

Information Cocoons in Economic Statistics Education: A Study of Teaching Reform Pathways

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Abstract: In the era of digital economy, intelligent algorithms—such as platform recommendations, short video pushes, and search ranking mechanisms—have become primary channels for college students to access information and data. The resulting "information cocoon" effect is increasingly influencing how economic statistics students perceive data and approach statistical reasoning. Specifically, students' data sources are becoming more platform-dependent, visually oriented, and preference-driven, which undermines their ability to collect diverse data and weakens their awareness of sampling principles. Moreover, the filtering mechanisms embedded in recommendation algorithms impair students' capacity to identify key statistical biases—including sampling bias, selection bias, and availability bias—leading to an overreliance on procedural tools, reinforced learning pathways, and a neglect of statistical ethics. To address these challenges, this study proposes a three-stage teaching reform model centered on "cognitive awakening – bias detection – anti-cocoon practice." It introduces a data bias measurement formula and an operational flowchart to guide implementation. The reform is designed to enhance statistical literacy and strengthen data governance competence, ultimately aiming to shift the focus of statistics education from merely using data toward understanding, evaluating, and governing data. By doing so, it seeks to cultivate a new generation of ethically responsible data professionals capable of thriving in an increasingly digital society.

Keywords: Information Cocoons; Economic Statistics Education; Sampling Bias; Data Literacy; Statistical Ethics.

1. Introduction

With the rapid spread of intelligent recommendation algorithms in the information dissemination system, the way college students obtain information is shifting from the past "active search" to the passive "algorithmic feeding". In this structural shift, the data students come into contact with is no longer statistically "overall representative", but is more of a "local sample" generated based on personal interests and preferences. This "feed-solidifying - reinforcing" information filtering mechanism gradually forms an "information cocoon" at the social level and influences students' cognitive patterns and data concepts in the digital learning environment (Pariser, 2011) [1]. For the economics and statistics major, this impact poses a significant professional risk. On the one hand, students overly rely on the data presented by the platform and are prone to mistaking "hot data", "list data", "visualization indicators" with significant selection bias for the real population, thereby reducing their sensitivity to sampling design, data sources, and survey errors; On the other hand, students' statistical analysis often stays at the presentation of "hot phenomena" rather than focusing on core issues such as "whether the statistical inference is reasonable" and "whether the sample is representative", which leads to the gradual alienation of statistical learning into "operational skills training" and "tool display", lacking the necessary scientific

inquiry and critical judgment.

This problem is not an isolated teaching phenomenon, but a concentrated manifestation of the changes in the data structure of the digital society in the educational field. Especially since 2020, due to the pandemic accelerating the online transformation of society, the polarization of public opinion space, and the policy focus at the national level on the governance of platforms and algorithms, the "information cocoon" has gradually shifted from an academic concept to a widely recognized social issue (Baidu Index, Figure 1). This means that economics and statistics students currently entering university campuses are naturally in an environment where information is filtered by algorithms, and when they engage in financial market analysis, industrial research or policy evaluation in the future, they will inevitably face real challenges such as algorithmic bias, data distortion and ethical risks [2, 3]. Therefore, economic statistics education must shift from the traditional "model teaching under ideal data conditions" to "data governance education oriented towards the real world". The core of education lies not only in teaching computational skills, but also in developing students' ability to judge, question and correct biases based on data, to identify structural biases formed by information filtering, to understand the social mechanisms behind the data, and to take responsibility for data justice and governance in future work.

