




Università di Pisa **Data Mining and Machine Learning** *Francesco Marcelloni*
Bioinspired computational methods
Biological data mining




Introduction

Francesco Marcelloni
Department of Information Engineering
University of Pisa
ITALY
e-mail: francesco.marcelloni@unipi.it





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Università di Pisa **Programme 6 ECTs** *Francesco Marcelloni*



- Data Preprocessing
- Classification and Prediction
- Cluster Analysis
- Mining Frequent Patterns, Associations and Correlations



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Programme 12 ECTs

- Data Preprocessing
- Classification and Prediction
- Cluster Analysis
- Mining Frequent Patterns, Associations and Correlations
- Outlier detection
- Mining Stream, Time-series and Sequence Analysis
- Graph Mining
- Streaming Data Mining
- Hadoop and Mahout



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Course

- Lectures: prof. Francesco Marcelloni
 - Reception hours: Wednesday 15-18 (please, send me an e-mail for confirmation)
- Practical and laboratory work: prof. Alessandro Renda
 - Reception hours: Monday 16-18



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Material

■ Teaching material:

- J. Han and M. Kamber. Data Mining: Concepts and Techniques. Morgan Kaufmann, Third edition
- Jiawei Han, Jian Pei, Hanghang Tong, Data Mining: Concepts and Techniques. Morgan Kaufmann, Fourth edition
- Papers on the different algorithms described during the course

■ Slides available in the Teams channel of the course



Acknowledgement

Some slides belong to the collection

Jiawei Han, Micheline Kamber, and Jian Pei
University of Illinois at Urbana-Champaign &
Simon Fraser University
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This collection accompanies the book

J. Han and M. Kamber. Data Mining: Concepts and
Techniques. Morgan Kaufmann, 3rd ed., 2011



Project (12 CFU)

■ Project

- The project consists of the **development of one application** which exploits one or more techniques introduced in the lectures. Applications can be stand-alone, Web applications, mobile applications and so on. For instance, event detection, recommender systems, user profiling, sentiment analysis. **The applications can be developed in groups of two persons at most.**

■ Schedule

- Application specification (**has to be approved**)
- Analysis and Design of the application
- Implementation (**recommended Python**) and Validation
- Presentation of the application 2-3 days before the examination (you have to deliver source code, executable version and documentation)



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Examination (12 CFU)

■ Examination

- 2-3 days before the official date of the examination:
 - **Presentation of the application and discussion.** The date for the presentation will be agreed with us. The application will be assessed in terms of originality, coherence with the specifications, appropriate choice of the data mining techniques, results obtained in the validation, usability.
 - A mark (between 18 and 30 in case of positive evaluation) will be assigned to each project. Only students who have obtained positive evaluations will be entitled to take the examination.
- Date of the examination: written test on theoretical aspects of the course and final discussion on the answers or directly an oral.
 - The final mark will be computed as the average of the marks obtained in the two evaluations.



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Examination (6 CFU)

■ Examination

- Date of the examination:
 - **Practical test using Python.** We will give you some dataset to be analyzed by using specific methods
 - After the analysis you will prepare a short presentation with the results of the analysis
 - A mark (between 18 and 30 in case of positive evaluation) will be assigned to the presentation. Only students who have obtained positive evaluations will be entitled to participate to the second test.
- Written\Oral test: written test on theoretical aspects of the course and final discussion on the answers or oral.
 - The final mark will be computed as the average of the marks obtained in the two evaluations.



Glossary

- **Machine Learning**
 - (The English Oxford Dictionary)
"The capacity of a computer to learn from experience, i.e. to modify its processing on the basis of newly acquired information."
- **Data Mining**
 - (Merriam-Webster)
 - the practice of searching through large amounts of computerized data to find useful patterns or trends



Data mining techniques



Machine Learning

• Machine Learning

Supervised Learning

Learning with a labeled training set

Unsupervised Learning

Discovering patterns in unlabeled data

Reinforcement Learning

Learning based on feedback or reward

Semi-supervised Learning



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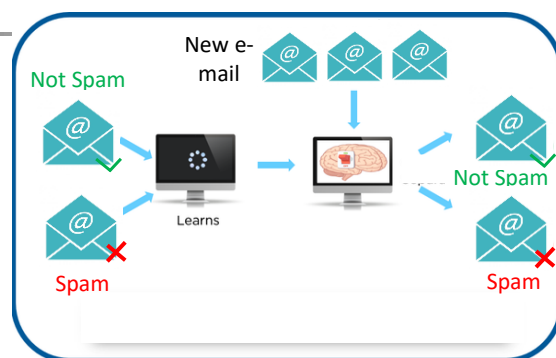
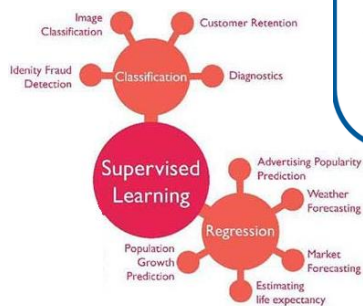


