

# IS4242

# INTELLIGENT SYSTEMS & TECHNIQUES

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L1 – Introduction

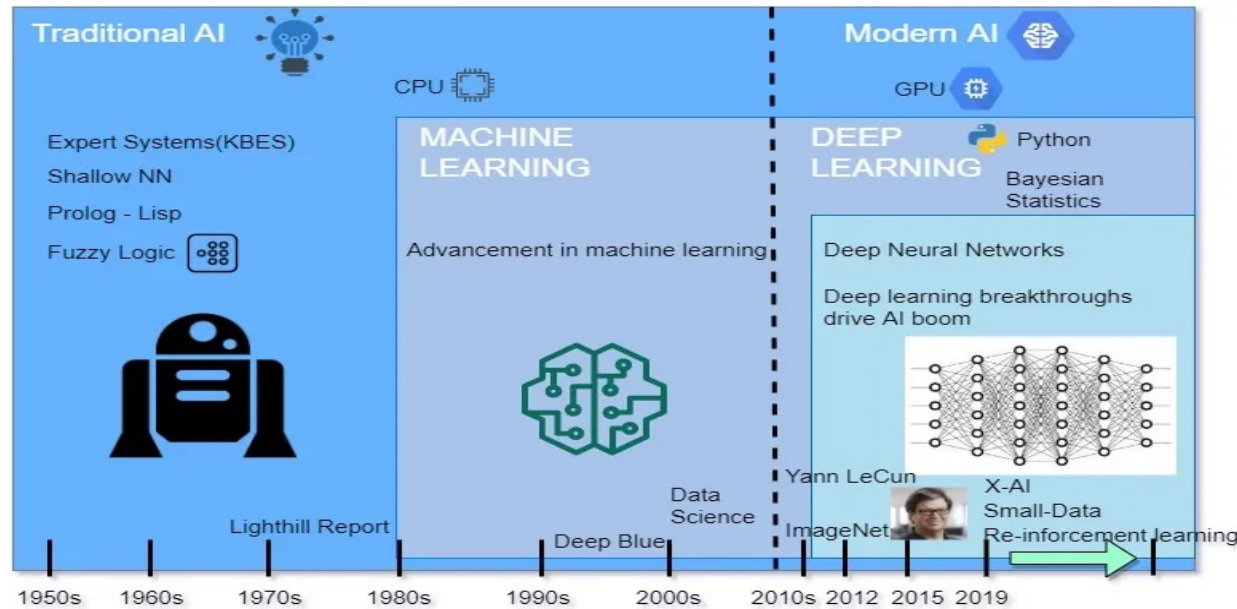
Aditya Karanam

# Intelligent Systems for Business

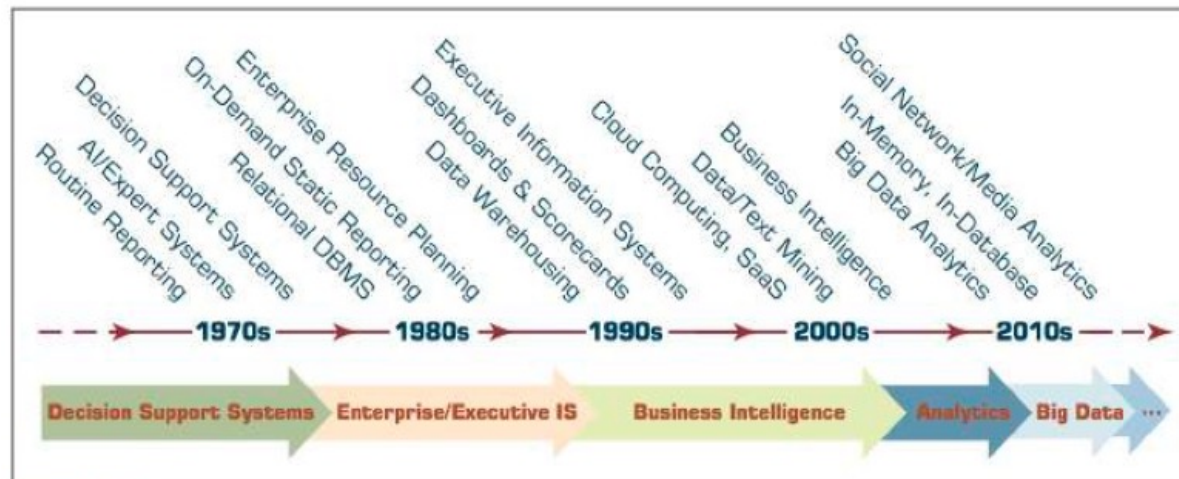
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- ▶ Every company wants to know:
  - ▶ What is happening?
  - ▶ What is likely to happen?
  - ▶ What strategy provides the best outcome in this situation?
- ▶ *Collect data and extract meaningful insights* to make every possible decision
  - ▶ Transform itself into an AI-first company!
- ▶ But AI has been around since 1950, why now?
  - ▶ “*Machines will be capable of doing any work a man can do.*” – Herbert A. Simon (1965)