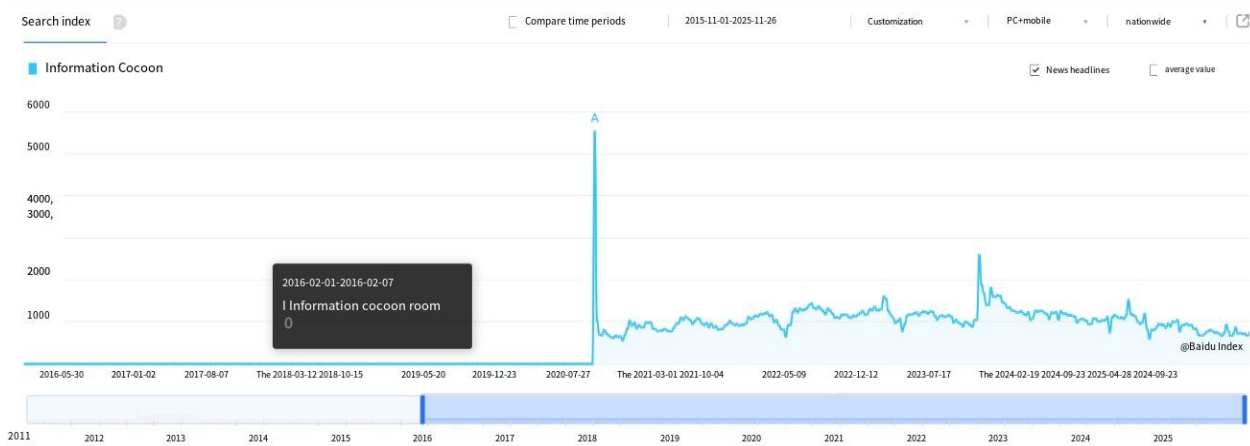


Figure 1. Search graph of Baidu Index for "information cocoon"

Based on the above analysis, this paper constructs a three-step teaching reform model of "cognitive awakening - bias detection - reverse cocoon practice", aiming to incorporate the cultivation of critical data literacy throughout the entire process of economic statistics teaching. Systematically explain the erosion path of the information cocoon on the core statistical competence, and construct a systematic teaching program aimed at empowering students to identify, measure and go beyond the information cocoon, with the expectation of cultivating the next generation of economic statistics talents who can handle complex data environments, possess critical thinking and data ethical responsibility.

2. Implications of the Information Cocoon Phenomenon for Economic Statistics Education

(1) Data sources tend to be single, with platform reliance instead of active acquisition

In course practice, graduation thesis, and data course design, students typically use data automatically filtered by the platform, such as "Douyin business data", "Taobao sales ranking", "Platform Heat Index", "Toutiao Visualization Case", etc. Although these data are visually appealing and technically presentable, they often have a strong screening preference, no sampling mechanism, and no statistical "overall representativeness". In contrast, real public data usually contains features such as "multidimensional variables, sampling background, source institutions, statistical criteria and survey error range", while the data disseminated by platforms often lacks these attributes [4, 5]. Students treat "platform data as real statistics", which prompts statistical education to shift from "survey, sampling, inference, verification" to "imitation visualization". Long-term reliance on platform data weakens students' ability to judge data and prevents them from taking on the responsibility of data

discrimination in a real working environment.

(2) Weakened ability to identify statistical biases, with recommendation mechanisms masking sampling biases

The information cocoon is essentially a kind of "non-probability sampling" and belongs to self-selection sampling. Different users see different data, and the platform provides different sets of information based on differences in interest, showing obvious sampling bias. Therefore, the platform's recommended data often meets the characteristics of "high frequency - small coverage - heavy bias".

However, in the teaching practice research, many students regarded the popular data as "more real and more representative", showing a weakened ability to identify biases. To help students quantitatively identify biases, the following deviation rate measurement formula can be introduced for classroom teaching [6]:

$$D = \frac{|\bar{X}_r - \bar{X}_t|}{\bar{X}_t} \times 100\% \quad (1)$$

Where $\bar{X}_r$ represents the mean of an indicator of the platform's recommended sample (such as the average salary of the recommended position), $\bar{X}_t$ represents the mean of that indicator of the population or authoritative reference sample; D is the quantified deviation rate, the larger the value, the more likely the cognitive distortion caused by the information cocoon is. Through this formula, students can compare differences such as "the recommended price versus the actual market price" and "the recommended job position versus the real industry structure" to understand bias intuitively.

