Data Analytics
- 18.8.3. Integration of Spark with Other Big Data Tools in Finance

18.9. Data Security and Privacy in the **Financial Sector**

- 18.9.1. Data Protection Rules and Regulations (GDPR, CCPA)
- 18.9.2. Encryption and Access Management Strategies for Sensitive Data
- 18.9.3. Impact of Data Breaches on Financial Institutions

18.10. Impact of Cloud Computing on Large-Scale Financial Analysis

- 18.10.1. Advantages of the Cloud for Scalability and Efficiency in Financial Analysis
- Comparison of Cloud Providers and Their 18.10.2. Specific Financial Services
- 18.10.3. Case Studies on Migration to the Cloud in Large Financial Institutions

- 18.8.2. Predictive Model Building Using Spark MLlib

Module 19. Algorithmic Trading Strategies			
 19.1. Fundamentals of Algorithmic Trading 19.1.1. Algorithmic Trading Strategies 19.1.2. Key Technologies and Platforms for the Development of Algorithmic Trading Algorithms 19.1.3. Advantages and Challenges of Automated Trading versus Manual Trading 	 19.2. Design of Automated Trading Systems 19.2.1. Structure and Components of an Automated Trading System 19.2.2. Algorithm Programming: from the Idea to the Implementation 19.2.3. Latency and Hardware Considerations in Trading Systems 	 19.3. Backtesting and Evaluation of Trading Strategies 19.3.1. Methodologies for Effective Backtesting of Algorithmic Strategies 19.3.2. Importance of Quality Historical Data in Backtesting 19.3.3. Key Performance Indicators for Evaluating Trading Strategies 	 19.4. Optimizing Strategies with Machine Learning 19.4.1. Applying Supervised Learning Techniques in Strategy Improvement 19.4.2. Using Particle Swarm Optimization and Genetic Algorithms 19.4.3. Challenges of Overfitting in Trading Strategy Optimization
19.5. High Frequency Trading (HFT) 19.5.1. Principles and Technologies behind HFT 19.5.2. Impact of HFT on Market Liquidity and Volatility 19.5.3. Common HFT Strategies and Their Effectiveness	 19.6. Order Execution Algorithms 19.6.1. Types of Execution Algorithms and Their Practical Application 19.6.2. Algorithms for Minimizing the Market Impact 19.6.3. Using Simulations to Improve Order Execution 	 19.7. Arbitration Strategies in Financial Markets 19.7.1. Statistical Arbitrage and Price Merger in Markets 19.7.2. Index and ETF Arbitrage 19.7.3. Technical and Legal Challenges of Arbitrage in Modern Trading 	 19.8. Risk Management in Algorithmic Trading 19.8.1. Risk Measures for Algorithmic Trading 19.8.2. Integrating Risk Limits and Stop-Loss in Algorithms 19.8.3. Specific Risks of Algorithmic Trading and How to Mitigate Them
 19.9. Regulatory Aspects and Compliance in Algorithmic Trading 19.9.1. Global Regulations Impacting Algorithmic Trading 19.9.2. Regulatory Compliance and Reporting in an Automated Environment 19.9.3. Ethical Implications of Automated Trading 	 19.10. Future of Algorithmic Trading and Emerging Trends 19.10.1. Impact of Artificial Intelligence on the Future Development of Algorithmic Trading 19.10.2. New Blockchain Technologies and Their Application in Algorithmic Trading 19.10.3. Trends in Adaptability and Customization of Trading Algorithms 		

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Module 20. Module 20. Ethical and Regulatory Aspects of Al in Finance

20.1. Ethics in Artificial Intelligence Applied to Finance

- 20.1.1. Fundamental Ethical Principles for the Development and Use of Al in Finance
- 20.1.2. Case Studies on Ethical Dilemmas in Financial Al Applications
- 20.1.3. Developing Ethical Codes of Conduct for Financial Technology Professionals

