

# Whisper ASR Evaluation Report

WPM & Background Noise Impact Analysis

Generated: December 09, 2025

40 Recordings | Single Speaker | Controlled Variables

*Analysis by Claude (Anthropic AI)*

# Executive Summary

## KEY FINDINGS

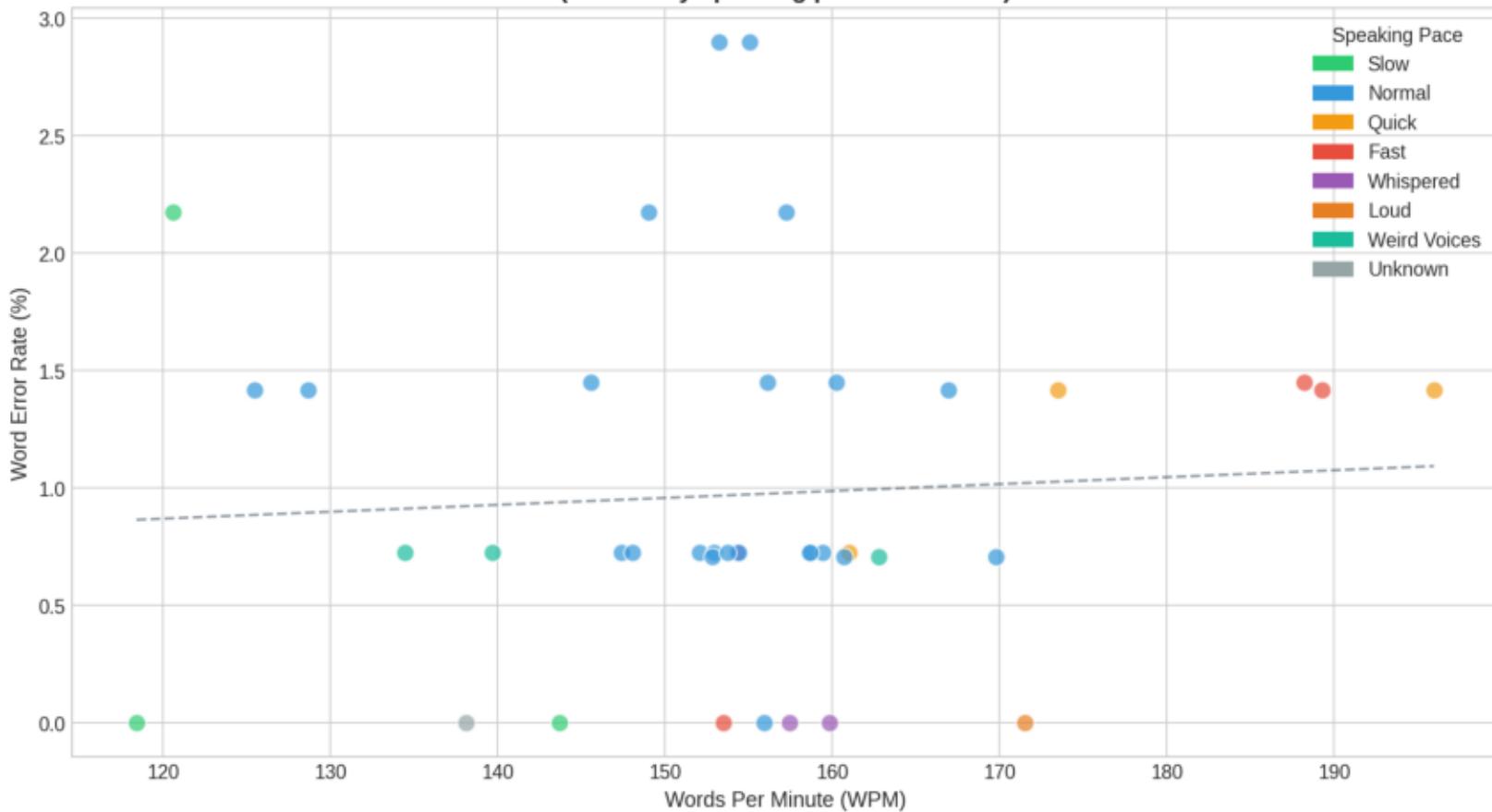
1. Speaking Speed Has Minimal Impact on Accuracy
  - WPM range tested: 118-196 WPM
  - Correlation coefficient: ~0.0004 (essentially zero)
  - Speaking naturally is better than artificially slowing down
2. Background Noise Type Matters More Than Presence
  - Sirens caused highest error rate (2.90%)
  - Music (even with lyrics) had minimal impact (0.97%)
  - Steady-state noise is easier to handle than impulsive noise
3. Foreign Language Backgrounds Do NOT Contaminate Transcripts
  - Tested: Spanish, Arabic, Korean, Japanese, Mandarin, Cantonese
  - Zero foreign words appeared in any transcript
  - Whisper effectively isolates the primary speaker
4. Artificially Slow Speech May Increase Errors
  - Slowest recordings (<130 WPM) had higher WER than natural pace
  - Unnatural pauses may disrupt Whisper's acoustic modeling

