

A Methodology for Deriving VoIP Equipment Impairment Factors for a mixed NB/WB Context

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Outline

- 1 Background and Motivation
- 2 Our Goal
- 3 Our Approach
- 4 Our Results
- 5 Our Achievements

Background

- VoIP enables **packet-based** telecommunication.



- It is cheap and cost effective.
- Supports **WB** telephony conveniently.
- QoS provisioning remains a problem.
- Speech quality estimation is important.

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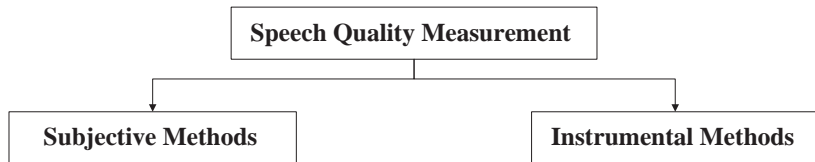


Figure: Various categories of speech quality assessment methods.

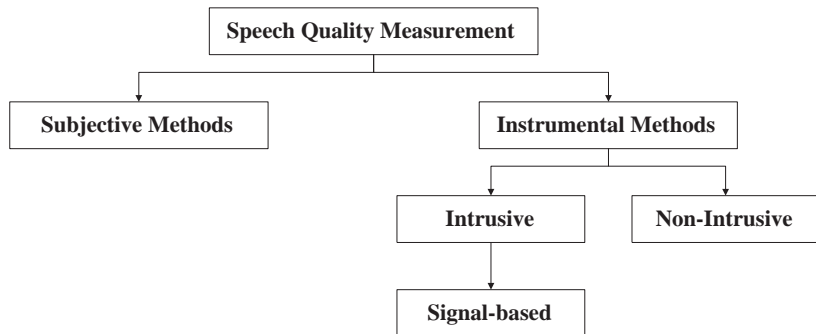


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Speech Quality

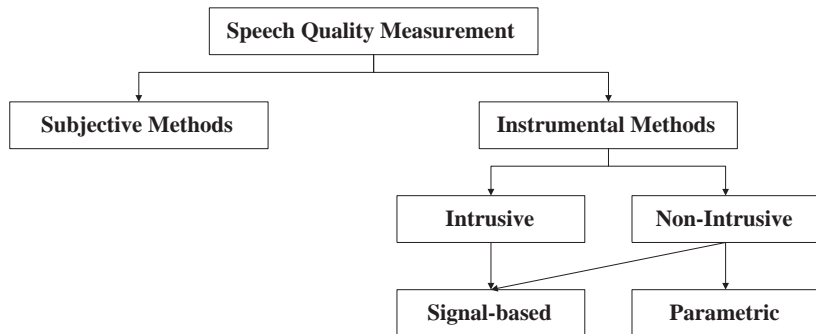


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Background

Speech Quality Standardization

- International Telecommunications Union – **ITU-T** – deals with the standards.
- Intrusive: ITU-T P.862.* – NB-PESQ and **WB-PESQ**.
- Nonintrusive:
 - 1 Parametric: ITU-T G.107 – The **E-Model**.
 - 2 Signal-based: ITU-T P.563.

Our Goal

Equipment Impairment Factor

To derive equipment impairment factors for **WB** VoIP as described by the **E-Model**:

$$SpeechQuality = f(Impairments) \quad (1)$$

$$I_{e,WB,eff} = f(codec, loss, \dots) \quad (2)$$

Our Approach

To evolve $I_{e,WB,eff}$ using GP.

Table: Input domain parameters used in evolutionary modeling

No.	Input Parameter	Description
1	$I_{e,WB}$	Impairments due to codecs
2	mlr	mean loss rate
3	PI	packetization interval (ms)
4	mbl	mean burst length
5	grad	Coarse estimate of codec specific loss robustness factor

Our Approach

Simulation Environment

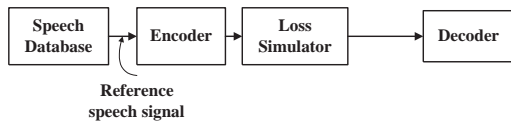


Figure: Simulation system for derivation of $I_{e,WB,eff}$

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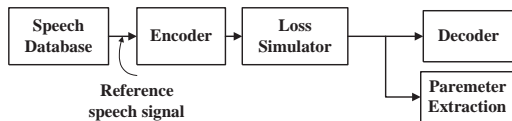


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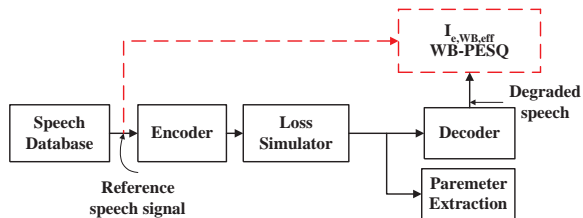


