

Internet of Things:

Communication Technologies
Codecs

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Lesson 1

The big picture about codecs

Lesson 1 | The big picture about codecs

1 - Why compress data?

2 - Why does it work?

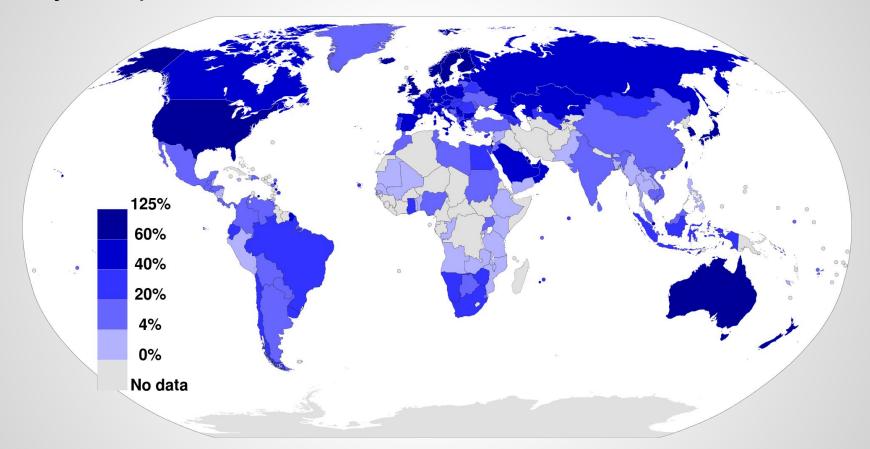
3 - What to look for?





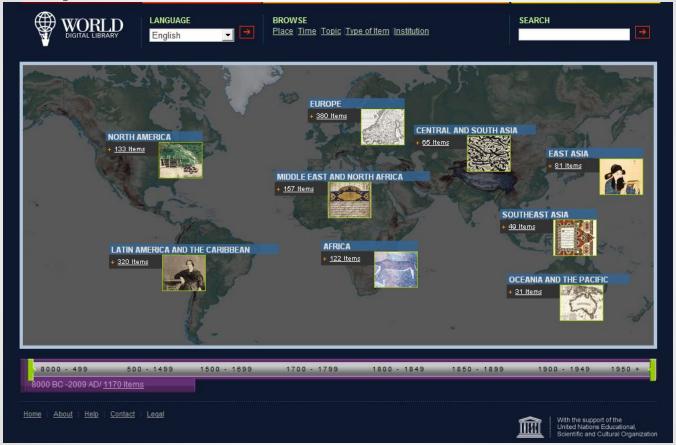
1 - Why compress data?

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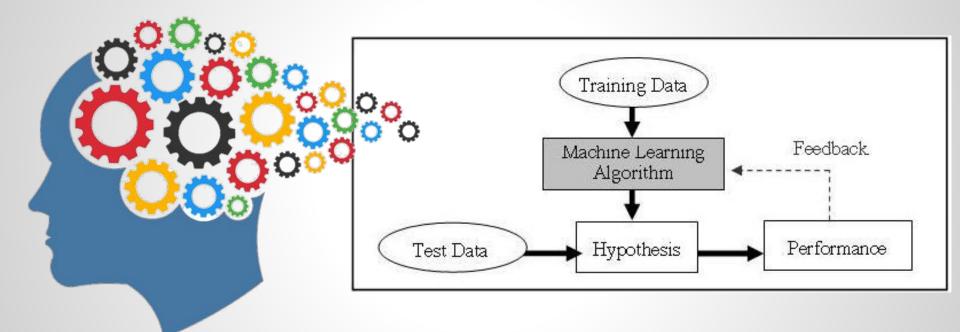
Mobile broadband subscriptions as a percentage of the population in 2012 (ITU-T)

1 | Why compress data?