# Evolution of AI and Business Intelligent Systems



AI has gone through many ups and downs



But companies were always after the Data

# Why do companies need Intelligent Systems?

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- ▶ Data helps you to measure
  - ▶ You can't solve problems you don't know about
- ▶ This data revolution has not only created efficient processes but also created data-driven products!
  - ▶ Ex: Tesla builds cars by developing software on unique hardware, much in the way Apple develops the iPhone or Microsoft leverages Intel chips and Dell PCs
- ▶ This digital transformation has created a *hyper-turbulent* and *hyper-competitive* market in almost every industry

# Why do companies need Intelligent Systems?

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- ▶ Hyper – turbulent:
  - ▶ Highly volatile market: difficult to maintain the sustainable and competitive advantage for a long term
- ▶ Hyper-competitive:
  - ▶ Highly concentrated market, where winner takes all
  - ▶ The entire demand goes to the best products, but these best products keep changing over time
  - ▶ Decisions should not be made by the whim but based on facts (data)!
- ▶ Crucial to be an AI-First company: leverage data to make strategic decision and build competitive advantage

# Business Use Cases: Pricing Strategy

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- ▶ Nothing is more important than ensuring products are priced appropriately
  - ▶ Charge too little – you leave revenue on the table
  - ▶ Charge too much – alienate and push customers to competitors
- ▶ Willingness to pay
  - ▶ The maximum price that a customer is ready to pay for a product or service
- ▶ How do you learn the willingness to pay of your potential customers?
  - ▶ Survey?
    - ▶ They would say: they want *the best of everything* by paying *as little as possible*
  - ▶ Don't believe what they say but what they do!
    - ▶ Past purchase behaviours that reveal consumer preferences

# Business Use Cases: Coupons

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- ▶ Coupons /Promotions
  - ▶ Target *valued* customers to improve *profitability*
- ▶ Groupon sold coupons called Groupons
  - ▶ Purchasers buy goods at discount prices from participating merchants
  - ▶ Merchants paid nothing for advertising unless a customer made a purchase
  - ▶ Groupon's sales grew rapidly and exceeded \$2.5 billion in 2013
- ▶ Is this Business Model viable?



# Business Use Cases: Coupons

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- ▶ Groupon spent heavily on marketing to acquire customers
  - ▶ Customers got tired of receiving discount offers
  - ▶ Merchants found that offering Groupons only brought them unprofitable customers, who don't come back.
- ▶ Where is the problem?
  - ▶ In the data! Groupon does not know :
    - ▶ Consumer purchase behaviour when the Groupons are not used
    - ▶ Who are the valued customers for each of their merchants
- ▶ A better way is to build a classifier to see which *valued* customers would encash the coupon



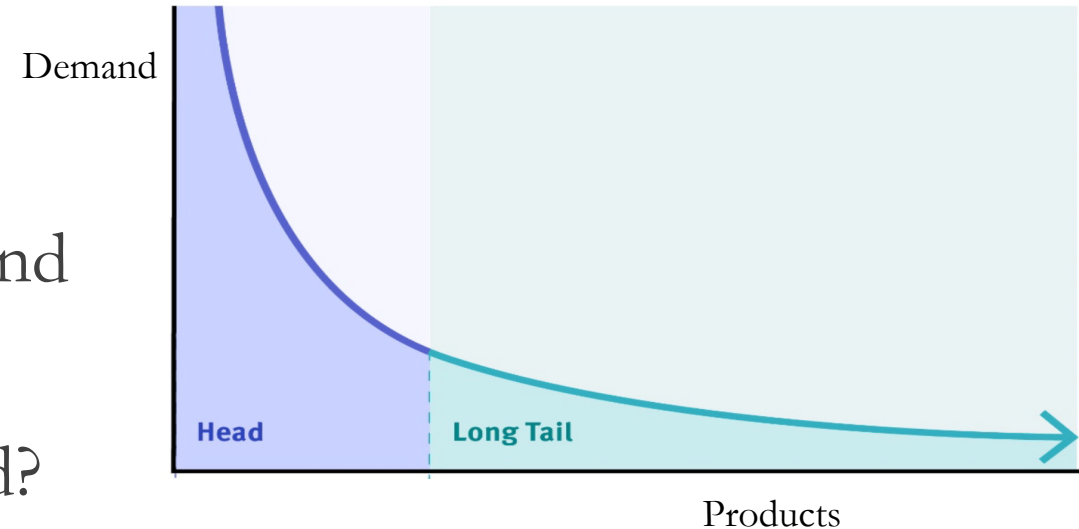
# Business Use Cases Example: Market Segmentation

