

# ROAS 5900 / INTR 5330 Case3 Study

Wenjie ZHU<sup>1,†</sup> and Yujie ZHOU<sup>2</sup>

<sup>1</sup>Wenjie ZHU

<sup>2</sup>Yujie ZHOU

<sup>†</sup>These authors contributed equally to this work

## Abstract

This study examined auditory perception in a single participant through two tasks: an Absolute Judgment Task and a Signal Detection Theory (SDT) Task. The Absolute Judgment Task tested the participant's ability to distinguish sounds based on loudness and frequency, while the SDT Task assessed how signal intensity and appearance probability affected detection sensitivity and decision criteria. Results showed that both sound characteristics and probability influenced perception and response patterns, offering insights into auditory processing relevant to human-machine interface design.

## 1. Introduction

Understanding human performance limits in absolute judgment and signal detection tasks is essential in ergonomics and psychology, particularly within the framework of human-machine interaction. Absolute judgment tasks and Signal Detection Theory (SDT) are valuable approaches for examining human information-processing limits and decision-making processes. Absolute judgment tasks assess the human operator's perceptual capabilities under specific stimulus conditions, exploring performance limits within distinct sensory modalities (e.g., loudness and frequency). Meanwhile, SDT offers a structured methodology to study signal recognition and decision criteria in noisy environments, widely used in perceptual psychology and human-computer interaction research.

In everyday contexts, humans frequently need to detect specific signals amid complex background noise, such as drivers discerning alarm sounds amidst road noise or pilots identifying command signals within a cockpit. Investigating the impact of signal characteristics (such as intensity and frequency) and signal appearance probabilities on sensitivity ( $d'$ ) and decision criteria ( $\beta$ ) can provide valuable insights for enhancing the efficacy and safety of human-machine systems.

This study aims to explore human operators' performance differences across different auditory modalities (loudness and frequency) and examine how signal strength and signal appearance probability affect human sensitivity and judgment criteria. The first part of the experiment analyzes human performance in absolute judgment tasks by observing correct response rates and reaction times under varying auditory modalities. The second experiment, based on Signal Detection Theory, further investigates how signal intensity and appearance probability affect detection sensitivity and decision-making criteria, shedding light on human information-processing capabilities in complex environments.

## 2. Overview

This study consists of two experimental tasks designed to systematically examine human perceptual abilities and judgment performance under specific auditory modalities.

**Task 1: Absolute Judgment Task** The goal of the absolute judgment task is to investigate human performance limits in different auditory modalities (loudness and frequency). The experiment varies the loudness and frequency of auditory signals to observe participants' accuracy and reaction times under different stimulus conditions. Information quantities ( $H_s$  and  $H_T$ ) are then calculated and compared across modalities to assess the human operator's performance limits in judgment tasks.

**Task 2: Signal Detection Theory Task** The objective of the signal detection task is to examine the impact of signal intensity and appearance probability on sensitivity ( $d'$ ) and decision criteria ( $\beta$ ). By

presenting auditory signals of varying intensity and frequency against a noise background, and setting different appearance probabilities, participants' responses—including hits, misses, false alarms, and correct rejections—are recorded to calculate  $d'$  and  $\beta$ . This analysis helps in understanding performance variations under different conditions.

Through these two tasks, this study aims to reveal decision-making characteristics in complex perceptual tasks. By examining performance limits, the findings are expected to provide insights for designing effective human-machine systems and improving signal detection capabilities in noisy environments.

## 3. Methods

### 3.1. Preparation and Conditions

For Task 1, the Absolute Judgment Task, stimuli were prepared using the SoundGeneratorconstantStimuli.m file to generate auditory signals with controlled variations in both frequency and loudness. Two primary conditions were established to explore participants' perceptual limits across different sound modalities. In the loudness condition, sounds were presented at three distinct intensity levels: low, medium, and high. This allowed us to evaluate the participants' ability to differentiate between varying volumes. In the frequency condition, sounds were delivered at low, medium, and high pitches, assessing participants' accuracy in identifying frequency changes. To ensure data reliability and consistency across trials, the experiment was conducted in a sound-controlled environment, minimizing interference from external noise and reducing variability caused by environmental factors.