20.2. Global Regulations Affecting the Use of AI in Financial Markets

- 20.2.1. Overview of the Main International Financial Regulations on Al
- 20.2.2. Comparison of AI Regulatory Policies among Different Jurisdictions
- 20.2.3. Implications of AI Regulation on Financial Innovation

20.3. Transparency and Explainability of Al Models in Finance

- 20.3.1. Importance of Transparency in Al Algorithms for User Confidence
- 20.3.2. Techniques and Tools to Improve the Explainability of Al Models
- 20.3.3. Challenges of Implementing Interpretable Models in Complex Financial Environments

20.4. Risk Management and Ethical Compliance in the Use of Al

- 20.4.1. Risk Mitigation Strategies Associated with the Deployment of Al in Finance
- 20.4.2. Ethics Compliance in the Development and Application of Al Technologies
- 20.4.3. Ethical Oversight and Audits of Al Systems in Financial Operations

20.5. Social and Economic Impact of Al in Financial Markets

- 20.5.1. Effects of AI on the Stability and Efficiency of Financial Markets
- 20.5.2. Al and Its Impact on Employment and Professional Skills in Finance
- 20.5.3. Benefits and Social Risks of Large-Scale Financial Automation

20.6. Data Privacy and Protection in Al Financial Applications

- 20.6.1. Data Privacy Regulations Applicable to Al Technologies in Finance
- 20.6.2. Personal Data Protection Techniques in Al-Based Financial Systems
- 20.6.3. Challenges in Managing Sensitive Data in Predictive and Analytics Models

20.7. Algorithmic Bias and Fairness in Al Financial Models

- 20.7.1. Identification and Mitigation of Bias in Financial Al Algorithms
- 20.7.2. Strategies to Ensure Fairness in Automated Decision-Making Models
- 20.7.3. Impact of Algorithmic Bias on Financial Inclusion and Equity

20.8. Challenges of Regulatory Oversight in Financial Al

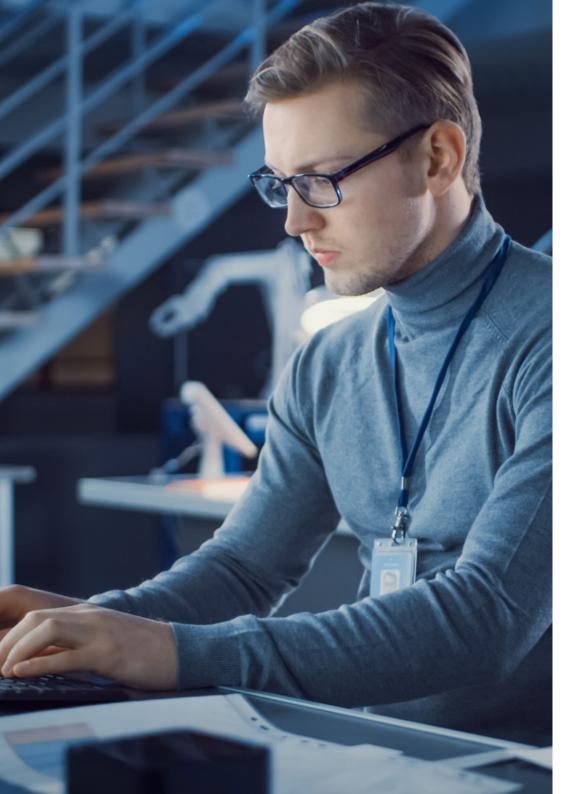
- 20.8.1. Difficulties in the Supervision and Control of Advanced Al Technologies
- 20.8.2. Role of Financial Authorities in the Ongoing Supervision of Al
- 20.8.3. Need for Regulatory Adaptation in the Face of Advancing Al Technology

20.9. Strategies for Responsible Development of Al Technologies in Finance

- 20.9.1. Best Practices for Sustainable and Responsible Al Development in the Financial Sector
- 20.9.2. Initiatives and Frameworks for Ethical Assessment of Al Projects in Finance
- 20.9.3. Collaboration between Regulators and Businesses to Encourage Responsible Practices