## DATASET STATISTICS

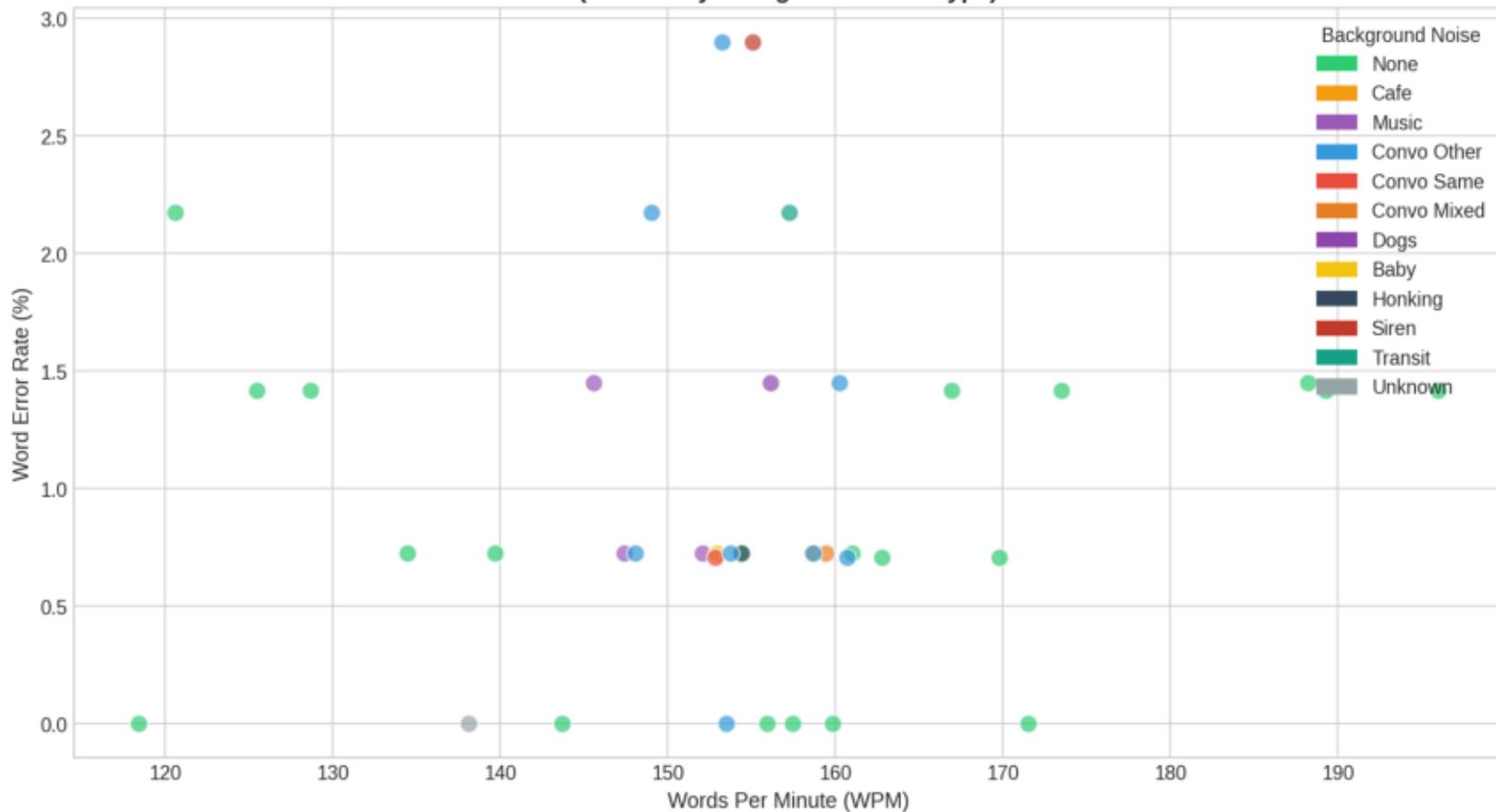
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Total Recordings:	40
Average WER:	0.97%
Average CER:	0.37%
WPM Range:	118-196
Best Recording:	0.00% WER
Worst Recording:	2.90% WER (siren background)