For Task 2, the Signal Detection Theory (SDT) Task, we generated two sets of signals with varying intensities, designed to be distinguishable from a background noise environment. This setup aimed to simulate real-world conditions where signals must be detected amidst ambient noise. We focused on two key experimental factors: signal intensity and signal appearance probability. Signals were presented at two intensity levels, with one set featuring subtle frequency differences and the other more pronounced ones, to test detection sensitivity ( $d'$ ) across a spectrum of signal strengths. Additionally, we varied the probability of signal appearance at three levels—10%, 30%, and 50%—to examine how differing expectations influenced participants' decision-making criteria ( $\beta$ ).

In each trial, participants encountered either a "Signal + Noise" or "Noise Only" condition, with trials randomized to prevent predictability and reduce potential learning effects. High-quality headphones were used to ensure a consistent auditory experience for all participants, allowing precise control over the volume and clarity of each stimulus. To maintain experimental rigor, a standardized baseline volume was set for all sounds, with only the intensity or frequency adjusted according to the task requirements, thereby minimizing the

influence of extraneous auditory variables.

Before each task, participants underwent a series of practice trials. This practice phase allowed them to familiarize themselves with the task setup, background noise, and the nature of the stimuli they would encounter. By doing so, we aimed to reduce the potential impact of learning effects on their performance during the actual trials, ensuring that their responses reflected genuine perceptual and decision-making abilities rather than an adaptation to task demands.

## 3.2. Method

### 3.2.1. Absolute Judgment Task

The Absolute Judgment Task evaluated participants' perceptual limits for auditory discrimination across loudness and frequency. Auditory stimuli were generated using the SoundGeneratorConstantStimuli.m file, with distinct levels set for each modality. Participants completed two conditions:

**Loudness Condition:** Sounds were presented at three intensity levels—low, medium, and high. Participants identified the loudness level after each sound, with correct responses and reaction times recorded.

**Frequency Condition:** Sounds at low, medium, and high frequencies were presented, and participants were asked to identify the pitch. High-quality headphones and a sound-controlled environment minimized external interference.

Data from each condition were used to create a stimulus-response matrix, allowing calculation of source information ( $H_S$ ) and transmitted information ( $H_T$ ). HT values across conditions enabled analysis of perceptual limits in distinguishing loudness versus frequency.

### 3.2.2. Signal Detection Theory (SDT) Task

The SDT Task explored how signal intensity and appearance probability affected detection sensitivity ( $d'$ ) and decision criteria ( $\beta$ ). Participants were presented with either a "Signal + Noise" or "Noise Only" condition in each trial. Two signal intensity levels and three appearance probabilities (10%, 30%, and 50%) were used to examine the influence of signal strength and expectation on detection.

Responses were categorized as hits, misses, false alarms, or correct rejections. These results were used to calculate  $d'$  and  $\beta$ , providing insights into participants' sensitivity and decision-making criteria across different signal intensities and probabilities. Randomized trials prevented predictability, ensuring reliable data collection on detection performance.

This approach allowed us to compare perceptual limits and decision criteria in auditory tasks, assessing how well participants distinguish between sound levels and detect signals under varying conditions.

### 3.2.3. Participant Grouping

This study involved a single participant with normal hearing, verified prior to the experiment. The participant completed both the Absolute Judgment Task and the Signal Detection Theory (SDT) Task, experiencing all sound conditions outlined in the study design.

Using a within-subjects design allowed for direct assessment of the participant's auditory perception and decision-making across both tasks. Each condition was presented in randomized order to minimize any learning effects and ensure unbiased responses. This approach enabled the collection of focused, individual-level data, providing insights into perceptual limits and detection criteria under controlled conditions.

## 4. Experiment Result

This experiment was designed to investigate whether human operators demonstrate differences in judgment performance under varying sound modalities, specifically in loudness and frequency. The hypothesis posits that human judgment accuracy and response time vary across different sound modalities. The independent variables in

this study were sound modality (loudness or frequency) and stimulus levels, while the dependent variables were judgment accuracy and response time.