20.10. Future of Al Regulation in the Financial Sector

- 20.10.1. Emerging Trends and Future Challenges in Al Regulation in Finance
- 20.10.2. Preparation of Legal Frameworks for Disruptive Innovations in Financial Technology
- 20.10.3. International Dialogue and Cooperation for Effective and Unified Regulation of AI in Finance





You will address time series modeling and the application of explainable Artificial Intelligence, facilitating informed and accurate decision making in dynamic financial environments"



This academic program offers students a different way of learning. Our methodology uses a cyclical learning approach: **Relearning.**

This teaching system is used, for example, in the most prestigious medical schools in the world, and major publications such as the **New England Journal of Medicine** have considered it to be one of the most effective.





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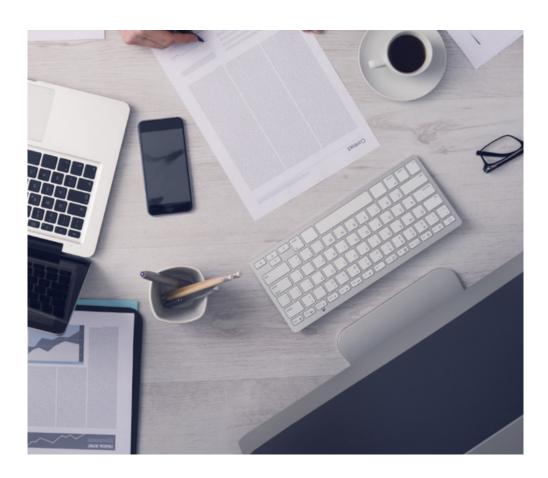
TECH Business School uses the Case Study to contextualize all content

Our program offers a revolutionary approach to developing skills and knowledge. Our goal is to strengthen skills in a changing, competitive, and highly demanding environment.





This program prepares you to face business challenges in uncertain environments and achieve business success.



Our program prepares you to face new challenges in uncertain environments and achieve success in your career.

A learning method that is different and innovative

This TECH program is an intensive educational program, created from scratch to present executives with challenges and business decisions at the highest level, whether at the national or international level. This methodology promotes personal and professional growth, representing a significant step towards success. The case method, a technique that lays the foundation for this content, ensures that the most current economic, social and business reality is taken into account.



You will learn, through collaborative activities and real cases, how to solve complex situations in real business environments"

The case method has been the most widely used learning system among the world's leading business schools for as long as they have existed. The case method was developed in 1912 so that law students would not only learn the law based on theoretical content. It consisted of presenting students with real-life, complex situations for them to make informed decisions and value judgments on how to resolve them. In 1924, Harvard adopted it as a standard teaching method.

What should a professional do in a given situation? This is the question we face in the case method, an action-oriented learning method. Throughout the program, the studies will be presented with multiple real cases. They must integrate all their knowledge, research, argue and defend their ideas and decisions.

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Relearning Methodology

TECH effectively combines the Case Study methodology with a 100% online learning system based on repetition, which combines different teaching elements in each lesson.

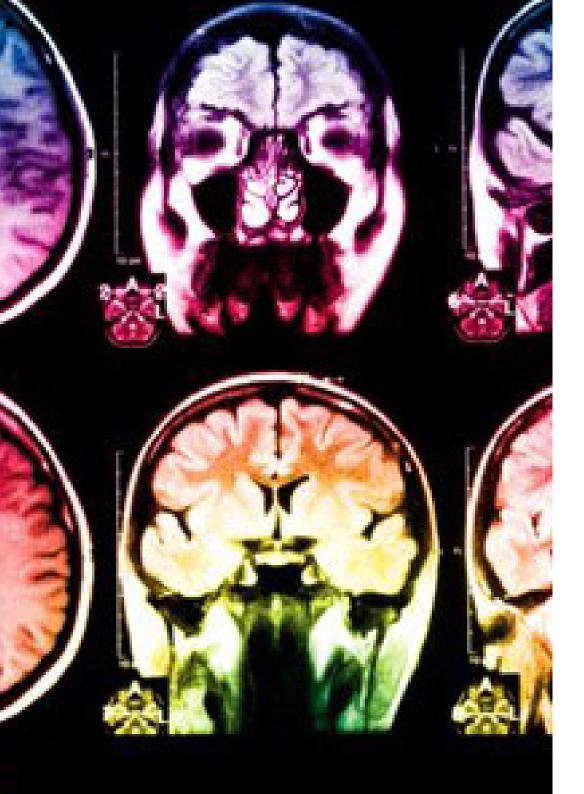
We enhance the Case Study with the best 100% online teaching method: Relearning.