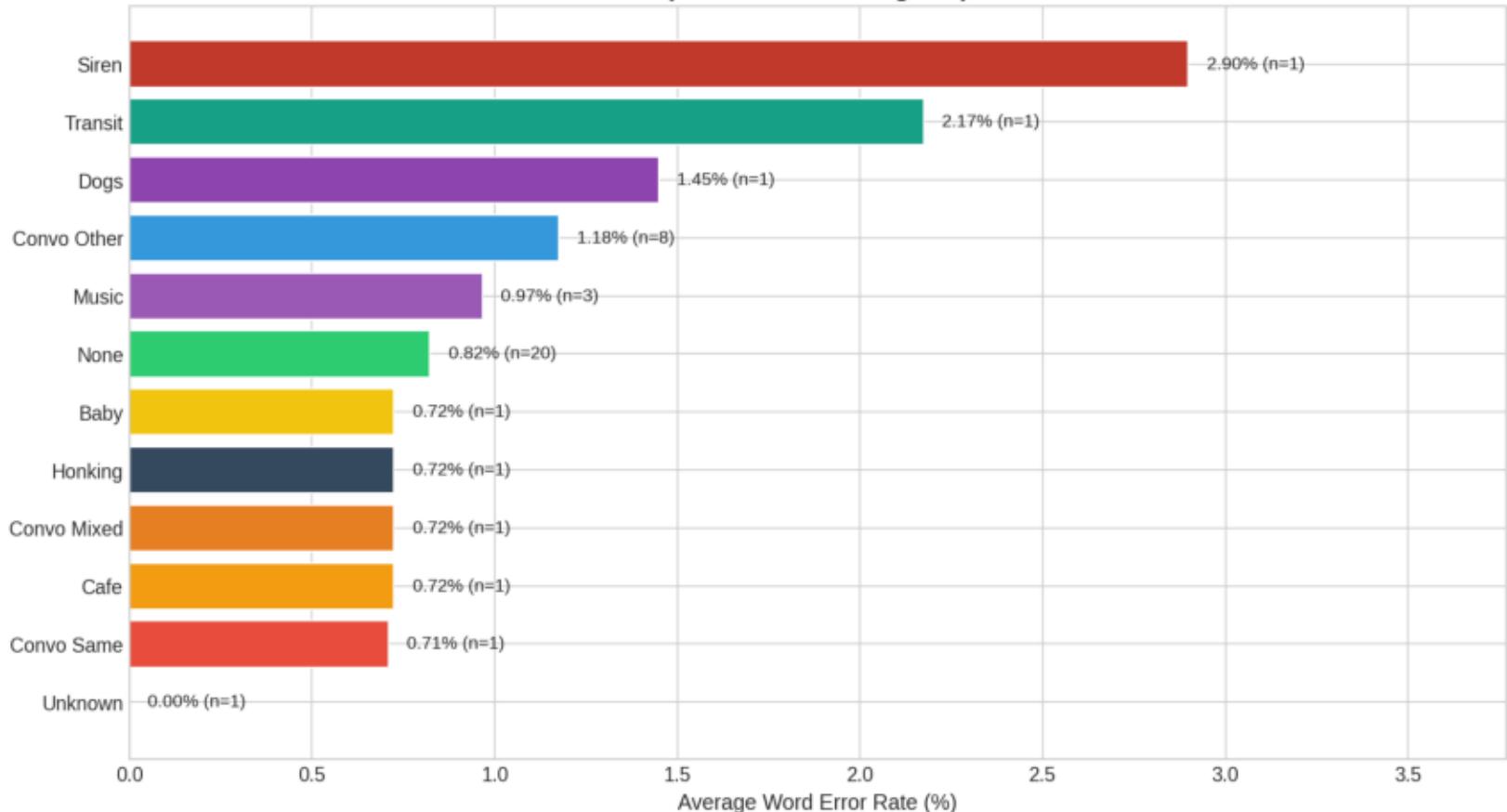
## WPM vs Word Error Rate (colored by speaking pace annotation)



