



Model-based Discrimination Analysis: A Position Paper

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Basics

- Decision-making software may lead to undesirable discrimination:
 - exploiting sensitive data (e.g., race)
 - learning correlations between a set of data.

- 1 Black box approaches
- White box approaches

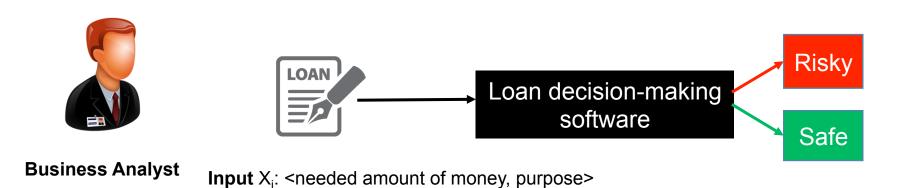
Require the software to be implemented.

How to discover potential discrimination during the software design phase? (i.e., before having a faulty implementation)

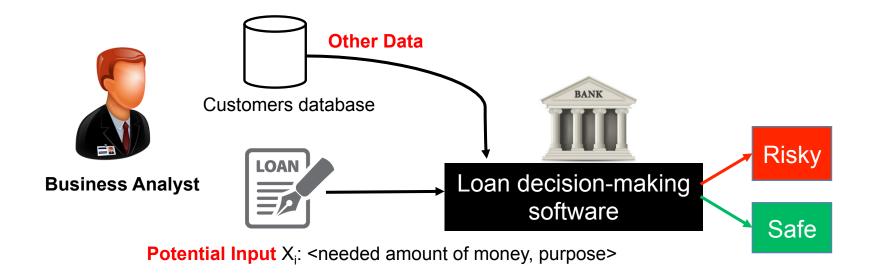
Motivating Example 1



- Bank offers three services
 - Zero-Fee Money Transfer (for international merchants)
 - Vacancies Announcement (for domestic persons educated in accounting).
 - Apply for a loan.
- Policy: The bank disallows discriminating between the loans applicants based on their citizenship.