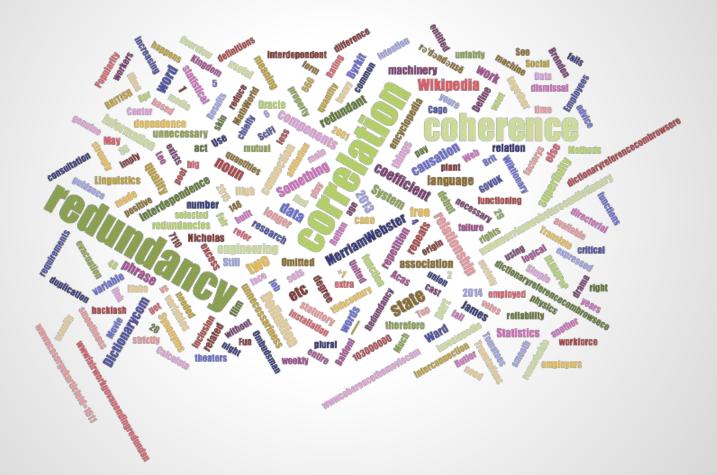


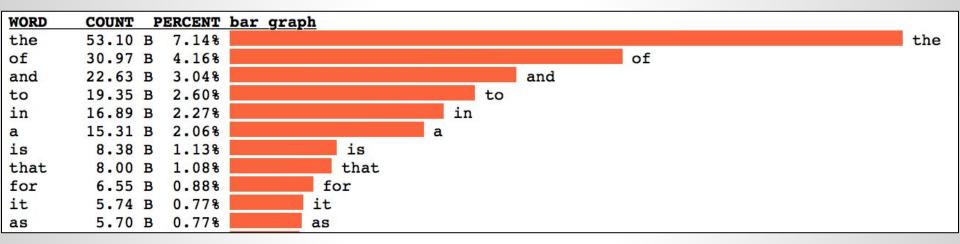
https://en.wikipedia.org/wiki/World_Digital_Library

1 | Why compress data?