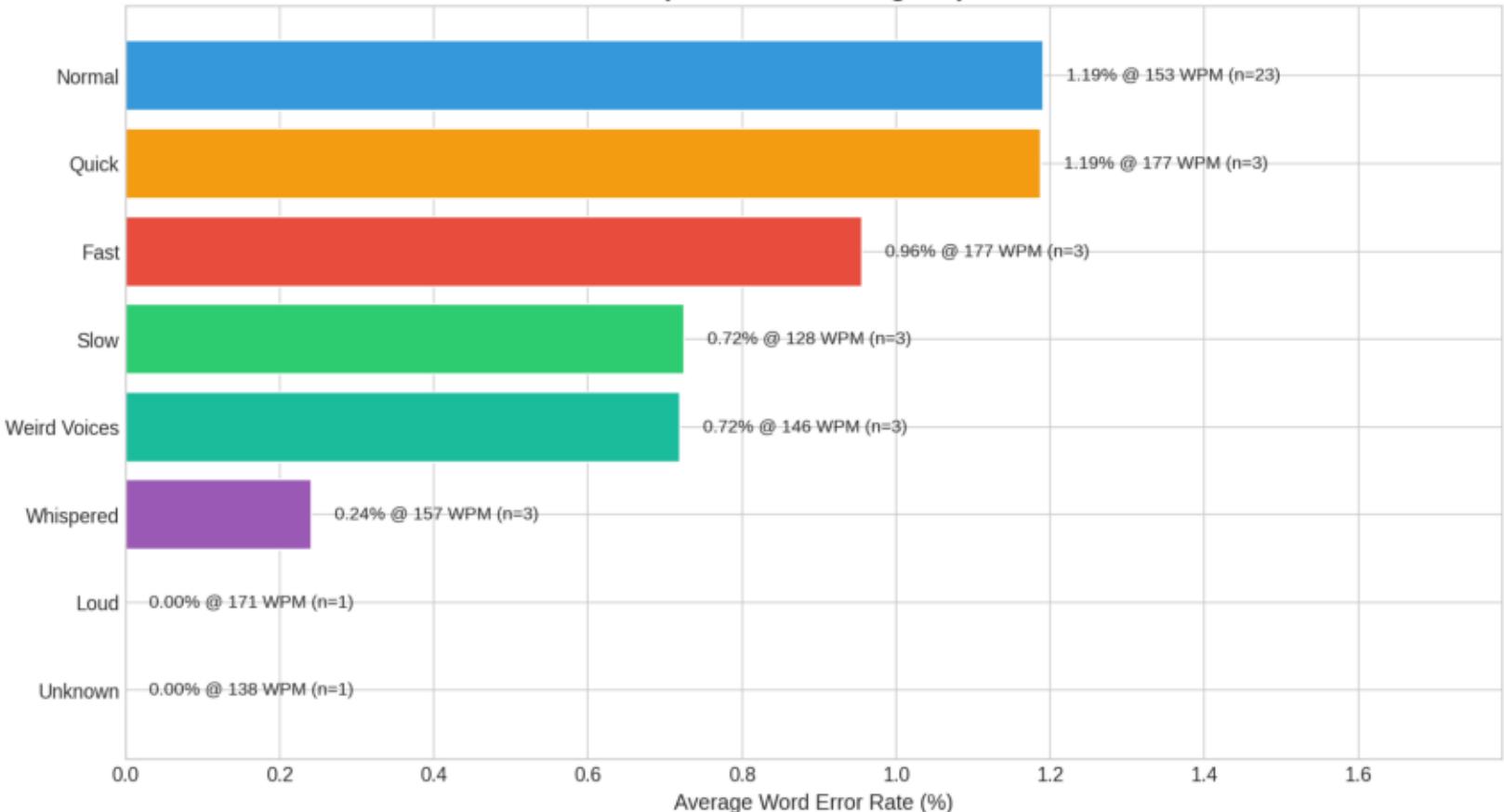
**WPM vs Word Error Rate  
(colored by background noise type)**