(3) Analytical Perspective path dependence, popular data obscuration important information

The platform algorithm has a "hotspot first" mechanism, resulting in the platform presenting "high-frequency groups" rather than "real groups" [7, 8]. As a result, there are structural biases in student statistical analysis, and typical problems encountered in teaching are shown in Table 1 below.

Table 1. Analysis of Themes Affected by the Information Cocoon

Typical topics	Analysis directions influenced by the information cocoon	A comprehensive statistical perspective that should be advocated
Analysis of Consumption Trends	Live-streaming sales, blind box consumption, internet-famous dining	Basic consumption, service consumption, and consumption structure for different income and age groups
Job Market research	Internet "big companies", emerging flexible jobs	Manufacturing demand, public service jobs, and employment in small and medium-sized enterprises
Research on Enterprise Ecology	Top platform companies, unicorns	"Specialized, refined, distinctive and innovative" small and medium-sized enterprises and regional characteristic industrial clusters

The "algorithm-shaped perspective" of these cases in Table

1 weakens the diversity and exploratory nature of statistical

thinking. If statistical education is not intervened, economic statistics talents will become "data operators in the algorithmic perspective".

(4) Lack of awareness of statistical ethics and neglect of data discrimination and governance responsibilities

There are obvious ethical risks behind the information cocoon - different users receive different data, which may lead to problems such as algorithmic discrimination, price discrimination, and job mismatch. However, students generally understand data only from a technical or economic efficiency perspective and do not examine platform data from the perspective of "fairness, social justice, and information transparency." For example, students will discuss "algorithms

improve sales efficiency", but rarely "whether algorithms exclude vulnerable groups" or "whether data causes cognitive discrimination", which implies an urgent need to introduce the dimension of data ethics education in statistical education. The information cocoon is not only a cognitive issue, but also an ethical one. Algorithms may solidify or even exacerbate inherent biases in society, leading to phenomena such as "big data price discrimination" and "employment discrimination". If statistics education only teaches "how to analyze" without exploring "why to analyze" and "for whom to analyze", students will have difficulty shouldering social responsibilities in data governance in the future.

Table 2. Main Manifestations of the information Cocoon and Its Impact

Dimensions	Specific manifestations	The impact on statistical education
Data Sources	Rely on recommendation data	Downplay sampling awareness and ignore the population
Bias identification	Heat = Truth	It's impossible to judge whether the inference is reasonable
Analysis Perspective	Focus on Hotspots	The conclusion is one-sided and not of public value
Ethical awareness	Ignore discriminatory data	Lack of data governance responsibility

3. The Core Benefits of Shifting to Responsible Data Governance

Returning to the essence of statistics: From 'Computational skills' to 'Critical thinking'. The soul of statistics lies in making robust inferences amid uncertainty. Teaching must guide students through the fog of algorithms, examine the reliability of data sources, and make the leap from "knowing how to calculate" to "knowing how to distinguish".

Rebuilding Data Capabilities: From Passive Acceptance to active Governance. Future statistical talents must have the ability to proactively design data acquisition schemes and cross-validate multi-source data in a mixed real data environment, becoming data masters rather than passive consumers.

Building an ethical defense line: From 'value neutrality' to

'accountability'. Statistical professionals are important builders of the data society. Education must be instilled with an ethical gene, enabling it to keenly identify data bias and actively advocate and practice fair, transparent, and good data governance principles in future work.

4. Teaching Reform Plan: Three-step anti-cocooning Statistical Teaching Model

Based on the challenges identified, this study proposes a three-stage pedagogical model. The framework aims to systematically deconstruct information cocoons in economic statistics education. It progresses through three structured phases, as detailed in Table 3. This approach transforms students from passive data recipients into critical data governance practitioners.