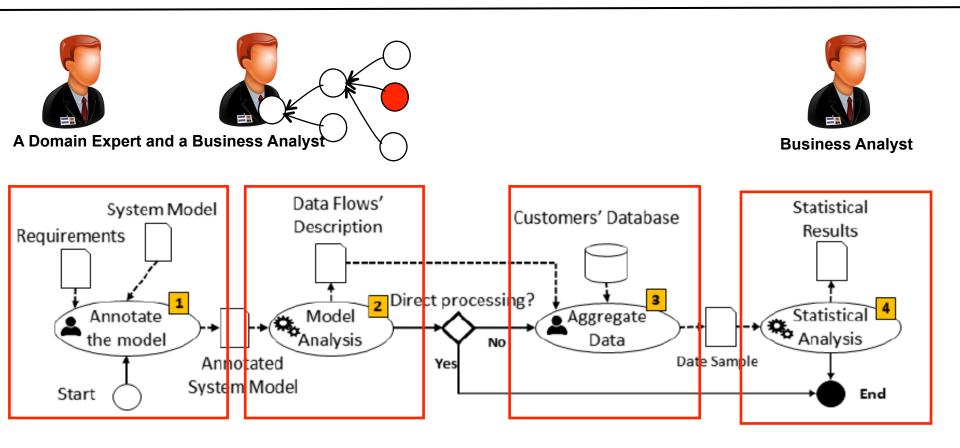


Motivating Example 1



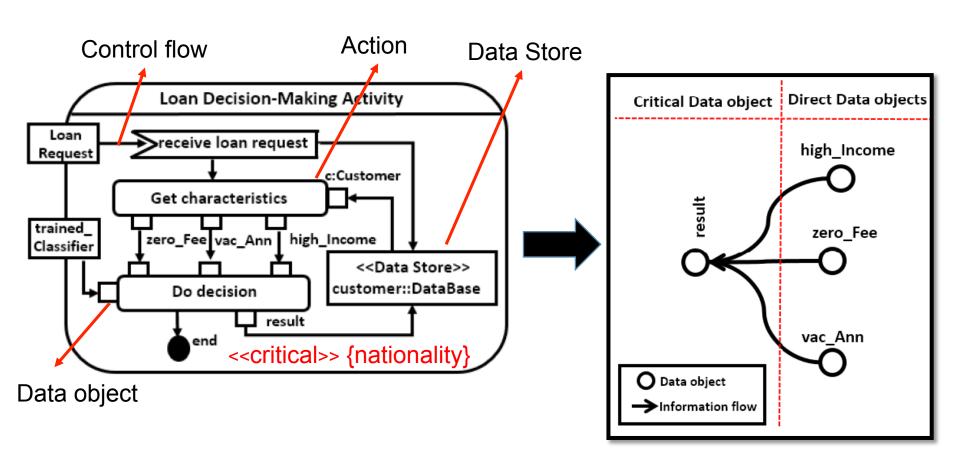
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Roadmap: Model-based Analysis



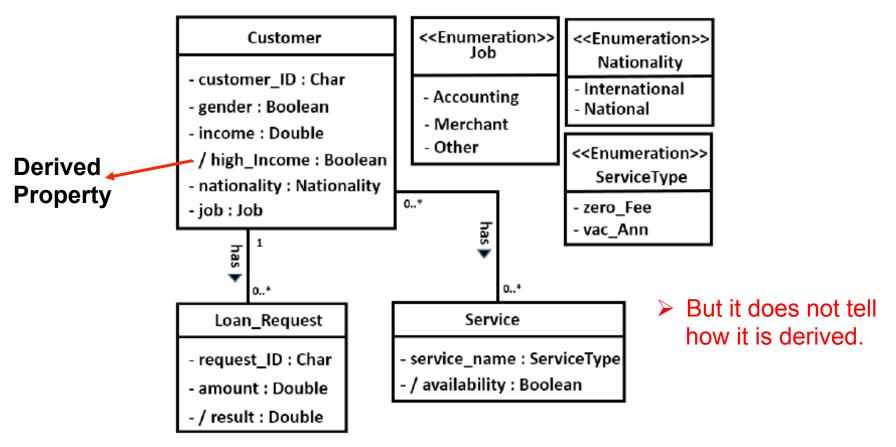
High-level overview of a model-based discrimination analysis framework.

Information Flow Analysis



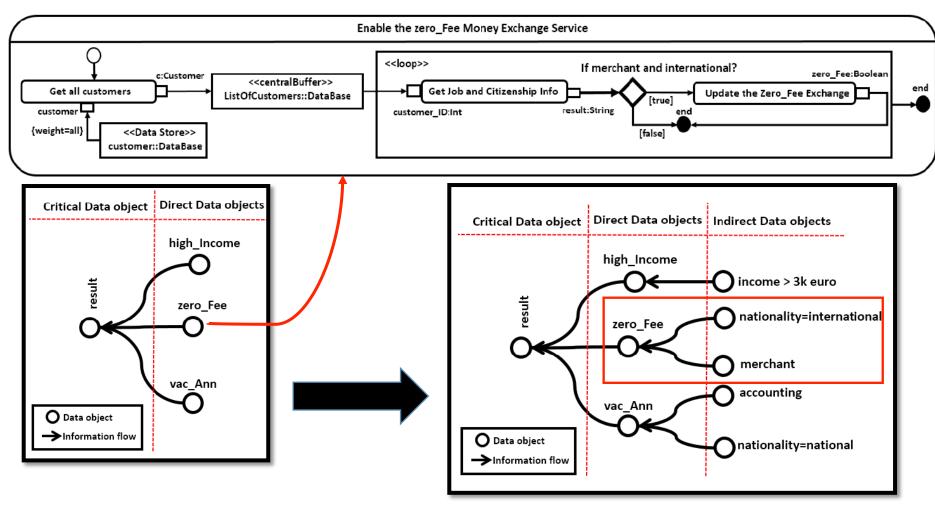
Database Schema

 Information about whether a data object is derived or not can be represented in the database schema.