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Supervised Learning

• Supervised Learning

Learning with a labeled training set



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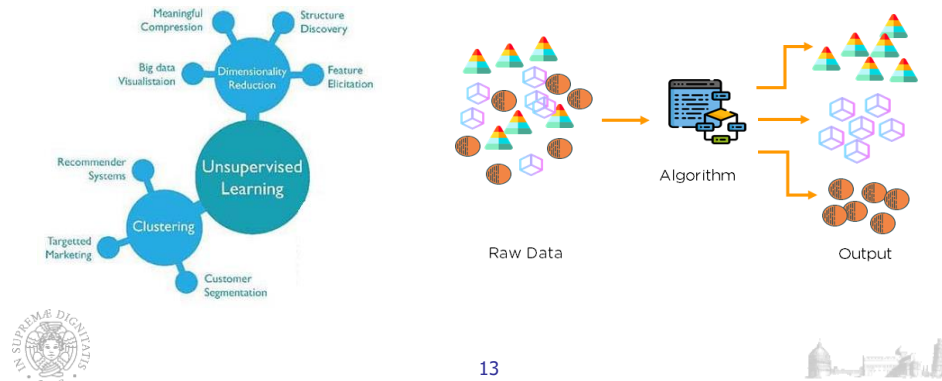


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Unsupervised Learning

• Unsupervised Learning

Discovering patterns in unlabeled data

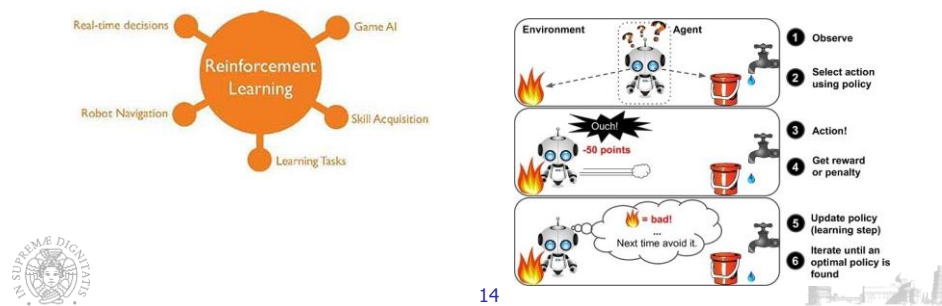


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Reinforcement Learning

• Reinforcement Learning

Learning based on feedback or reward.
No dataset needed at beginning



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What is Data Mining

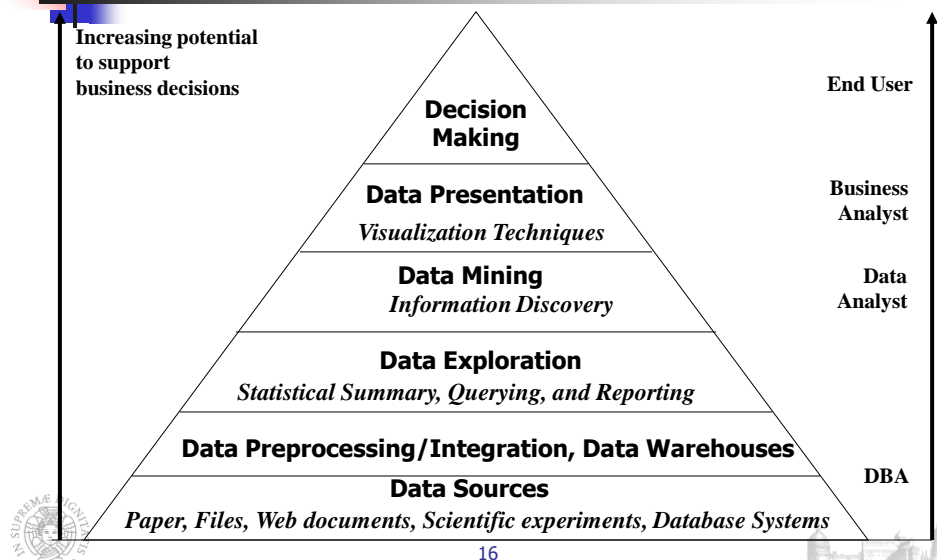


- **Data mining** (knowledge discovery from data)
 - Extraction of interesting (non-trivial, implicit, previously unknown and potentially useful) patterns or knowledge from huge amount of data
 - Data mining: a misnomer?
- Alternative names
 - **Knowledge discovery** (mining) in databases (KDD), knowledge extraction, data/pattern analysis, data archeology, data dredging, information harvesting, business intelligence, etc.
- Watch out: Is everything "data mining"?
 - Simple search and query processing
 - (Deductive) expert systems