Table 3. Teaching Reform Path

Stages	Core Objectives	Teaching activities	Anticipatory ability
Cognitive arousal	Recognizing data is selected	Show different recommendation pages	See the information cocoon
Deviation detection	Learn to quantify bias	Group sampling + deviation rate calculation	Proving "bias"
Anti-cocoon practice	Improving data acquisition	Multi-source sampling + data ethics norms	Be able to propose governance suggestions

(1) Cognitive awakening stage - Revealing the "cocoon"

Core activity: Comparative experiments. Display in the classroom the recommendations about "consumption upgrade" or "employment prospects" received by different user profiles (such as "Generation Z in first-tier cities" and "middle-aged people in third - and fourth-tier cities") on the same platform (such as Taobao, Toutiao). Guide students to discuss: "Is the world you see the whole picture of the real world or a 'theater' tailored for you by algorithms?" This session aims to stimulate students' metacognition, making them aware that they may be trapped in an information cocoon and that "it's not us who choose the data, but the data that chooses us."

(2) Deviation detection stage from "perception" to "measurement"

Core activities: Group sampling and deviation calculation. Group the students and crawl the recommended data of a

certain economic indicator (such as "average salary for junior positions") from the platform interface (such as the public API), while obtaining the corresponding overall data from official channels such as the National Bureau of Statistics and the Ministry of Human Resources and Social Security. Guide students to use the aforementioned deviation rate formula (1) for quantitative calculation, supplemented by group box plots or violin plots for visual comparison. This process transforms the abstract concept of "deviation" into concrete, verifiable numerical evidence, reinforcing students' empirical spirit. Next, students are asked to explain the reasons behind the bias: Is it due to user behavior, algorithmic preferences, platform interests, sample limitations? Develop students' ability to question the data - explain the bias - trace the source.

(3) Counter-cocooning Practice - Proposing governance solutions and data ethics evaluation

This section aims to cultivate students to be "data

responsibility proposers" rather than "data users", with the core activity being the design of "data governance solutions". Students are required to complete a comprehensive task in the capacity of a "data governance advisor" based on the findings from the previous two stages. Guide students to propose counter-cocooning strategies such as multi-platform sampling, weighted fusion of data from different sources, introducing samples of vulnerable groups, and suggestions for data transparency.

The proposal should include the following three steps.

(1) Multi-source data sampling plan: How to integrate platform data, official data, survey data, and even "dark data" to form a more comprehensive view.

(2) Algorithmic Bias audit recommendations: Design a simple process for a fictional recruitment platform or credit platform to evaluate the fairness of its recommendation algorithm.

(3) Data Ethics Charter: Draft a simple code of ethics for data scientists, emphasizing the responsibility to be vigilant about and correct the information cocoon in statistical analysis.

The final output works include an algorithm improvement proposal, a fair sampling scheme, a visual report on information discrimination, and an anti-bias data collection specification, thus transitioning from technical practice to the expression of social responsibility and filling the ethical gap in statistical education. This session aims to develop students' ability to go beyond criticism and propose constructive solutions, completing the initial role transition from "data user" to "data manager".

5. Conclusions and Prospects

The information cocoon is a deep chasm in front of statistics education in the digital age. It challenges not only students' habits of information acquisition, but also the objectivity, systematicness and inferential validity of the foundation on which statistics thrives. The "three-step reverse cocoon teaching model" proposed in this paper is a reform path aimed at empowering students, returning to the essence, and oriented towards governance. It integrates the cultivation of critical data literacy throughout the teaching process, enabling students to master statistical tools while gaining insight into the premises and limitations behind the tools.

In the future, the deepening of this reform could be carried

out in the following directions: First, develop a standardized statistical bias identification ability evaluation scale to scientifically assess teaching effectiveness; Second, build an interdisciplinary data ethics case library and collaborate with fields such as law, sociology, and computer science in teaching; Third, explore cooperation with Internet companies to establish "data transparency" teaching laboratories, allowing students to interact with algorithms in a real and controllable environment. Ultimately, the mission of economic statistics education is to cultivate outstanding talents who can penetrate the fog of data, reveal the truth of the economy, and have an ethical warmth.

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