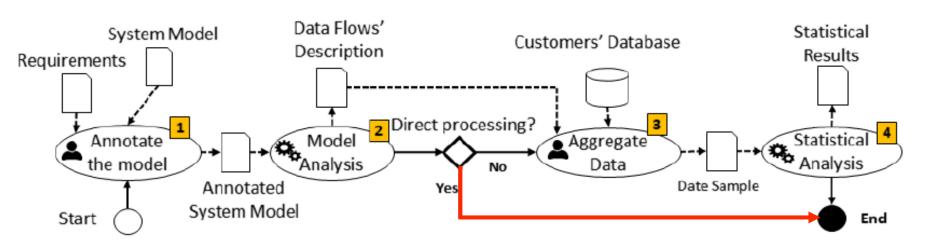
Information Flow Analysis

Activity diagram describing how the value of the zero_Fee data object is derived.



There is indirect leakage of the citizenship data to the result data object.

Roadmap: Model-based Analysis



High-level overview of a model-based discrimination analysis framework.

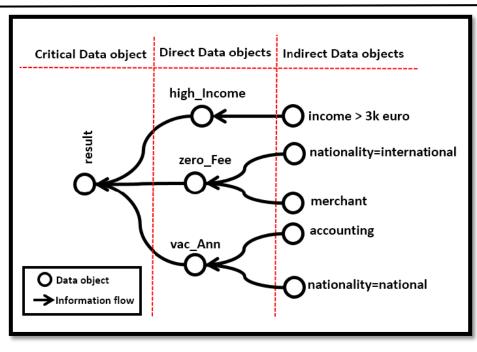
Motivating Example 2



What about dependencies with the gender?

are not true)

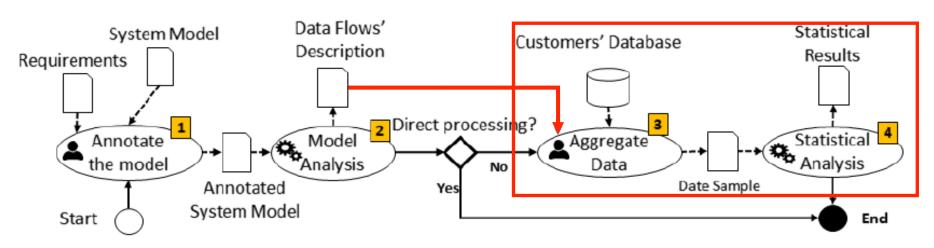
Business Analyst

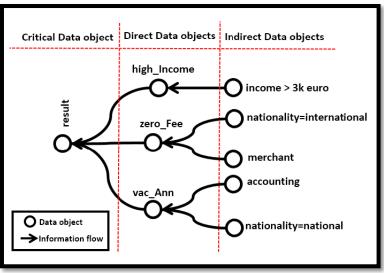


P(Female | ¬(">3k" €∩ international ∩ merchant) ∩ accounting) = 66.67% (i.e., given a national customer with educational background in accounting and the other data objects

The nationality and the education background in this context can act as a proxy for the gender.

Roadmap: Model-based Analysis



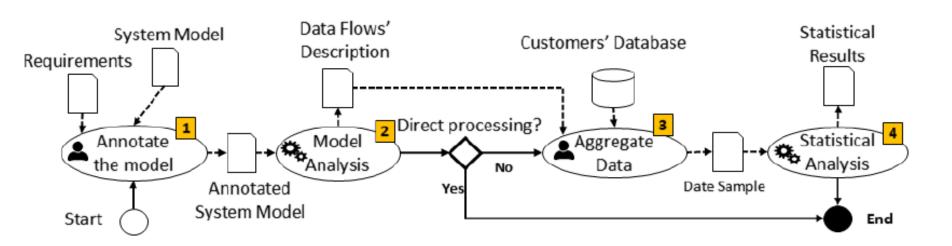


P(Female)	">3k" €	nationality	merchant	accounting
50.00%	1	1	1	1
66.67%	0	1	0	1
0.00%	0	0	1	0
100.00%	0	0	1	1
100.00%	1	1	0	1
0.00%	1	0	1	1

Conclusion



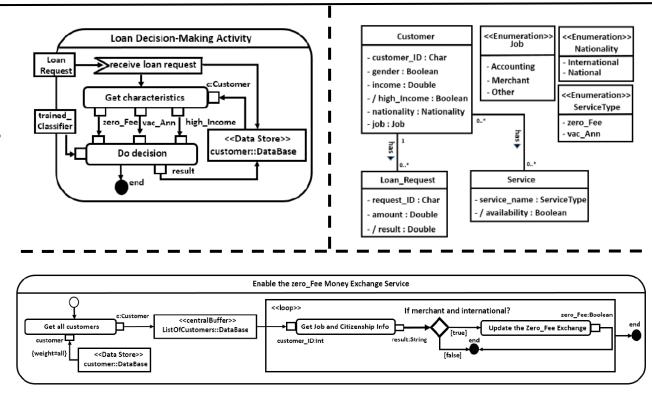




High-level overview of a model-based discrimination analysis framework.