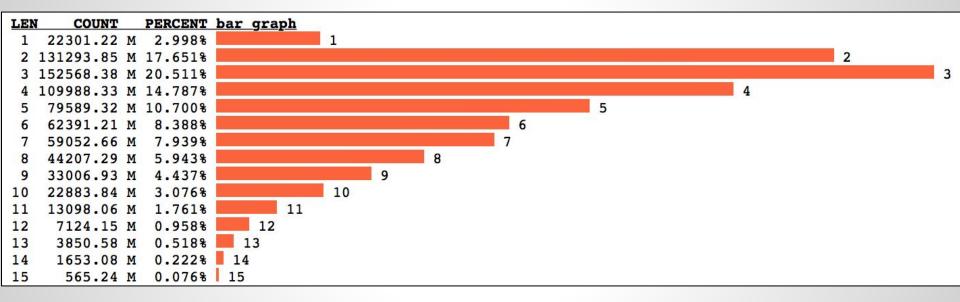




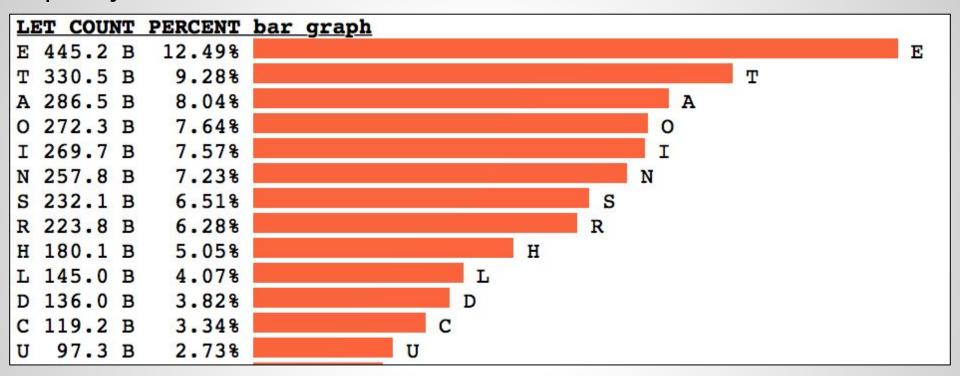




Redundancy in English Text: http://norvig.com/mayzner.html – ETAOIN SRHLDCU



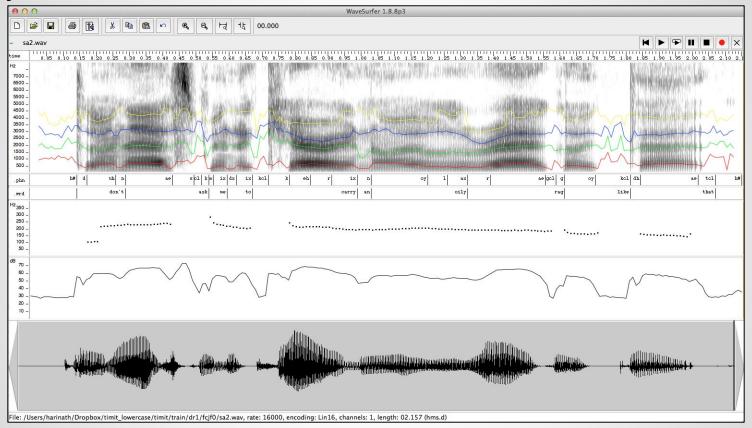
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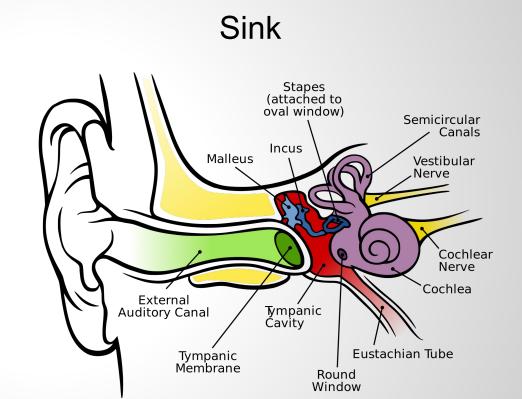


ETAOIN SRHLDCU – QZ



Redundancy in Amplitude, Time and Frequency

Source **Nasal Cavity Palate Oral Cavity** Lips Pharynx **Tongue Epiglottis** Jaw Larynx opening into pharynx **Esophagus** Larynx



Knowledge of Speech Production and Perception



3 - What to look for?

- 3 | What to look for?
 - Compression Efficiency
 - Reconstruction Quality
 - Algorithm Latency
 - Error Resiliency

Computational Complexity





Lesson 2

Waveform Codecs

Lesson 2 | Waveform Codecs

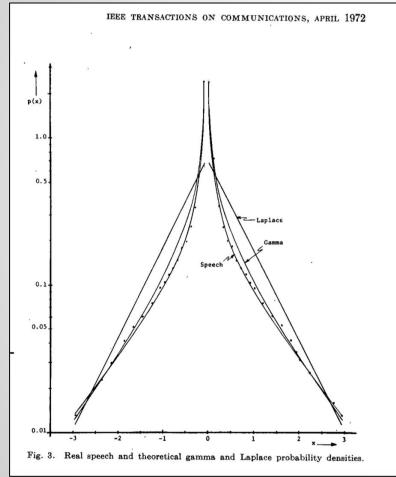
1 - Amplitude - Log Companding (PCM)

2 - Time - Adaptive, Delta Quantization (ADPCM)



3 - Frequency - ADPCM in multiple subbands (Subband ADPCM)





M.D. Paez and T.H. Glisson, "Minimum Mean-Squared-Error Quantization in Speech PCM and DPCM Systems", IEEE Transactions on Communications, April, 1972

G.711 μ -Law

$$y(n) = \operatorname{sign}(x) \frac{\ln(1 + \mu |x(n)|)}{\ln(1 + \mu)},$$

$$sign(x) = \begin{cases} +1, & x(n) > 0, \\ 0, & x(n) = 0, \\ -1, & x(n) < 0 \end{cases}$$

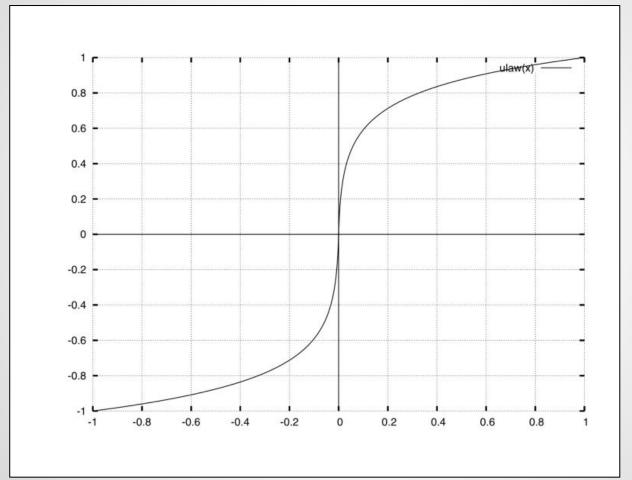
$$\mu = 255$$

G.711 A-Law

$$y(n) = \begin{cases} sign(x) \frac{A |x(n)|}{1 + \ln(A)}, & \text{for } 0 \le |x(n)| < \frac{1}{A}, \\ sign(x) \frac{A |x(n)|}{1 + \ln(A)}, & \text{for } \frac{1}{A} \le |x(n)| < 1. \end{cases}$$