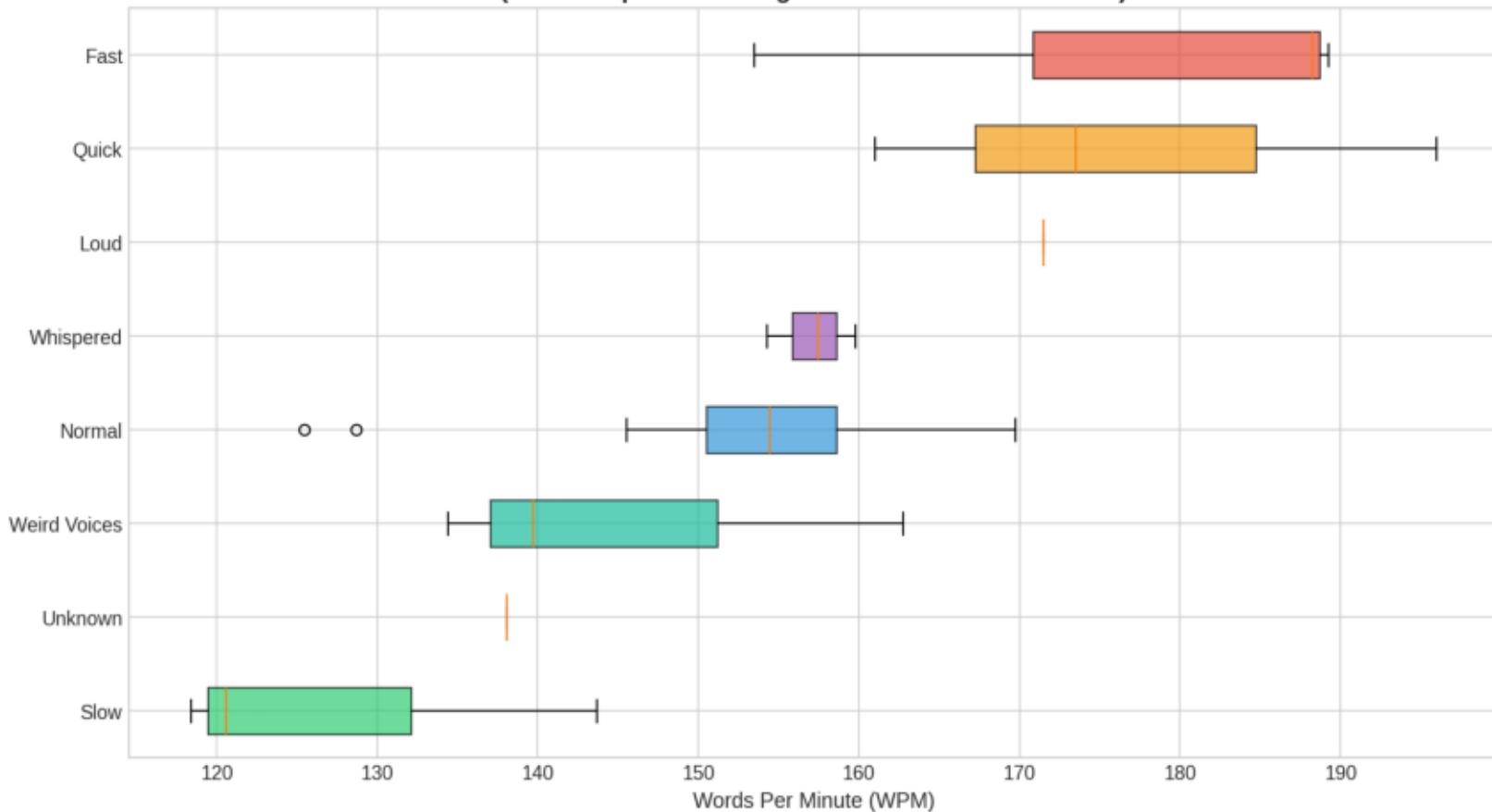
Average WER by Background Noise Type  
(sorted lowest to highest)