Challenges

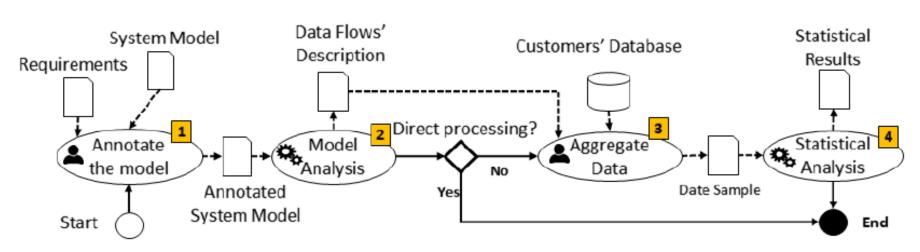
 Information about the derived data are distributed in multiples diagrams.



How to measure the discrimination by proxy? (e.g., information gain)

Conclusion





High-level overview of a model-based discrimination analysis framework.

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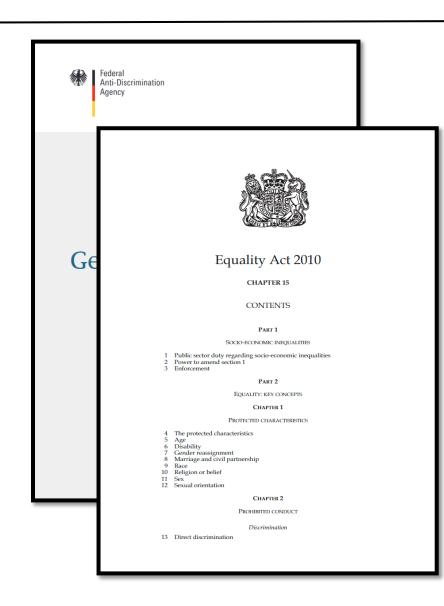
Backup slides

Protected Characteristics

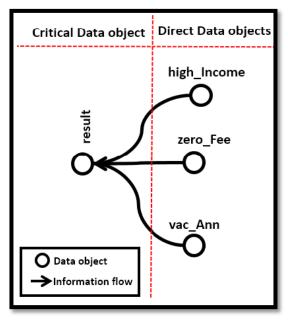
 legally protected characteristics (e.g., age, gender, race, ...).

 But not limited to those listed by the laws and regulations.

Example: a bank may disallow discriminating between the loans applicants based on their citizenship.



Initial Statistical Analysis



P(international | zero_Fee ∩ vac_Ann ∩ high_Income)

P(international)	zero_Fee	vac_Ann	high_Income
66.67%	0	0	0
100.00%	1	0	0
0.00%	1	1	0
100.00%	1	0	1
0.00%	0	1	0
0.00%	0	1	1
100.00%	0	0	1

- ➤ a societal fact (e.g., a taxi driver in Saudi Arabia).
- > They could be **derived** from processing the citizenship information.

Table 1: Personal Data

14210 2.1 0.1001141 2.414						
customer_ID	gender	income	high_Income	nationality	merchant	accounting
BA01	0	3000	0	0	1	1
BA02	0	4500	1	1	1	1
BA03	0	2500	0	1	0	1
BA04	0	3200	1	1	0	1
BA05	0	2900	0	1	0	1
BA06	1	5000	1	0	1	1
BA07	1	2450	0	0	1	0
BA08	1	3600	1	1	1	1
BA09	1	3100	1	0	1	1
BA10	1	1800	0	1	0	1

Table 2. Services Data

customer_ID	zero_Fee	vac_Ann
BA01	0	1
BA02	1	0
BA03	0	0
BA04	0	0
BA05	0	0
BA06	0	1
BA07	0	0
BA08	1	0
BA09	0	1
BA10	1	0