Our online system will allow you to organize your time and learning pace, adapting it to your schedule. You will be able to access the contents from any device with an internet connection.

At TECH you will learn using a cutting-edge methodology designed to train the executives of the future. This method, at the forefront of international teaching, is called Relearning.

Our online business school is the only one in the world licensed to incorporate this successful method. In 2019, we managed to improve our students' overall satisfaction levels (teaching quality, quality of materials, course structure, objectives...) based on the best online university indicators.





Methodology | 53 tech

In our program, learning is not a linear process, but rather a spiral (learn, unlearn, forget, and re-learn). Therefore, we combine each of these elements concentrically. With this methodology we have trained more than 650,000 university graduates with unprecedented success in fields as diverse as biochemistry, genetics, surgery, international law, management skills, sports science, philosophy, law, engineering, journalism, history, markets, and financial instruments. All this in a highly demanding environment, where the students have a strong socio-economic profile and an average age of 43.5 years.

Relearning will allow you to learn with less effort and better performance, involving you more in your specialization, developing a critical mindset, defending arguments, and contrasting opinions: a direct equation to success.

From the latest scientific evidence in the field of neuroscience, not only do we know how to organize information, ideas, images and memories, but we know that the place and context where we have learned something is fundamental for us to be able to remember it and store it in the hippocampus, to retain it in our long-term memory.

In this way, and in what is called neurocognitive context-dependent e-learning, the different elements in our program are connected to the context where the individual carries out their professional activity.

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This program offers the best educational material, prepared with professionals in mind:



Study Material

All teaching material is produced by the specialists who teach the course, specifically for the course, so that the teaching content is highly specific and precise.

These contents are then applied to the audiovisual format, to create the TECH online working method. All this, with the latest techniques that offer high quality pieces in each and every one of the materials that are made available to the student.



Classes

There is scientific evidence suggesting that observing third-party experts can be useful.

Learning from an Expert strengthens knowledge and memory, and generates confidence in future difficult decisions.



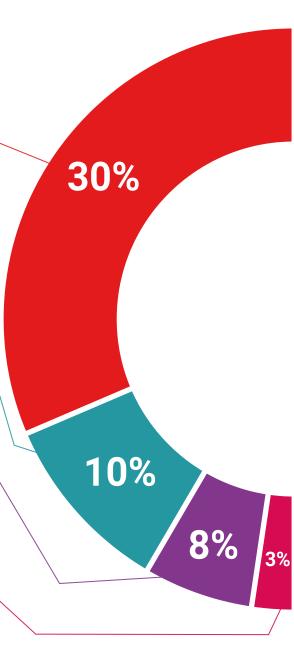
Management Skills Exercises

They will carry out activities to develop specific executive competencies in each thematic area. Practices and dynamics to acquire and develop the skills and abilities that a high-level manager needs to develop in the context of the globalization we live in.



Additional Reading

Recent articles, consensus documents and international guidelines, among others. In TECH's virtual library, students will have access to everything they need to complete their course.





Students will complete a selection of the best case studies chosen specifically for this program. Cases that are presented, analyzed, and supervised by the best senior management specialists in the world.



Interactive Summaries

The TECH team presents the contents attractively and dynamically in multimedia lessons that include audio, videos, images, diagrams, and concept maps in order to reinforce knowledge.