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- ▶ How do you find new customers?
- ▶ Learn as much as we can about current customers and their similarities
- ▶ Group customers based on their demographics or purchase behavior
- ▶ These groups should provide information on potentially new customers as well as existing customers!
- ▶ Design products to target these groups accordingly
  - ▶ Ex: In an audio equipment company:
    - ▶ Casual listeners: any speakers, headphones or ear buds
    - ▶ Music lovers: the highest quality in audio equipment

# Business Use Cases: Matching Demand & Supply

- ▶ Amazon has a catalog of 12 billion products and 310 million users
  - ▶ Long tail of suppliers
  - ▶ Few suppliers have high demand and a lot of suppliers have very low demand
- ▶ How to help quality suppliers gain demand?
- ▶ How to help customers find quality products they like?
- ▶ One simple way: Efficient search functionality
  - ▶ Better way: Recommendation systems!

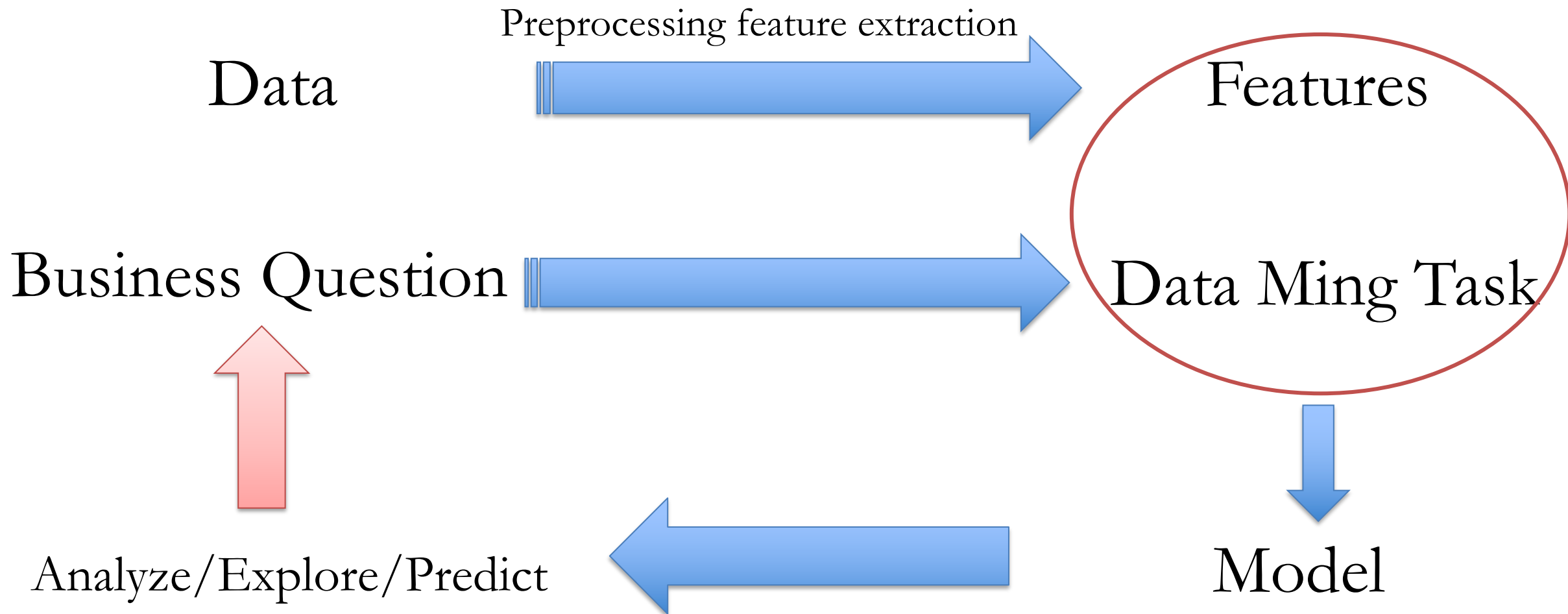


# Other Pertinent Business Questions

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- ▶ Mobile Apps: How do you retain customers?
- ▶ Product design: N ways of designing a feature but how do you know which is the best?