$$A = 87.6$$

A-Law -- 13 bit signed sample → 8 bit quantized sample u-Law -- 14 bit 2's complement → 8 bit quantized sample



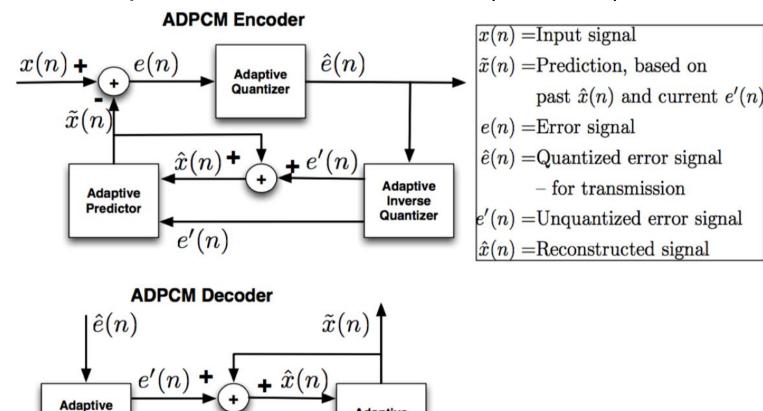


2 - Time - Adaptive, Delta Quantization (ADPCM)

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Inverse

Quantizer



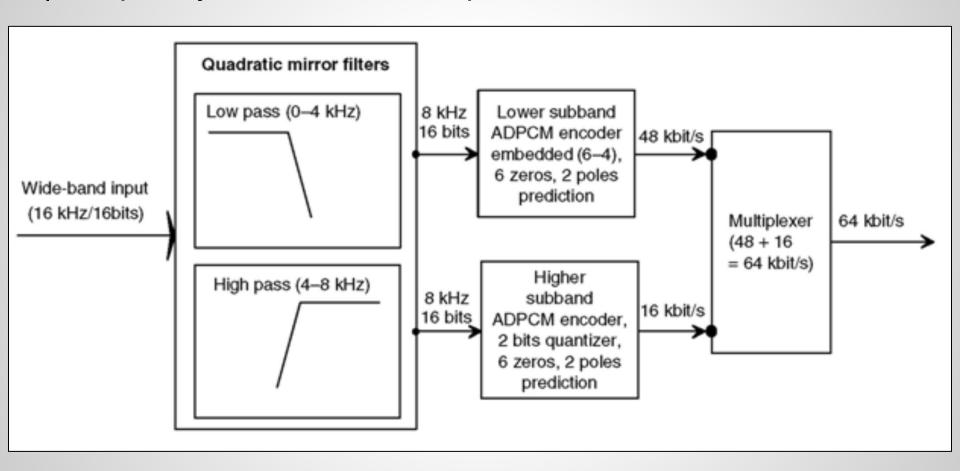
Adaptive

Predictor

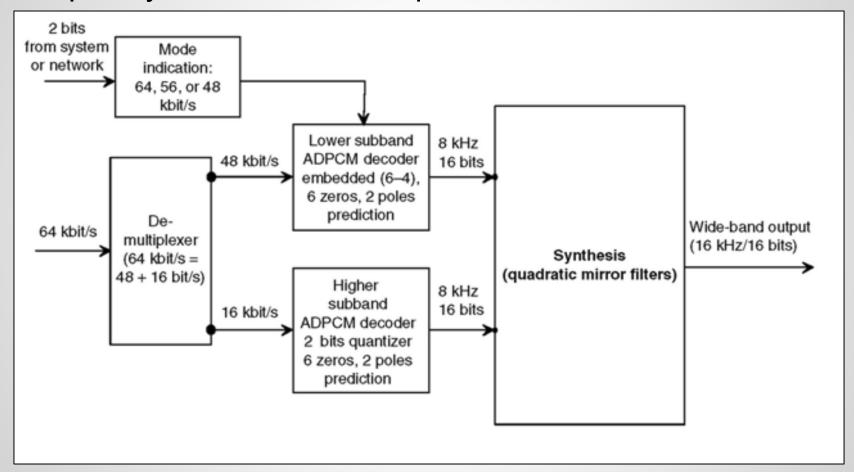


3 - Frequency - ADPCM in multiple subbands (Subband ADPCM)

