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AI-Enabled Underwriting Brings New Challenges for Insurance

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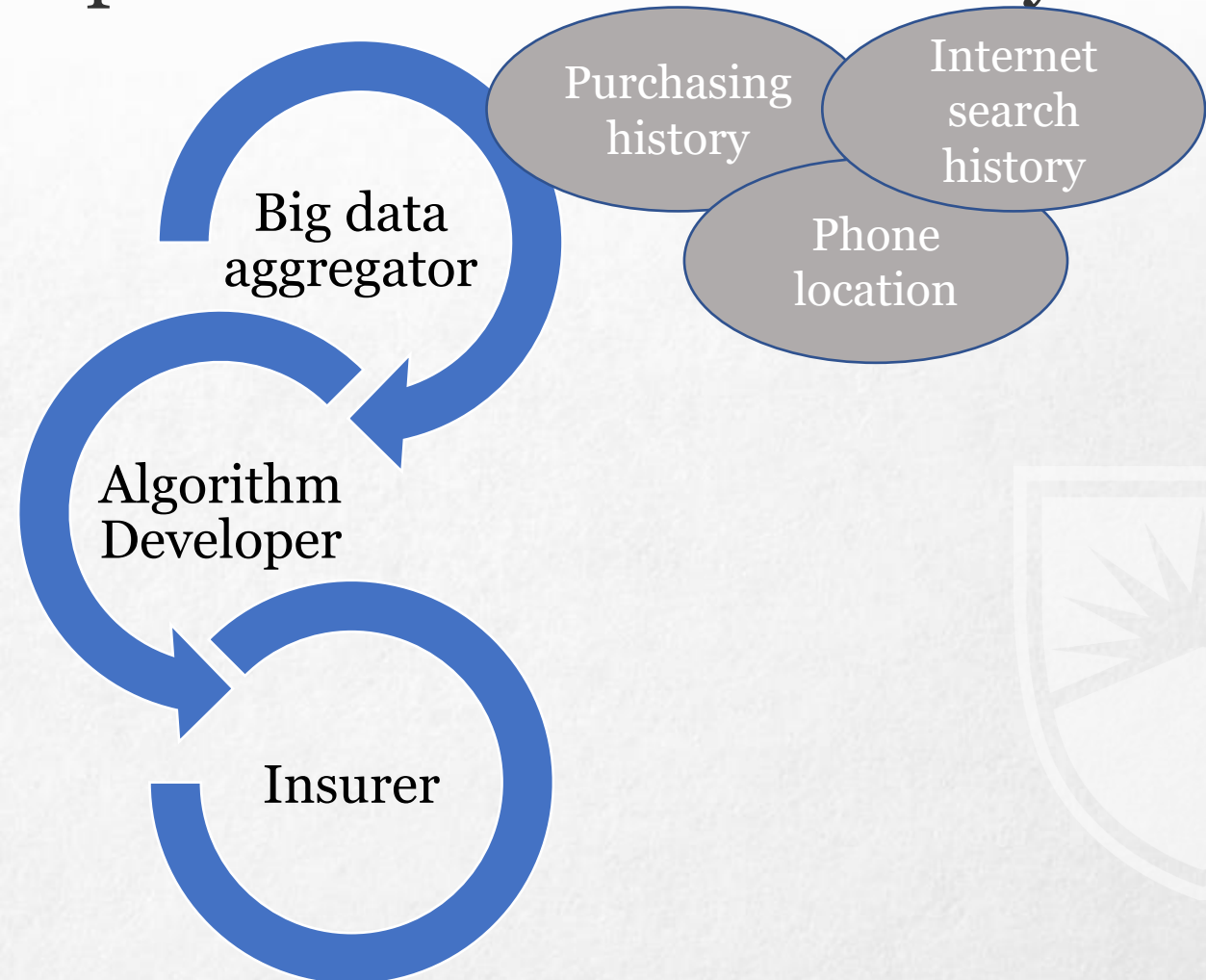
Overview

1. New Challenges, described
2. Regulatory approaches, and their limits
3. A multi-pronged approach
 - Standards; Certification; Testing
4. Discussion/Q&A



New Challenge:

Chain of Data Ownership Obscures Accountability



New Challenge: Three Types of Algorithm Risks & “Mistakes”

Category Mistakes Incorrect label/Misclassified Data	Process Mistakes Inappropriate Design	Social Implications Legal & Accurate, but...
<ul style="list-style-type: none"> - <u>False positives</u>: facial recognition might falsely identify someone as a “smoker” when they are not - <u>False negatives</u>: credit score indicates that you are unlikely to repay debt 	<ul style="list-style-type: none"> - <u>Use of irrelevant data</u>: Use of Social Media data: joining an African-American Breast Cancer Survivor Group - <u>Use of unlawful data</u>: race as a category for insurance decisions - <u>Use of incorrect data</u>: a credit card you never owned 	<ul style="list-style-type: none"> - Ethical implications of outcomes; social impact considerations

Category mistakes and process mistakes with attribution to Martin, K. (2018). Designing Ethical Algorithms. *MIS Quarterly Executive*