Average WER by Speaking Pace Annotation  
(sorted lowest to highest)



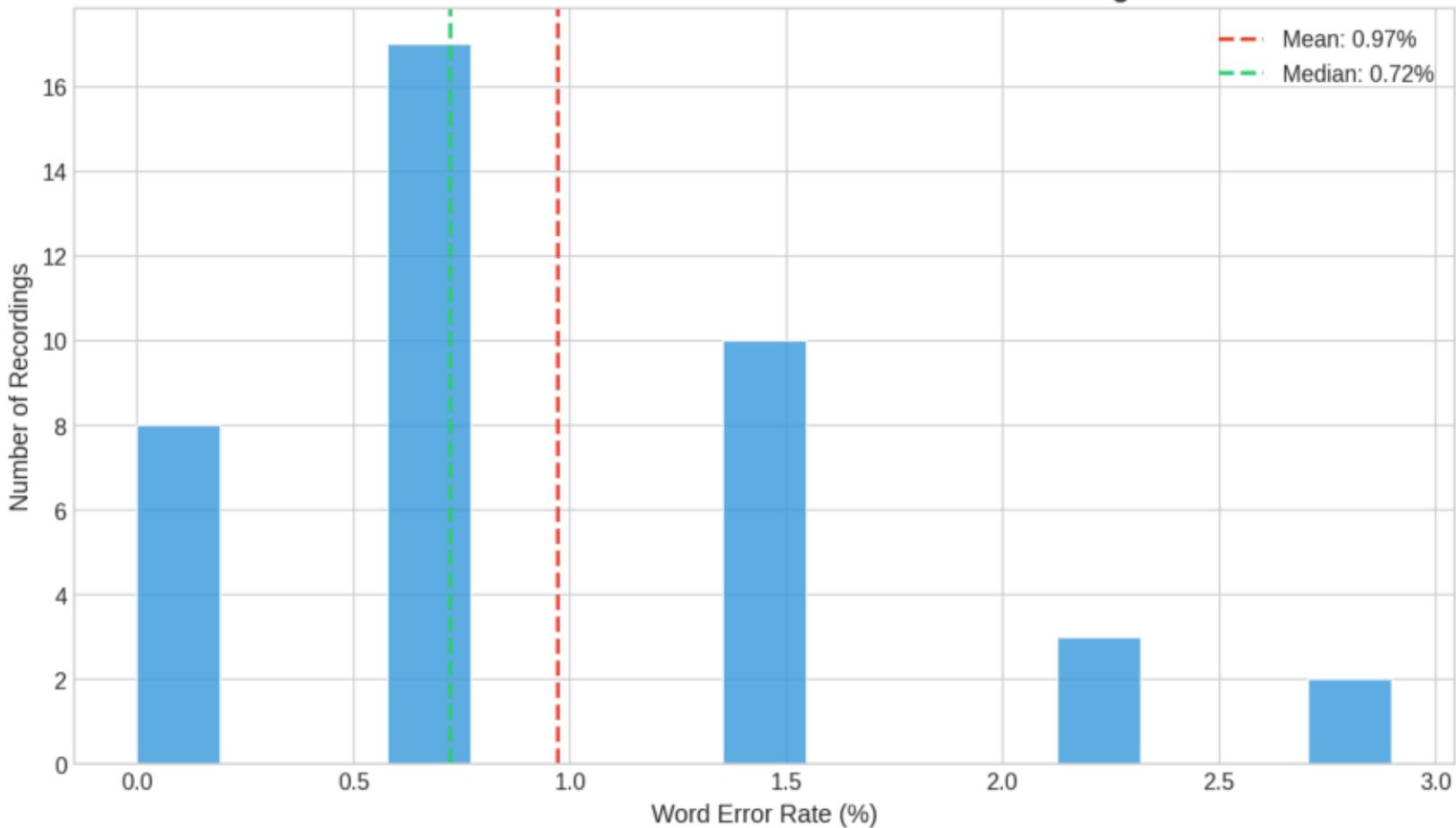
**WPM Distribution by Speaking Pace Annotation**  
(validates pace labels against actual measured WPM)