- ▶ Social media monitoring:
  - ▶ Enormous amounts of user data, which can be used to obtain customer preferences
  - ▶ Data is unstructured: text, images, and videos
- ▶ Competitor Analysis: What is the market structure and who are your competitors?

# 'Intelligence' of the Intelligent System: Model



# Learning Paradigms of Data Mining Tasks

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- ▶ Supervised
- ▶ Unsupervised
- ▶ Others
  - ▶ Semi-supervised
  - ▶ Active Learning
  - ▶ Reinforcement Learning

# Supervised Learning

- ▶ Training Data:  $(\mathbf{X}_1, y_1), (\mathbf{X}_2, y_2), \dots, (\mathbf{X}_n, y_n)$

- ▶ Learning function:  $y = g(\mathbf{X})$

- ▶  $y$  is *numerical or ordinal* : Regression

	$f_1$	$f_2$	...	$Y$
$\mathbf{X}_1$				$y_1$
$\mathbf{X}_2$				$y_2$
$\mathbf{X}_3$				$y_3$

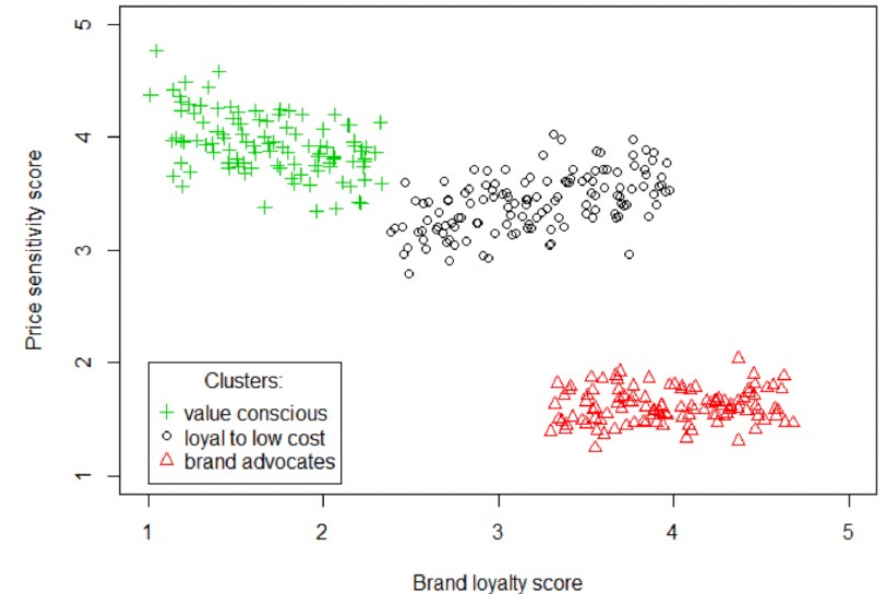
- ▶  $y$  is *categorical* : Classification (binary or multiclass)

- ▶ Prediction:  $g(\mathbf{X}_i) = \hat{y}_i$

# Unsupervised Learning

- ▶ No labeled training data:  $\{(\mathbf{X}_1, \mathbf{X}_2, \dots, \mathbf{X}_n)\}$
- ▶ Discover structure in data
- ▶ Clustering is  $g(X) = y$ 
  - ▶ Partitioning: K-means, Fuzzy C-Means, etc.
  - ▶ Hierarchical Clustering