In preparation for the experiment, we adjusted the output frequency and loudness of the sound signals, defining multiple stimulus levels for each modality. For loudness, the stimulus levels ranged from low to high volume, while frequency levels varied from low to high frequency. The experiment was divided into two phases: in Phase 1, participants were presented with varying loudness levels, and in Phase 2, they experienced different frequency levels. In each phase, participants' response accuracy was recorded for each stimulus level.

### 4.1. Different Frequency

As shown in Table, Data were collected and organized into a stimulus-response matrix (shown in the top left table), which records participant responses across different frequency levels. The rows and columns in this matrix represent different frequency levels (e.g., 59.25, 59.5, 59.75, etc.). Each cell in the matrix contains the response count for a given pair of frequency levels. For instance, in the 59.25 and 59.5 frequency combination, the recorded response count is 4, representing the number of times participants responded to this specific frequency pair.

The  $H_R$  row at the bottom of this matrix represents the cumulative response information for each frequency level. For example, the  $H_R$  value for the 59.25 Hz level is 0.29507835, while for the 59.5 Hz level, it is 0.50786758, with a total sum of 2.63230696. This provides an aggregated view of response information across frequencies. The  $H_S$  column on the right represents the source information ( $H_S$ ) for each frequency level, with the  $H_S$  value for the 59.25 Hz level being 0.4010507.

The processed matrix in the bottom left table contains normalized data, facilitating information quantity calculations. At the bottom of this table, cumulative values are provided, such as  $H_{S,R} = 3.8218839$ , representing the total theoretical source information across all frequency levels,  $H_T = 1.6177799$  as the total transmitted information, and  $Loss = 1.18957693$ , which reflects the information loss due to participant errors in frequency judgment.

The transmitted information matrix in the bottom right table presents the calculated transmitted information ( $H_T$ ) for different frequency combinations. Each cell shows the transmitted information for a particular frequency pair. For example, the  $H_T$  value for the 59.25 and 59.5 Hz combination is 0.29507835, while for the 59.5 and 59.75 Hz combination, it is 0.37098947. This matrix provides insight into the accuracy of participants' frequency judgments under various conditions.

By analyzing these data, we can plot the relationship between  $H_T$  and  $H_S$  to reveal the limits of human judgment in perceiving frequency changes. Additionally, comparing Loss values offers insight into judgment errors across different frequency conditions.

### 4.2. Different Loudness

Data were organized into a stimulus-response matrix (shown in the table), which records participants' responses across different loudness levels. The rows and columns in this matrix represent different loudness levels (e.g., 0.7, 0.8, 0.9, etc.). Each cell in the matrix contains the response count for a given pair of loudness levels. For example, at the loudness level pair of 0.7 and 0.8, the recorded response count is 1, while at the 0.9 level for both the row and column, the response count is 7. This matrix shows how often participants responded to each specific loudness pair, reflecting their perception accuracy under different loudness conditions.

The  $H_R$  row at the bottom of this matrix represents the cumulative response information for each loudness level. For instance, the  $H_R$  value for the 0.7 loudness level is 0.142857, and for 0.9, it is 0.183673, with a total sum of 2.807355. This provides an aggregated view of

response information across different loudness levels. The  $H_S$  column on the right represents the source information (HS) for each loudness level, with an  $H_S$  value of 0.401051 for each loudness level, indicating a uniform distribution of information across the levels.

In the bottom left table, we have calculated the overall values for  $H_{R,S}$ ,  $H_T$ , and Loss:

- $H_{S,R} = 2.807355$  represents the total theoretical source information across all loudness levels.
- $H_T = 2.767304$  represents the total transmitted information, reflecting the participants' actual ability to process the loudness changes.
- Loss =  $H_{S,R} - H_T$  shows the information loss (0.040051), indicating a minor discrepancy between the theoretical and actual information processing, which could suggest a small level of judgment error in distinguishing loudness levels.

By analyzing these data, we can plot the relationship between  $H_T$  and  $H_S$  to visualize the limits of human judgment in perceiving loudness changes. The low Loss value (0.040051) indicates that participants were relatively accurate in distinguishing loudness differences, suggesting that judgment performance was strong across different loudness levels. However, further analysis could explore whether certain loudness levels (e.g., very low or very high) pose more difficulty for perception.