3 | Frequency - ADPCM in multiple subbands



3 | Frequency - ADPCM in multiple subbands





Lesson 3

Parametric Codecs

Lesson 3 | Parametric Codecs

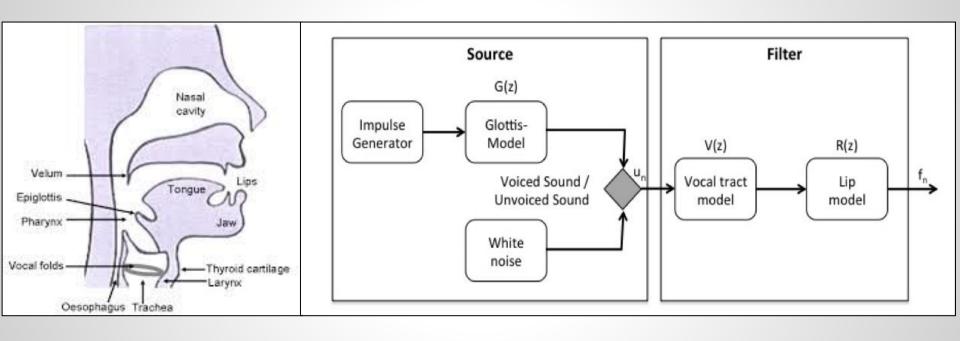
- 1 Speech production model
- 2 Linear Predictive Coding of Speech
- **3** 2G/3G Codecs
- 4 Voice Quality
- 5 VoIP over Wireless



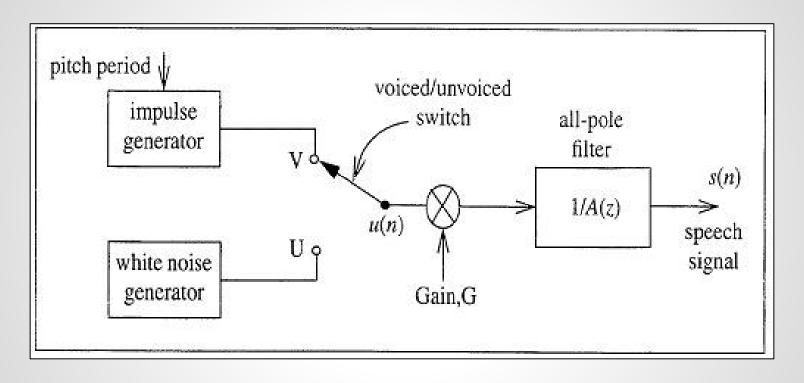


1 - Speech production Model

1 | Speech production model



1 | Speech production model



0.5 to 1.5 bits / sample instead of 8 bits / sample



2 - Linear Predictive Coding of Speech

LPC Model of Speech Production

$$x(n) = \sum_{i=1}^{p} a(i) \ x(n-i) + G u(n)$$

Given a speech segment x(n), n = 0..N - 1, the task is to estimate

- the LPC filter a(i), i = 1..p that approximates the vocal tract filter;
- input to the filter u(n) that approximates the source; and
- the gain G for the source excitation signal.

2 | Linear Predictive Coding of Speech

LPC Formulation

x(n) is predicted as linearly weighted sum of past samples

$$x(n) \approx \tilde{x}(n) = \sum_{i=1}^{p} a(i) x(n-i)$$

The error captures the ignorance missed by the model

$$e(n) = x(n) - \tilde{x}(n)$$

$$= x(n) - \sum_{i=1}^{p} a(i) \ x(n-i)$$

$$E_n = \sum_{i=1}^{p} [e(n)]^2$$

Minimize E_n by differentiating E_n w.r.t each a(i) and set to 0

Autocorrelation Method

The set of equations for LPC parameters a(i) are

$$\sum_{k=1}^{p} a(i)R(|i-k|) = R(i) \ 1 \le i \le p,$$

where, R(i), the autocorrelation at lag i is given by

$$R(i) = \sum_{m=0}^{N-1-i} x(m)x(m+i)$$

The matrix R(|i-k|), $1 \le k \le p$, $1 \le i \le p$ is a Toeplitz matrix in which each descending diagonal from left to right is constant, i.e. $R_{i,j} = R_{i+1,j+1} = R(|i-k|)$. Levinson-Durbin recursive solution is an efficient method to solve for a(i).