	$f_1$	$f_2$	...
$\mathbf{X}_1$			
$\mathbf{X}_2$			
$\mathbf{X}_3$			



<https://select-statistics.co.uk/blog/customer-segmentation/>

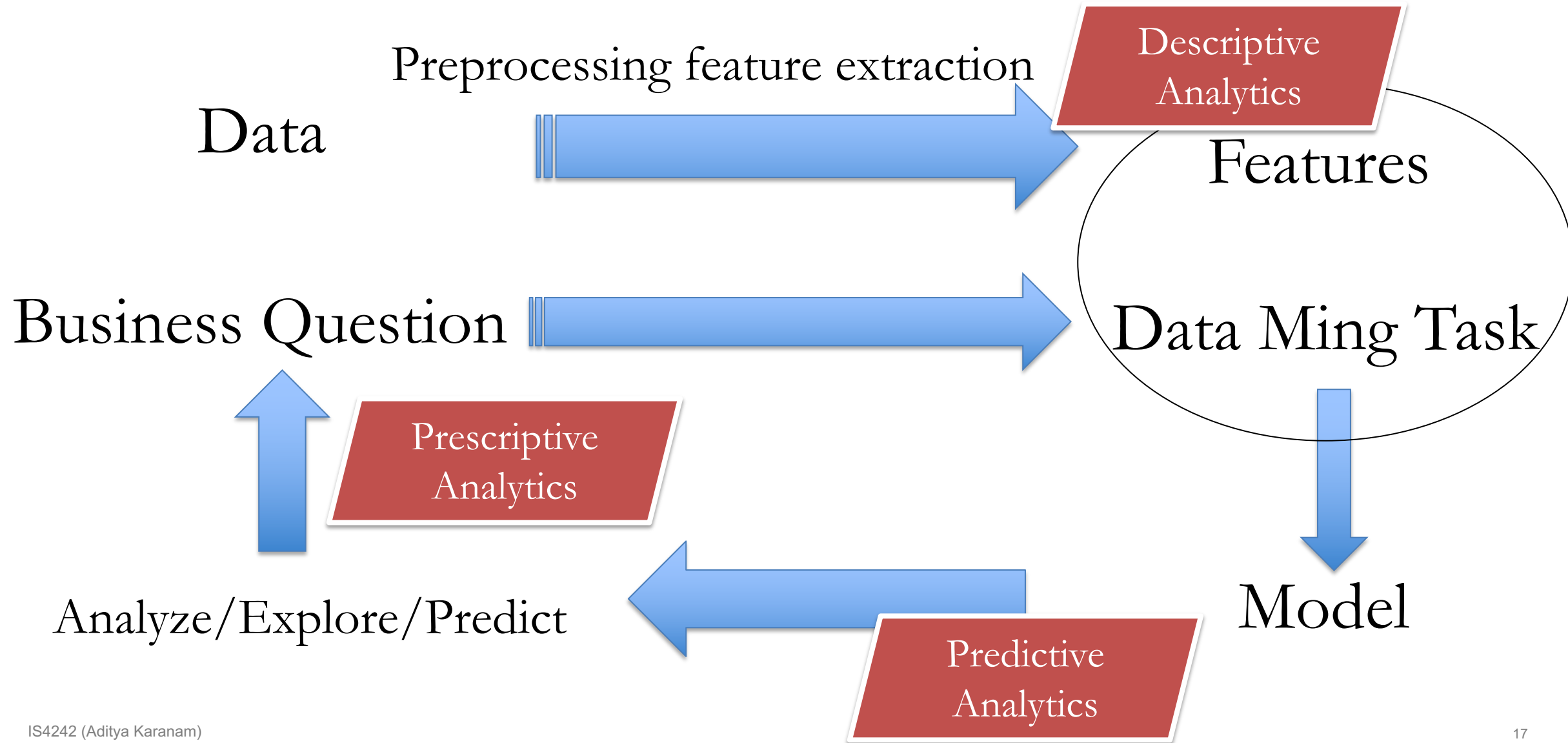
# Building Intelligent Systems: Example

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- ▶ Does providing discounts improve the sales of a product?
  - ▶ *Data Mining Task: Prediction (Regression)*
  - ▶  $Sales = \alpha_0 + \alpha_1 discount + \varepsilon$ 
    - ▶ Building the model corresponds to identifying  $\alpha$
  - ▶ Algorithm: Maximum Likelihood Estimation (MLE)
  - ▶ Theory: Which algorithm is correct – MLE, Least Squares, etc.?
- ▶ Primarily: Modeling and Application
  - ▶ Algorithm and Theory in some cases



# 'Intelligence' of the Intelligent System: Model



# References

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- ▶ Sharda et al., Business Intelligence, Analytics, and Data Science: A Managerial Perspective
- ▶ <https://hbr.org/2008/07/investing-in-the-it-that-makes-a-competitive-difference>

# Thank You

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