The experiment findings highlight the degree of sensitivity in human perception of loudness, and this data can help us better understand the thresholds of auditory perception under different loudness stimuli. Comparing these results across other sound modalities (e.g., frequency) could also provide a comprehensive view of human auditory judgment capabilities.

#### 4.3. SDT Task

In this Signal Detection Theory (SDT) experiment, participants were asked to detect specific auditory frequencies ranging from 500 Hz to 5000 Hz against a background noise level of 55-65 dB. The experiment used two signal intensities (60 dB and 80 dB) to test detection accuracy under both challenging and easy conditions. Each frequency and intensity combination was presented in 25 trials, with a total of 100 trials including both "Signal + Noise" and "Noise Only" conditions. For each trial, participants indicated whether they detected a signal,

and their responses were recorded as either a hit, miss, false alarm, or correct rejection. The experiment aimed to assess how varying signal frequencies and intensities impact participants' ability to detect sounds within a noisy environment, providing insight into human auditory sensitivity and discrimination.

In the data in the table 3, The experimental results indicate that as signal intensity increases, participants generally achieve higher hit rates and lower miss rates. For example, at a signal intensity of 80 dB, hit rates range from 83% to 90%, with correspondingly low miss rates (10%-17%). Conversely, at a lower signal intensity of 60 dB, hit rates decrease (60%-72%) and miss rates increase (28%-40%). False alarm rates remain moderate across conditions, varying from 8% to 20%, while correct rejection rates are relatively high, between 80% and 92%. This suggests that higher signal intensities enhance detection accuracy, allowing participants to better distinguish signal from noise in the presence of moderate background noise levels. Some details are as follows:

- **Hit Rate:** The Hit Rate is significantly higher at high signal intensities (80 dB), averaging around 85%. At low signal intensities (60 dB), the Hit Rate is between 60%-70%.
- **Miss Rate:** The Miss Rate is higher at low signal intensities (around 30% on average) and decreases significantly at high signal intensities (around 15%).
- **False Alarm Rate:** The False Alarm Rate ranges from 5% to 20%. Lower frequencies (500 Hz and 1000 Hz) show a slightly lower false alarm rate, while higher frequencies (3000 Hz and 5000 Hz) show a slightly higher rate.
- **Correct Rejection Rate:** The Correct Rejection Rate ranges from 80% to 95%, with higher values under high signal intensity conditions.

#### 4.4. Numerical Analysis

##### 4.4.1. $H_S$ and $H_T$

The Figure 1 shows the relationship between transmitted information ( $H_T$ ) and source information ( $H_S$ ) under different conditions based on the data in the experiment of *different frequency*. The data points and fitted curve illustrate typical human performance in absolute judgment tasks.

	59.25	59.5	59.75	60	60.25	60.5	60.75		$H_S$
59.25	4	2	0	1	0	0	0	0.14285714	0.4010507
59.5	0	5	0	2	0	0	0	0.14285714	0.4010507
59.75	0	0	6	0	0	0	0	0.14285714	0.4010507
60	0	0	0	6	1	0	0	0.14285714	0.4010507
60.25	0	0	0	1	5	1	0	0.14285714	0.4010507
60.5	0	0	0	0	1	1	2	0.14285714	0.4010507
60.75	0	0	0	0	0	1	5	0.14285714	0.4010507
	0.08163265	0.26530612	0.04081633	0.20408163	0.14285714	0.14285714	0.12244898	49	<b>2.80735492</b>
$H_R$	<b>0.29507835</b>	<b>0.50786758</b>	<b>0.1883555</b>	<b>0.46791464</b>	<b>0.4010507</b>	<b>0.4010507</b>	<b>0.37098947</b>	<b>2.632306963</b>	