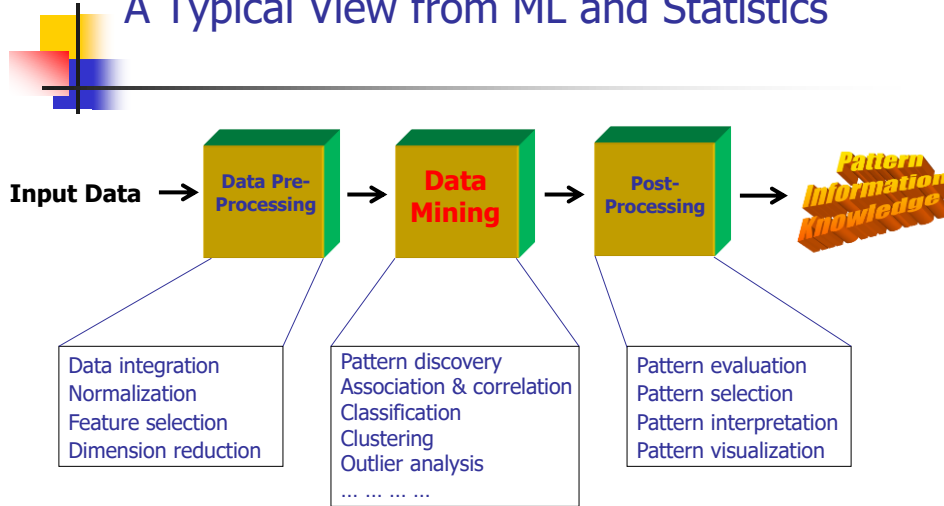
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Data Mining in Business Intelligence



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A Typical View from ML and Statistics



This is a view from typical machine learning and statistics communities

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Data Mining: On What Kinds of Data?

- Database-oriented data sets and applications
 - Relational database, data warehouse, transactional database
- Advanced data sets and advanced applications
 - Data streams and sensor data
 - Time-series data, temporal data, sequence data (incl. bio-sequences)
 - Structure data, graphs, social networks and multi-linked data
 - Object-relational databases
 - Heterogeneous databases and legacy databases
 - Spatial data and spatiotemporal data
 - Multimedia database
 - Text databases
 - The World-Wide Web

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Data Mining Function: Classification

- **Classification and label prediction**

- Construct models (functions) based on some training examples
- Describe and distinguish classes or concepts for future prediction
 - E.g., classify countries based on (climate), or classify cars based on (gas mileage)
- Predict some unknown class labels

- **Typical methods**

- Decision trees, naïve Bayesian classification, support vector machines, neural networks, rule-based classification, pattern-based classification, logistic regression, ...

- **Typical applications**

- Credit card fraud detection, direct marketing, classifying stars, diseases, web-pages, ...



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Data Mining Function: Cluster Analysis

- **Unsupervised learning** (i.e., Class label is unknown)
- Group data to form new categories (i.e., clusters), e.g., cluster houses to find distribution patterns
- Principle: Maximizing intra-class similarity & minimizing interclass similarity
- Many methods and applications



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Data Mining Function: Association and Correlation Analysis

- **Frequent patterns** (or frequent itemsets)
 - What items are frequently purchased together in your Walmart?
- **Association, correlation vs. causality**
 - A typical association rule
 - Diaper \rightarrow Beer [0.5%, 75%] (support, confidence)
 - Are strongly associated items also strongly correlated?
- How to mine such patterns and rules efficiently in large datasets?
- How to use such patterns for classification, clustering, and other applications?