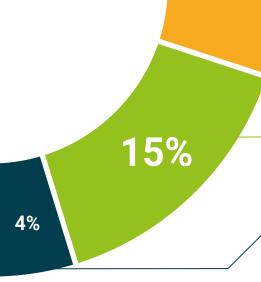


This exclusive educational system for presenting multimedia content was awarded by Microsoft as a "European Success Story".

Testing & Retesting

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We periodically evaluate and re-evaluate students' knowledge throughout the program, through assessment and self-assessment activities and exercises, so that they can see how they are achieving their goals.

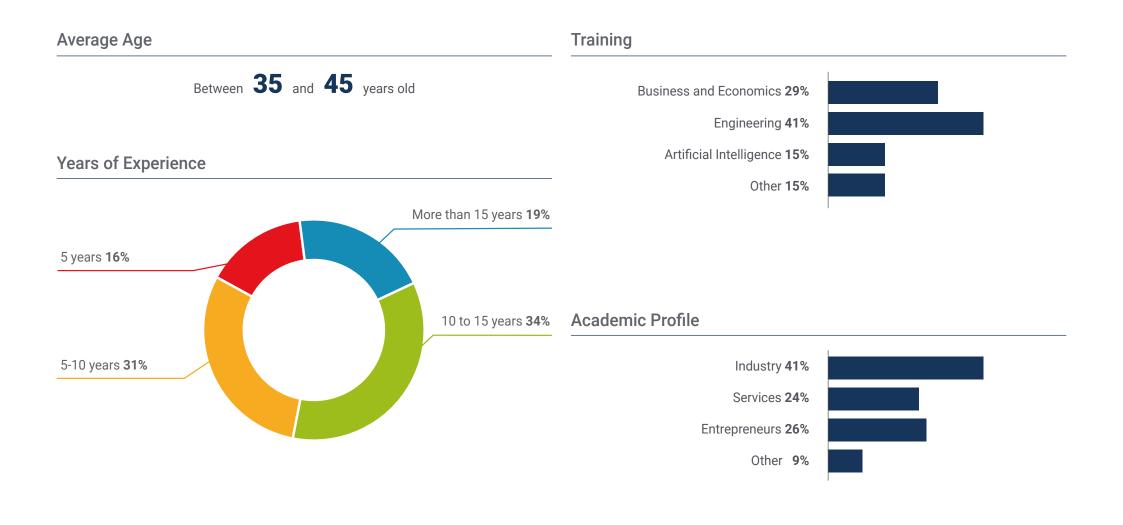


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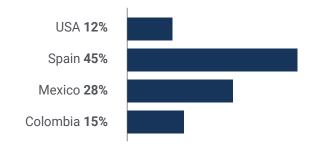




tech 58 | Our Students' Profiles



Geographical Distribution





Miriam Sánchez Aguado

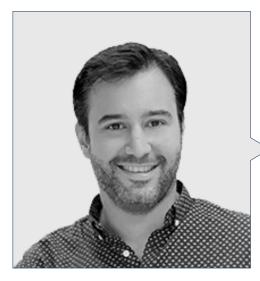
Financial Analyst

"Completing the Executive Master's Degree in Artificial Intelligence in Stock Market and Financial Markets has been an incredibly enriching experience for me. It has provided me with advanced tools and knowledge to apply Artificial Intelligence in a practical way in financial analysis and optimization of investment strategies. From learning how to use data visualization tools, to implementing Machine Learning techniques in algorithmic trading, the program has significantly broadened my perspective and skills. Now, I feel much more prepared to face market challenges with a technological edge and make more informed and accurate decisions in my role as a Financial Analyst."