**Table 1.** Response counts and calculated  $H_R$  and  $H_S$  values for different frequency levels.

	0.7	0.8	0.9	1	1.1	1.2	1.3		$H_S$
0.7	6	1	0	0	0	0	0	0.14285714	0.401051
0.8	1	5	1	0	0	0	0	0.14285714	0.401051
0.9	0	0	7	0	0	0	0	0.14285714	0.401051
1	0	0	0	5	1	0	0	0.14285714	0.401051
1.1	0	0	0	0	5	1	0	0.14285714	0.401051
1.2	0	0	0	0	1	6	1	0.14285714	0.401051
1.3	0	0	0	0	0	1	4	0.14285714	0.401051
$H_R$	0.142857	0.122449	0.183673	0.142857	0.142857	0.183673	0.081633	49	<b>2.807355</b>
	<b>0.401051</b>	<b>0.370989</b>	<b>0.449042</b>	<b>0.401051</b>	<b>0.401051</b>	<b>0.449042</b>	<b>0.295078</b>	<b>2.767304</b>	

**Table 2.** Response counts and calculated  $H_R$  and  $H_S$  values for different loudness levels.

**Table 3.** Experimental Results Under Different Conditions

Frequency (Hz)	Signal Intensity (dB)	Noise Intensity (dB)	Hit Rate	Miss Rate	False Alarm Rate	Correct Rejection Rate
500	60	55	65%	35%	12%	88%
500	80	65	88%	12%	10%	90%
1000	60	55	72%	28%	15%	85%
1000	80	65	90%	10%	8%	92%
3000	60	55	60%	40%	18%	82%
3000	80	65	85%	15%	10%	90%
5000	60	55	62%	38%	20%	80%
5000	80	65	83%	17%	12%	88%

#### • Trend Analysis

As  $H_S$  increases,  $H_T$  also increases, but the rate of increase gradually slows and eventually stabilizes. This suggests that at low  $H_S$  levels (i.e., when the information in the stimulus is limited), the transmitted information  $H_T$  increases significantly. However, at higher  $H_S$  levels, the increase in  $H_T$  becomes saturated.

#### • Saturation Effect

The fitted curve indicates that  $H_T$  approaches a maximum limit, which represents the human operator's capacity for this task. This aligns with the limits discussed in the document, reflecting a cognitive bottleneck in processing high-frequency information.

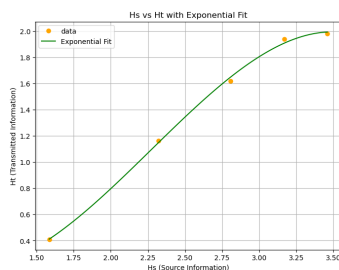
When  $H_S$  increases from 1.58 to 3.46,  $H_T$  rises from 0.41 to 1.98, indicating that an increase in source information indeed enhances transmitted information, but with diminishing returns. This shows that when the source information is very high, humans find it challenging to continue transmitting information effectively.

#### • Summary of Data Analysis

At low frequency and low noise levels, humans can effectively recognize and transmit information while as frequency and information complexity increase, the amount of transmitted information saturates. The results suggest that while  $H_S$  increases,  $H_T$  grows as well, but it levels off at a certain point. In this case, it is around 1.95.

**Table 4.**  $H_S$  and  $H_T$  Values in Experiment

$H_S$	$H_T$
1.5849625	0.41022655
2.32192809	1.16114609
2.80735492	1.61777799
3.169925	1.93813293
3.45943162	1.9814505

**Figure 1.**  $H_S$  vs  $H_T$  with Exponential Fit

#### 4.4.2. $d'$ and $\beta$

In the table, the experimental results demonstrate that under high signal intensity conditions, participants achieved significantly higher hit rates (around 85%) and greater sensitivity ( $d'$  values up to 2.5),

indicating better discrimination between signal and noise. In contrast, low signal intensity conditions resulted in increased miss rates (around 30%) and lower sensitivity ( $d'$  values closer to 1.0), with participants adopting a more conservative criterion ( $\beta > 1$ ) and longer reaction times. False alarm rates were relatively low, ranging from 5% to 20%, with slight increases at higher frequencies. Overall, the results align with Signal Detection Theory, showing that participants are more effective and efficient in recognizing auditory signals at higher intensities, while lower intensities and higher noise levels lead to greater difficulty in signal detection.