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Data Mining Function: Outlier Analysis

- **Outlier analysis**
 - Outlier: A data object that does not comply with the general behavior of the data
 - Noise or exception? — One person's garbage could be another person's treasure
 - Methods: by product of clustering or regression analysis, ...
 - Useful in fraud detection, rare events analysis



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Time and Ordering: Sequential Pattern, Trend and Evolution Analysis

- **Sequence, trend and evolution analysis**
 - Trend, time-series, and deviation analysis: e.g., regression and value prediction
 - Sequential pattern mining
 - e.g., first buy digital camera, then buy large SD memory cards
 - Periodicity analysis
 - Sequence Motifs (nucleotide or amino-acid sequence pattern) and biological sequence analysis
 - Approximate and consecutive motifs
 - Similarity-based analysis
- **Mining data streams**
 - Ordered, time-varying, potentially infinite, data streams

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Structure and Network Analysis

- **Graph mining**
 - Finding frequent subgraphs (e.g., chemical compounds), trees (XML), substructures (web fragments)
- **Information network analysis**
 - Social networks: actors (objects, nodes) and relationships (edges)
 - e.g., author networks in CS, terrorist networks
 - Multiple heterogeneous networks
 - A person could be multiple information networks: friends, family, classmates, ...
 - Links carry a lot of semantic information: Link mining
- **Web mining**
 - Web is a big information network: from PageRank to Google
 - Analysis of Web information networks
 - Web community discovery, opinion mining, usage mining, ...

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Evaluation of Knowledge

- Are all mined knowledge interesting?
 - One can mine tremendous amount of "patterns" and knowledge
 - Some may fit only certain dimension space (time, location, ...)
 - Some may not be representative, may be transient, ...
- Evaluation of mined knowledge → directly mine only interesting knowledge?
 - Descriptive vs. predictive
 - Coverage
 - Typicality vs. novelty
 - Accuracy
 - Timeliness
 - ...

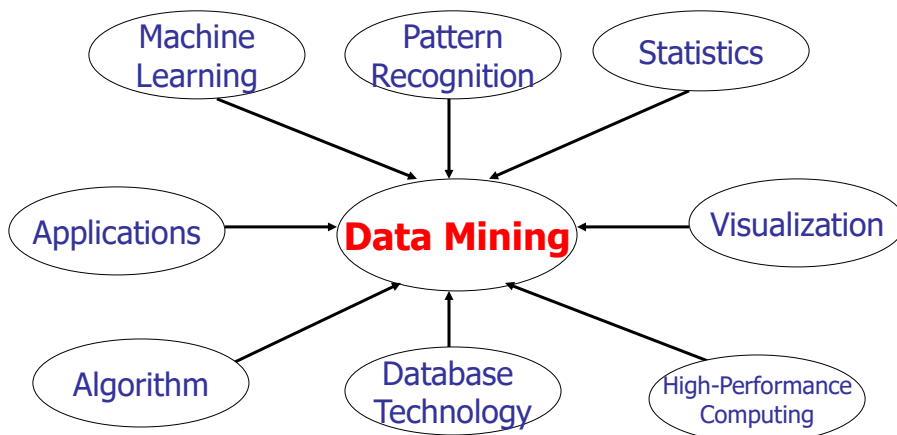


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Data Mining: Confluence of Multiple Disciplines



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Why Confluence of Multiple Disciplines?

- Tremendous amount of data
 - Algorithms must be highly scalable to handle such as tera-bytes of data
- High-dimensionality of data
 - Micro-array may have tens of thousands of dimensions
- High complexity of data
 - Data streams and sensor data
 - Time-series data, temporal data, sequence data
 - Structure data, graphs, social networks and multi-linked data
 - Heterogeneous databases and legacy databases
 - Spatial, spatiotemporal, multimedia, text and Web data
 - Software programs, scientific simulations
- New and sophisticated applications

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Major Issues in Data Mining (1)

- Mining Methodology
 - Mining various and new kinds of knowledge
 - Mining knowledge in multi-dimensional space
 - Data mining: An interdisciplinary effort
 - Boosting the power of discovery in a networked environment
 - Handling noise, uncertainty, and incompleteness of data
 - Pattern evaluation and pattern- or constraint-guided mining
- User Interaction
 - Interactive mining
 - Incorporation of background knowledge
 - Presentation and visualization of data mining results

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Major Issues in Data Mining (2)

- **Efficiency and Scalability**
 - Efficiency and scalability of data mining algorithms
 - Parallel, distributed, stream, and incremental mining methods
- **Diversity of data types**
 - Handling complex types of data
 - Mining dynamic, networked, and global data repositories
- **Data mining and society**
 - Social impacts of data mining
 - Privacy-preserving data mining
 - Invisible data mining
 - Trustworthiness



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A Brief History of Data Mining Society