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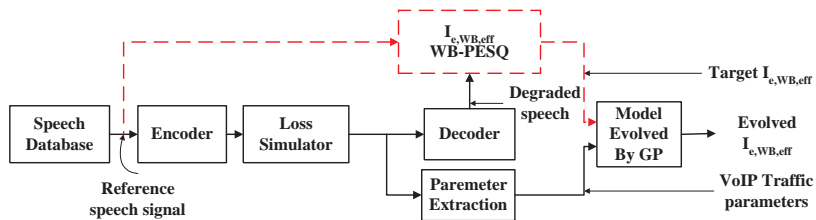


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Our Results

The Derived Models

$$I_{e,WB,eff} = \{11 - mbl + \ln(grad) + grad \times mlr + I_{e,WB} - 2.\log_2(PI)\} \times 0.8619 + 9 \quad (3)$$

$$I_{e,WB,eff} = \left\{ \ln \left(\frac{9 \times (I_{e,WB} + mlr \times grad^2)}{mbl^5 - mlr} \right) + mlr + I_{e,WB} + grad \times mlr \right\} \times 0.8303 + 8.9977 \quad (4)$$

$$I_{e,WB,eff} = (\log_{10}(\log_{10}(\log_2(I_{e,WB} - 2 \times mbl) + mlr))) \times 321.7017 + 95.3708 \quad (5)$$

Our Results

Comparison with E-Model

Table: Prediction gain over E-Model in various network operating conditions

Loss Type	Train	Test
General (Bursty)	14.54	16.36
Specific (Random)	36.72	35.89

Network conditions:

- Fulfilled by ITU-T G.1050.
- 11,280 input/output patterns.
- training data – 70%.
- test data – 30%.

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Codecs

Table: The following codecs were employed

Codec	Type	Modes
ITU-T G.722.1	WB	2
ITU-T G.722.2	WB	9
ITU-T G.729	NB	1
ITU-T G.723.1	NB	1
AMR-NB	NB	2
Total	–	15

Our Achievements

- Our model outperforms ITU-T recommendation G.107 i.e. The **E-Model** – **Criterion B** .
- Demonstrates superior performance for a wide range of ITU-T recommended NB/WB codecs.
- The **E-Model** is downloadable from www.itu.int . **Criterion C**.

Our Achievements

- The results have been accepted for publication in **domain specific IEEE Transactions on Multimedia. Criterion D.**
- Successful and **parsimonious** use of a richer feature space.
- Suitable for real-time evaluation.

Our Achievements

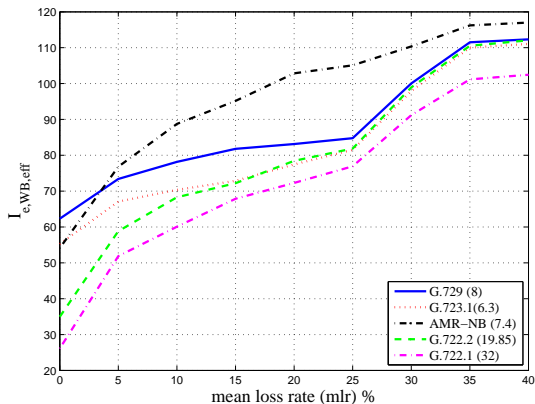


Figure: $I_{e,WB,eff}$ as a function of mlr for various NB/WB codecs. Past research proposed codec specific proposals.

Past approaches:

- Logarithmic – codec specific.
- Quadratic – codec specific.
- **The E-Model** – unified, superior.

Our model:

- Unified.
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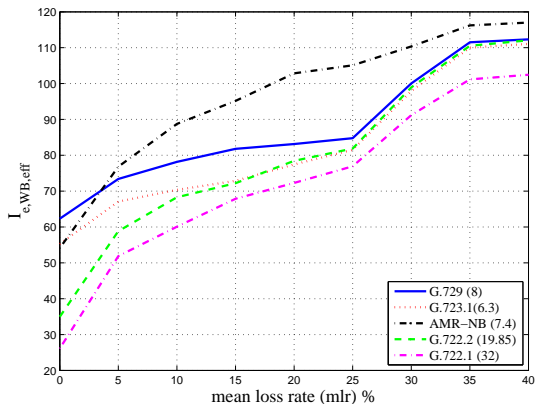


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- E-Model: used in transmission planning and live network monitoring.
- Clients: ISPs, network operators, telcos.

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Thank You