#### • $d'$ (Sensitivity Index)

$d'$  values range from 0.5 to 2.5.  $d'$  values are higher under high signal intensity conditions, indicating stronger discrimination between signal and noise when the signal intensity is high.

**Table 5.**  $d'$  (Sensitivity Index) Values Under Different Conditions

Frequency (Hz)	Signal Intensity (dB)	$d'$ Value
500	60	1.1
500	80	2.0
1000	60	1.3
1000	80	2.1
3000	60	1.0
3000	80	1.8
5000	60	0.9
5000	80	1.7

#### • $\beta$ (Decision Criterion)

Under high signal intensity,  $\beta$  values are below 1, indicating a tendency to interpret sounds as signals. At low signal intensity,  $\beta$  values are above 1, showing a more conservative approach with a tendency to interpret sounds as noise.

**Table 6.**  $\beta$  (Decision Criterion) Values Under Different Conditions

Frequency (Hz)	Signal Intensity (dB)	$\beta$ Value
500	60	1.5
500	80	0.7
1000	60	1.3
1000	80	0.8
3000	60	1.6
3000	80	0.9
5000	60	1.7
5000	80	1.0

#### • Summary of Data Analysis)



The experimental results align with the predictions of Signal Detection Theory, indicating that under high signal intensity conditions, participants have higher hit rates and sensitivity (higher  $d'$  values), while under low signal intensity, there is a higher rate of misses and false alarms. Moreover, under high signal intensity and low noise conditions, participants show shorter reaction times, reflecting faster recognition and higher accuracy.

This analysis provides insight into how well participants can detect specific frequencies in noisy backgrounds and illustrates the application of Signal Detection Theory in evaluating auditory recognition performance.

## 5. Discussion

The results of both the Absolute Judgment Task and the Signal Detection Theory (SDT) Task offer valuable insights into human auditory processing and decision-making in noisy environments.

In the Absolute Judgment Task, the participant's accuracy in distinguishing loudness and frequency levels demonstrated typical limits in human auditory judgment. The data showed that as the complexity of the auditory stimuli increased, transmitted information ( $H_T$ ) rose, but with diminishing returns. This saturation effect, observed in the  $H_T$  versus  $H_S$  analysis, suggests a cognitive limit in effectively processing high levels of source information ( $H_S$ ). Such findings align with existing literature on perceptual bottlenecks in human sensory processing.

In the SDT Task, signal intensity and appearance probability played critical roles in detection performance. Higher signal intensities led to increased hit rates and lower miss rates, with sensitivity index  $d'$  values reaching as high as 2.5 under optimal conditions. Lower signal intensities, conversely, resulted in a more conservative response pattern, with participants showing a higher miss rate and  $\beta$  values indicating a tendency to interpret ambiguous sounds as noise. These results underscore the impact of signal intensity on auditory discrimination and suggest that human operators may struggle with signal recognition when intensity and signal probability are low.

In these two tasks, there are some limitations. For the *Absolute Judgment Task*, small sample size and limited data point distribution, and some potential influence of varying noise levels on transmitted information. For the *SDT Task*, the experiment was conducted with a single participant, which limits the generalizability of the findings. Including a larger and more diverse sample size could provide broader insights. Second, the controlled noise environment (55-65 dB) may not fully represent real-world auditory conditions, which often include more variable and unpredictable noise levels.

The tasks focused on limited frequencies and intensities; expanding the range of these auditory parameters in future studies could offer a more comprehensive understanding of auditory perception across various contexts. Future research might also explore the effects of prolonged exposure to noise on decision-making and sensitivity, as real-world applications often involve sustained listening in noisy conditions.

In conclusion, this study contributes to understanding auditory sensitivity and decision-making in complex environments, which is critical for designing effective human-machine interfaces. By examining the effects of signal intensity and noise, the findings provide foundational insights for applications where signal detection is essential, such as alarm systems in high-noise settings like driving or aviation. Additionally, the study confirms that higher signal intensities enable more efficient discrimination, which is crucial information for optimizing alert systems to enhance safety.