- **1989 IJCAI Workshop on Knowledge Discovery in Databases**
 - Knowledge Discovery in Databases (G. Piatetsky-Shapiro and W. Frawley, 1991)
- **1991-1994 Workshops on Knowledge Discovery in Databases**
 - Advances in Knowledge Discovery and Data Mining (U. Fayyad, G. Piatetsky-Shapiro, P. Smyth, and R. Uthurusamy, 1996)
- **1995-1998 International Conferences on Knowledge Discovery in Databases and Data Mining (KDD'95-98)**
 - Journal of Data Mining and Knowledge Discovery (1997)
- **ACM SIGKDD conferences since 1998 and SIGKDD Explorations**
- **More conferences on data mining**
 - PAKDD (1997), PKDD (1997), SIAM-Data Mining (2001), (IEEE) ICDM (2001), etc.
- **ACM Transactions on KDD starting in 2007**



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Conferences and Journals on Data Mining

KDD Conferences

ACM SIGKDD Int. Conf. on Knowledge Discovery in Databases and Data Mining (**KDD**)

SIAM Data Mining Conf. (**SDM**)
(IEEE) Int. Conf. on Data Mining (**ICDM**)

European Conf. on Machine Learning and Principles of Knowledge Discovery and Data Mining (**ECML-PKDD**)

Pacific-Asia Conf. on Knowledge Discovery and Data Mining (**PAKDD**)
Int. Conf. on Web Search and Data Mining (**WSDM**)

Other related conferences

- DB conferences: ACM SIGMOD, VLDB, ICDE, EDBT, ICDT, ...
- Web and IR conferences: WWW, SIGIR, WSDM
- ML conferences: ICML, NIPS
- PR conferences: CVPR,

Journals

- Data Mining and Knowledge Discovery (DAMI or DMKD)
- IEEE Trans. On Knowledge and Data Eng. (TKDE)
- KDD Explorations
- ACM Trans. on KDD



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Where to Find References? DBLP, CiteSeer, Google

Data mining and KDD (SIGKDD: CDROM)

Conferences: ACM-SIGKDD, IEEE-ICDM, SIAM-DM, PKDD, PAKDD, etc.

Journal: Data Mining and Knowledge Discovery, KDD Explorations, ACM TKDD

Database systems (SIGMOD: ACM SIGMOD Anthology—CD ROM)

Conferences: ACM-SIGMOD, ACM-PODS, VLDB, IEEE-ICDE, EDBT, ICDT, DASFAA

Journals: IEEE-TKDE, ACM-TODS/TOIS, JIIS, J. ACM, VLDB J., Info. Sys., etc.

AI & Machine Learning

Conferences: Machine learning (ML), **AAAI**, IJCAI, COLT (Learning Theory), CVPR, NIPS, etc.

Journals: Machine Learning, Artificial Intelligence, Knowledge and Information Systems, IEEE-PAMI, etc.



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Where to Find References? DBLP, CiteSeer, Google

- Web and IR
 - Conferences: SIGIR, WWW, CIKM, etc.
 - Journals: WWW: Internet and Web Information Systems,
- Statistics
 - Conferences: Joint Stat. Meeting, etc.
 - Journals: Annals of statistics, etc.
- Visualization
 - Conference proceedings: CHI, ACM-SIGGraph, etc.
 - Journals: IEEE Trans. visualization and computer graphics, etc.



Why only today?

- **Technological development:** vast computing power and nearly infinite storage capacity enable the execution of increasingly complex machine learning algorithms using enormous amounts of data.

Market cap history of NVIDIA from 2001 to 2023



The sixth most valuable company in the world by market capitalization.



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Why only today?

- New learning algorithms:** increasingly advanced and sophisticated algorithms capable of training ever more complex models
 - GPT-1: 117 million parameters
 - GPT-2: 1.5 billion parameters
 - GPT-3: 175 billion parameters
 - Google Bard: 137 billion parameters

GPT-3: **26 days and 1248 MWh** for training (the equivalent of the annual electricity consumption of approximately 400 households with 4 people)

Google Bard: 13 days and 312 MWh for training






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Why only today?

- Availability of large amounts of data:** thanks to social networks, digital platforms, and the Internet of Things, which allows us to collect data from sensors deployed everywhere, etc.
 - GPT-3: **45TB** of text data from different datasets

Dataset	Quantity (tokens)
Common Crawl (filtered)	410 billion
WebText2	19 billion
Books1	12 billion
Books2	55 billion
Wikipedia	3 billion



Why only today?

- Impact on everyday life: AI has become pervasive



<https://techvidvan.com/tutorials/ai-in-human-life/>

- Significant investments: AI is a disruptive technology, and both companies and nations are investing heavily in it.

GPT-3: \$10 billion (with \$3 billion already invested)

Bard: \$300 million