AI-enabled Underwriting and Unfair Discrimination

More advanced risks in processing data

Big data source may have programming or data bias in facially neutral factors



Algorithm may correlate neutral factors with protected characteristics

These factors could indirectly represent race in the algorithm

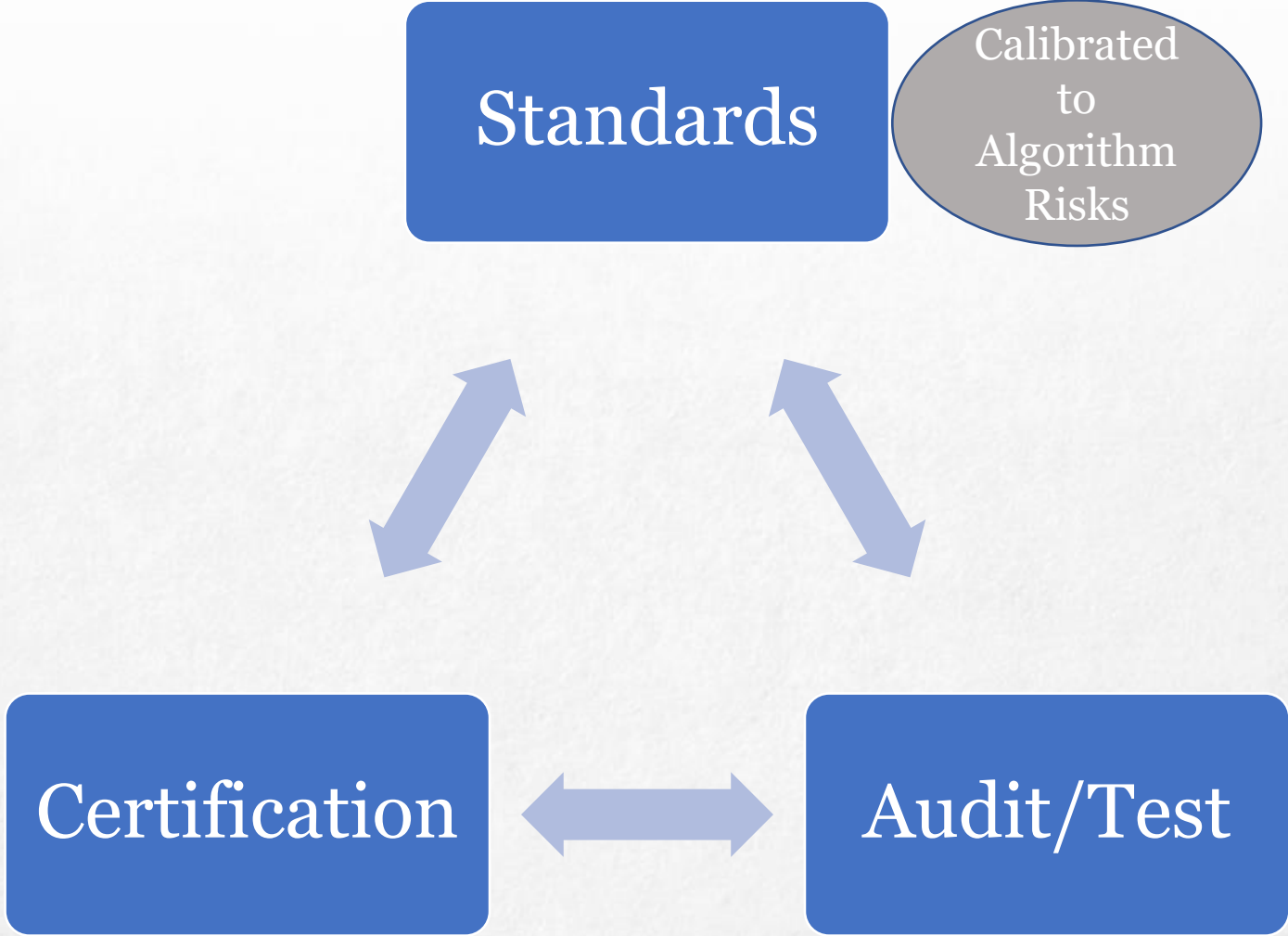
Unfair Discrimination?

Algorithm recommends price or product that could be discriminatory

Regulatory Approaches

Approach	Challenges
<p>Transparency:</p> <ul style="list-style-type: none">➤ Mandate transparency to consumers when an adverse action occurs	<ul style="list-style-type: none">➤ will the insurer know how the algorithm made its decision?➤ the code may be owned by a third party, and proprietary to them
<p>Factor-by-Factor Analysis:</p> <ul style="list-style-type: none">➤ Prohibition of Certain Rating Factors (e.g., criminal arrests, credit scores)	<ul style="list-style-type: none">➤ Big data growth...new rating factors regularly in development➤ facially neutral rating factors could be problematic: stand-in for a prohibited characteristic. E.g., cell phone data could provide zip code information
<p>Blanket Prohibitions:</p> <ul style="list-style-type: none">➤ Prohibition against “unfair discrimination”	<ul style="list-style-type: none">➤ good for “direct” use of data.... But what about the more advanced risks in processing data when race can be “inferred”
<p>Testing:</p> <ul style="list-style-type: none">➤ data should be tested to ensure that it is not “unfairly discriminatory”	<p>Will need to further address:</p> <ul style="list-style-type: none">➤ insurers don’t collect race-based data➤ need a standard for “unfair discrimination”

Multi-pronged approach



Categories of Standards

does it accurately reflect the actual behavior that is of interest?

Accurate

how much does the input contribute to the evaluation of risk?

Actuarial Significance

What decision processes and structures were used in designing the system?
What anticipated offer rates, acceptance rates, etc., among disparate demographic groups?

“Effective” Outcomes





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