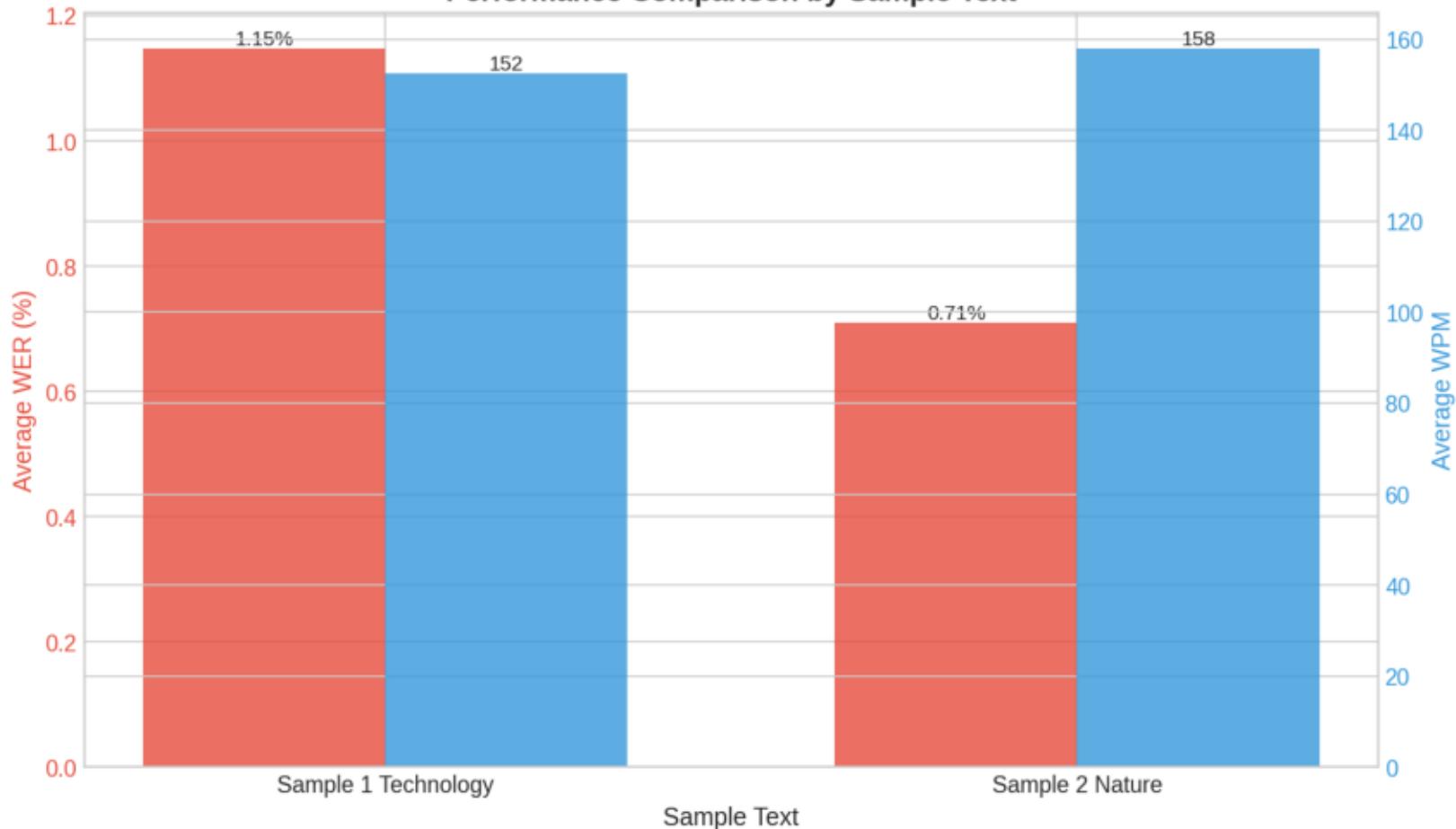
Average WER by Background Noise × Speaking Pace  
(green = lower WER, red = higher WER)