Management



Dr. Peralta Martín-Palomino, Arturo

- CEO and CTO at Prometeus Global Solutions
- CTO at Korporate Technologies
- CTO at AI Shepherds GmbH
- Consultant and Strategic Business Advisor at Alliance Medical
- Director of Design and Development at DocPath
- PhD in Psychology from the University of Castilla La Mancha
- PhD in Economics, Business and Finance from the Camilo José Cela University
- PhD in Psychology from University of Castilla La Mancha
- Master's Degree in Executive MBA from the Isabel I University
- Master's Degree in Sales and Marketing Management, Isabel I University
- Expert Master's Degree in Big Data by Hadoop Training
- Master's Degree in Advanced Information Technologies from the University of Castilla La Mancha
- Member of: SMILE Research Group





Mr. Sánchez Mansilla, Rodrigo

- Digital Advisor at Al Shephers GmbH
- Digital Account Manager at Kill Draper
- Head of Digital at Kuarere
- Digital Marketing Manager at Arconi Solutions, Deltoid Energy and Brinergy Tech
- Founder and National Sales and Marketing Manager
- Master's Degree in Digital Marketing (MDM) by The Power Business School
- Bachelor's Degree in Business Administration (BBA) from the University of Buenos Aires



Take the opportunity to learn about the latest advances in this field in order to apply it to your daily practice"







You will immediately apply the knowledge acquired in real scenarios, which will result in a comprehensive and updated preparation to face the challenges of the sector. With all TECH's quality guarantees!.

Are you ready to take the leap? Excellent professional development awaits you

The Executive Master's Degree in Artificial Intelligence in Stock Exchanges and Financial Markets from TECH is an intensive program that prepares you to face challenges and business decisions in the field of Artificial Intelligence applied to the Financial Department of companies. Its main objective is to promote your personal and professional growth. Helping you achieve success.

If you want to improve yourself, make a positive change at a professional level, and network with the best, then this is the place for you.

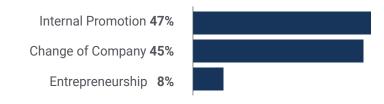
You will be able to demonstrate your skills in a growing discipline, enhancing your professional profile and opening doors to job opportunities in an increasingly technology-dependent industry.

Time of Change

During the program 63%

After 2 years 26%

Type of change



Salary increase

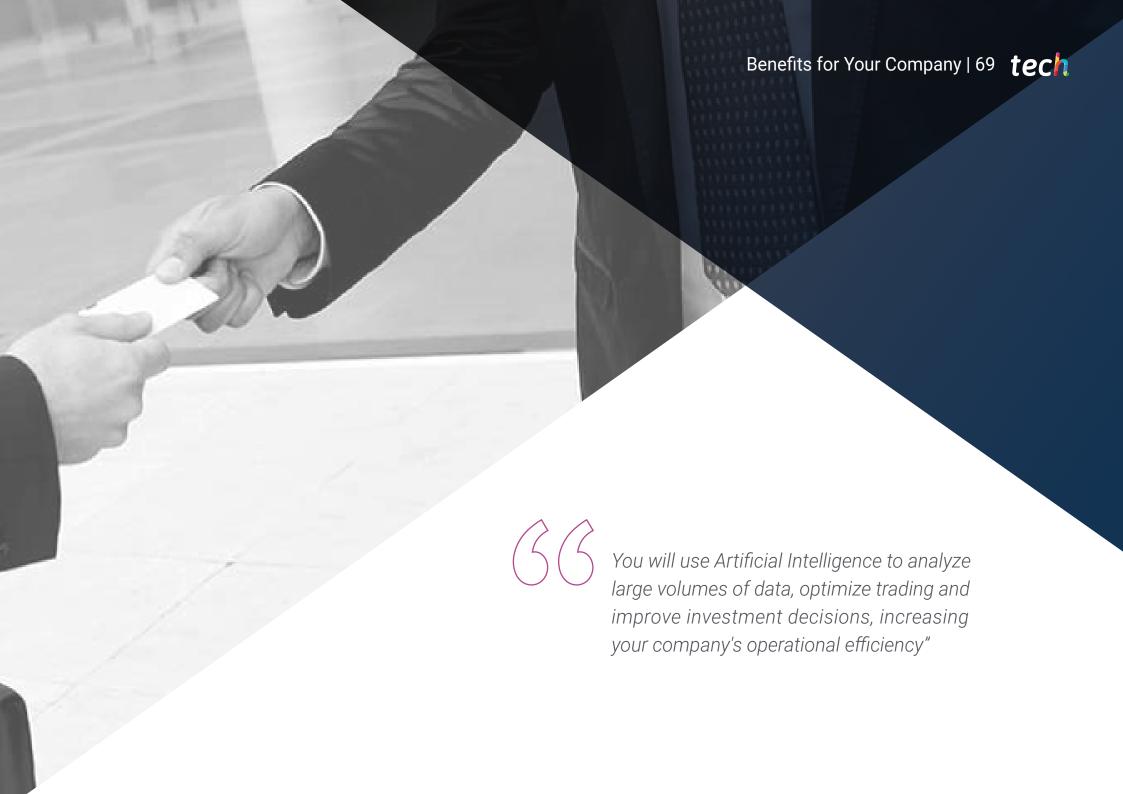
This program represents a salary increase of more than 26.24% for our students

