

# Speech Features for Depression Detection: A Taxonomy

## What is Speech? (Formal Definition)

Speech is a **time-varying acoustic signal** produced by the human vocal apparatus, consisting of:

1. **Phonation** - vocal fold vibration generating the fundamental frequency (F0)
2. **Articulation** - shaping of the vocal tract to produce distinct sounds
3. **Prosody** - suprasegmental features (rhythm, stress, intonation)

Speech can be analyzed in multiple domains:

- **Time domain** - amplitude over time (waveform)
  - **Frequency domain** - spectral content (via Fourier transform)
  - **Time-frequency domain** - spectrograms, MFCCs
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## Feature Categories

### 1. Prosodic Features (Suprasegmental)

#### 1.1 Pitch (Fundamental Frequency, F0)

**Definition:** The rate of vocal fold vibration, perceived as voice pitch.

**Measurement:** In Hertz (Hz), typically 85-180 Hz (male), 165-255 Hz (female)

**Features extracted:**

- Mean F0
- F0 standard deviation
- F0 range (max - min)
- F0 contour (rising/falling patterns)
- Jitter (cycle-to-cycle F0 variation)

**Depression association:** Lower mean F0, reduced F0 variability (monotone)

#### 1.2 Energy/Intensity

**Definition:** Amplitude of the speech signal, perceived as loudness.

**Measurement:** In decibels (dB)

**Features extracted:**

- Mean energy
- Energy standard deviation
- Energy range
- Shimmer (cycle-to-cycle amplitude variation)

**Depression association:** Lower overall energy, reduced variation

## 1.3 Temporal Features

**Definition:** Timing and rhythm of speech production.

**Features extracted:**

- Speech rate (syllables/phonemes per second)
- Articulation rate (excluding pauses)
- Pause duration (mean, max, total)
- Pause frequency
- Speech-to-pause ratio
- Response latency

**Depression association:** Slower rate, longer/more pauses, increased latency

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## 2. Spectral Features

### 2.1 Mel-Frequency Cepstral Coefficients (MFCCs)

**Definition:** Coefficients representing the short-term power spectrum on a mel scale (approximating human auditory perception).

**Typically extracted:**

- 13 MFCCs (static)
- Delta MFCCs (first derivative)
- Delta-delta MFCCs (second derivative)
- = 39 features total

**Why useful:** Captures vocal tract characteristics, widely used in speech recognition

### 2.2 Formants

**Definition:** Resonant frequencies of the vocal tract, F1, F2, F3...

**Features:**

- F1 (related to vowel height, ~300-800 Hz)
- F2 (related to vowel frontness, ~800-2500 Hz)
- F3 (speaker characteristics, ~2000-3500 Hz)
- Formant bandwidth
- Formant transitions

**Depression association:** Changes in articulation precision

### 2.3 Other Spectral Features

- **Spectral centroid** - "center of mass" of spectrum
  - **Spectral flux** - rate of change in spectrum
  - **Spectral rolloff** - frequency below which X% of energy lies
  - **Spectral entropy** - randomness/uniformity of spectrum
  - **Harmonic-to-Noise Ratio (HNR)** - voice quality measure
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### 3. Voice Quality Features

#### 3.1 Jitter

**Definition:** Cycle-to-cycle variation in fundamental frequency (F0)

**Types:** Local jitter, RAP, PPQ5

**Units:** Percentage or absolute (microseconds)

#### 3.2 Shimmer

**Definition:** Cycle-to-cycle variation in amplitude

**Types:** Local shimmer, APQ3, APQ5, APQ11

**Units:** Percentage or dB

#### 3.3 Harmonic-to-Noise Ratio (HNR)

**Definition:** Ratio of periodic (harmonic) to aperiodic (noise) components

**Units:** dB

**Interpretation:** Higher = clearer voice; lower = breathier/hoarser

**Depression association:** Increased jitter/shimmer, decreased HNR

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### 4. Linguistic/Content Features

#### 4.1 Lexical Features

- Word count
- Type-token ratio (vocabulary diversity)
- Use of first-person pronouns
- Negative emotion words
- Absolute terms ("always", "never")

#### 4.2 Semantic Features

- Sentiment scores
- Topic modeling
- LIWC (Linguistic Inquiry and Word Count) categories

**Note:** Requires transcript (ASR or manual)

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### 5. Deep Learning Representations

#### 5.1 Self-Supervised Representations

- **Wav2Vec 2.0** - Facebook/Meta's pre-trained model
- **HuBERT** - Hidden-Unit BERT
- **WavLM** - Microsoft's model

**Advantage:** Learn features directly from data, often outperform hand-crafted features

#### 5.2 Spectrogram-based

- Log-mel spectrograms → CNN
  - Raw waveform → 1D CNN
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## Statistical Methods for Measuring Differences

### Descriptive Statistics

- Mean, median, mode
- Standard deviation, variance
- Skewness, kurtosis
- Range, percentiles

### Functionals (Applied to LLDs)

Low-level descriptors (LLDs) computed frame-by-frame are summarized using functionals:

- Mean, std, min, max
- Quartiles (25th, 50th, 75th percentile)
- Linear regression coefficients
- Peaks (number, mean distance)

### Statistical Testing

- **T-tests** - compare means between groups
- **Mann-Whitney U** - non-parametric comparison
- **Effect size** (Cohen's d) - magnitude of difference
- **ANOVA** - multiple group comparison

### Machine Learning Evaluation

- Accuracy, Precision, Recall, F1-score
  - **Concordance Correlation Coefficient (CCC)** - standard for AVEC/regression
  - ROC-AUC (classification)
  - Mean Absolute Error, RMSE (regression)
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## Common Feature Extraction Tools

Tool	Description	Language
<b>OpenSMILE</b>	Comprehensive feature extraction	C++
<b>Praat</b>	Phonetic analysis, F0/formants	Praat script
<b>Librosa</b>	Python audio analysis	Python
<b>pyAudioAnalysis</b>	Audio feature extraction	Python
<b>SpeechBrain</b>	Deep learning toolkit	Python
<b>Parselmouth</b>	Praat in Python	Python

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## Feature Sets / Standards

**eGeMAPS (extended Geneva Minimalistic Acoustic Parameter Set)**

- 88 features
- Standardized, interpretable
- Used in AVEC challenges

### **ComParE (Computational Paralinguistics Challenge)**

- ~6000+ features
- Brute-force approach
- Used in INTERSPEECH challenges

### **IS09-IS13 Feature Sets**

- INTERSPEECH challenge feature sets
- Various sizes (384-6373 features)

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## **Key References**

- Schuller, B., et al. (2016). "The Geneva Minimalistic Acoustic Parameter Set (GeMAPS) for Voice Research and Affective Computing." IEEE TAC.
- Cummins, N., et al. (2015). "A review of depression and suicide risk assessment using speech analysis." Speech Communication.