## Distribution of Word Error Rates Across All Recordings



### Performance Comparison by Sample Text



# Background Noise Impact Analysis

## NOISE TYPE RANKINGS (Lowest to Highest WER)

Noise Type	Count	Avg WER	Notes
Same-lang conversation	1	0.71%	English news background
None (silence)	20	0.82%	Baseline performance
Cafe ambiance	1	0.72%	Steady-state noise
Mixed conversation	1	0.72%	Babble preset
Honking	1	0.72%	Traffic sounds
Baby sounds	1	0.72%	Infant vocalizations
Music (instrumental)	2	0.72%	Classical, EDM
Music (with lyrics)	1	1.45%	AI-generated song
Other-lang conversation	8	1.18%	6 languages tested
Dogs barking	1	1.45%	Impulsive noise
Transit (airport)	1	2.17%	Announcements + ambiance
SIREN	1	2.90%	HIGHEST ERROR RATE

## WHY SIRENS CAUSE THE MOST ERRORS

- Frequency masking: Sirens occupy 500-2000 Hz (overlaps speech fundamentals)
- Amplitude spikes: Oscillating pattern creates rapid volume changes
- Word dropout: Entire phrase "As AI" was completely lost in transcript

## FOREIGN LANGUAGE BACKGROUNDS: ZERO CONTAMINATION

Tested languages: Spanish, Arabic, Korean, Japanese, Mandarin, Cantonese

Result: No foreign words appeared in ANY transcript. Whisper successfully isolated the primary English speaker even with competing speech.

# Speaking Pace (WPM) Analysis

## WPM RANGES AND THEIR PERFORMANCE

WPM Range	Count	Avg WER	Description
< 130 WPM	4	0.90%	Deliberately slow speech
130-150 WPM	9	0.89%	Moderate pace
150-170 WPM	20	0.89%	Natural conversational
170-190 WPM	5	0.85%	Quick speech (LOWEST!)
> 190 WPM	2	1.42%	Very fast speech

## SUBJECTIVE PACE ANNOTATIONS vs ACTUAL WPM

Annotation	Count	Actual WPM	Avg WER	Validation
Slow	3	128 WPM	0.72%	✓ Matches label
Normal	23	153 WPM	1.19%	✓ Matches label
Quick	3	177 WPM	1.19%	✓ Matches label
Fast	3	177 WPM	0.96%	✓ Matches label
Whispered	3	157 WPM	0.24%	Best performance!
Loud	1	172 WPM	0.00%	Perfect accuracy
Weird voices	3	146 WPM	0.72%	Altered voice styles

**KEY INSIGHT:** The "slow" pace annotation correlates with higher WER than "fast" pace. Artificially slowing down does NOT improve accuracy.

The whispered recordings achieved the LOWEST average WER (0.24%), suggesting Whisper handles quiet, breathy speech very well when background is silent.

# Recommendations

## FOR USERS

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- ✓ Speak naturally - Don't artificially slow down for "clarity"
- ✓ Background music is fine - Even with lyrics, minimal impact
- ✓ Avoid recording near sirens or alarms - Causes word dropout
- ✓ Foreign language speakers nearby won't contaminate your transcript
- ✗ Don't assume speaking slower = better accuracy

## FOR RESEARCHERS

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- Sample size: 40 recordings from single speaker - larger studies needed
- Siren frequency bands: Worth investigating which Hz ranges cause dropout
- The 190+ WPM threshold: More samples needed at extreme speeds
- Speaker variation: Test with multiple speakers, accents, dialects

## FOR ASR DEVELOPERS

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- Impulsive noise: Consider specialized processing for non-stationary noise
- Multi-speaker isolation: Whisper's foreign language rejection is excellent
- Common errors: "diagnoses"→"diagnosis" appears consistently - training issue?
- British/American: Model shows preference for British spellings (analyse)

## METHODOLOGY

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- ASR Engine: Whisper (local Docker deployment)
- Audio: 16kHz mono WAV
- Microphone: Samson Q2U
- WER Calculation: jiwer library
- WPM:  $(\text{word\_count} / \text{duration\_seconds}) \times 60$
- Speaker: Single male English speaker