\$52,000

A salary increase **26.24%**

\$65,644





tech 70 | Benefits for Your Company

Developing and retaining talent in companies is the best long-term investment.



Growth of talent and intellectual capital

The professional will introduce the company to new concepts, strategies, and perspectives that can bring about significant changes in the organization.



Retaining high-potential executives to avoid talent drain

This program strengthens the link between the company and the professional and opens new avenues for professional growth within the company.



Building agents of change

You will be able to make decisions in times of uncertainty and crisis, helping the organization overcome obstacles.



Increased international expansion possibilities

Thanks to this program, the company will come into contact with the main markets in the world economy.







Project Development

The professional can work on a real project or develop new projects in the field of R & D or business development of your company.



Increased competitiveness

This program will equip students with the skills to take on new challenges and drive the organization forward.





tech 74 | Certificate

This private qualification will allow you to obtain a **Executive Master's Degree in Artificial Intelligence in Stock Exchanges and Financial Markets** endorsed by **TECH Global University**, the world's largest online university.

TECH Global University is an official European University publicly recognized by the Government of Andorra (*official bulletin*). Andorra is part of the European Higher Education Area (EHEA) since 2003. The EHEA is an initiative promoted by the European Union that aims to organize the international training framework and harmonize the higher education systems of the member countries of this space. The project promotes common values, the implementation of collaborative tools and strengthening its quality assurance mechanisms to enhance collaboration and mobility among students, researchers and academics.

This **TECH Global University** private qualification is a European program of continuing education and professional updating that guarantees the acquisition of competencies in its area of knowledge, providing a high curricular value to the student who completes the program.

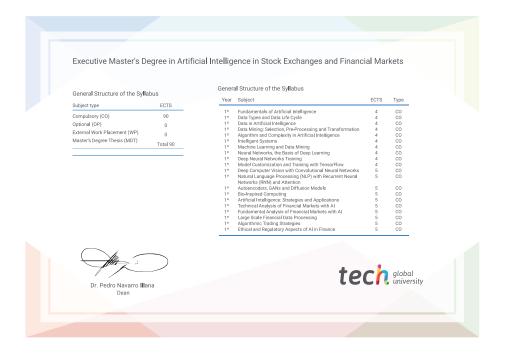
Title: Executive Master's Degree in Artificial Intelligence in Stock Exchanges and Financial Markets

Modality: online

Duration: 12 months.

Accreditation: 90 ECTS





^{*}Apostille Convention. In the event that the student wishes to have their paper diploma issued with an apostille, TECH Global University will make the necessary arrangements to obtain it, at an additional cost.



Executive Master's Degree

Artificial Intelligence in Stock Exchanges and Financial Markets

- » Modality: online
- » Duration: 12 months.
- » Certificate: TECH Global University
- » Accreditation: 90 ECTS
- » Schedule: at your own pace
- » Exams: online