2 | Linear Predictive Coding of Speech

Source: Gain and u(n)

$$G^{2} = R(0) - \sum_{k=1}^{p} a(k)R(k)$$

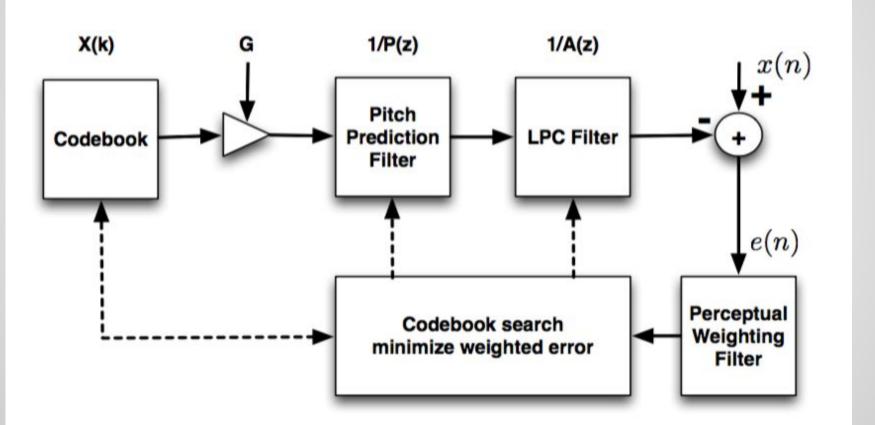
- u(n) is bimodal Voiced and Unvoiced. Voiced segments are periodic with the pitch period.
- Pitch period can be estimated using autocorrelation of low passed speech, center clipped speech, etc [1].
- When there is discernible peak in the autocorrelation function, it is labelled Unvoiced.
- u(n) is a train of impulses at pitch period for voiced signals.
- u(n) is random noise for unvoiced signals.

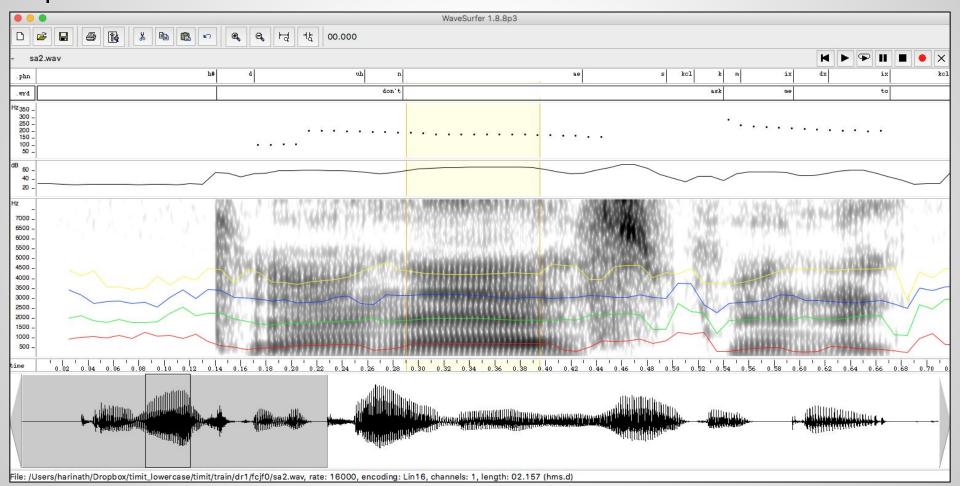
[1] Lawrence Rabiner, "On the Use of Autocorelation Analysis for Pitch Detection" IEEE Transactions On Acoustics, Spech, and signal Processing, Vol. ASSP-25, No.1, February 1977

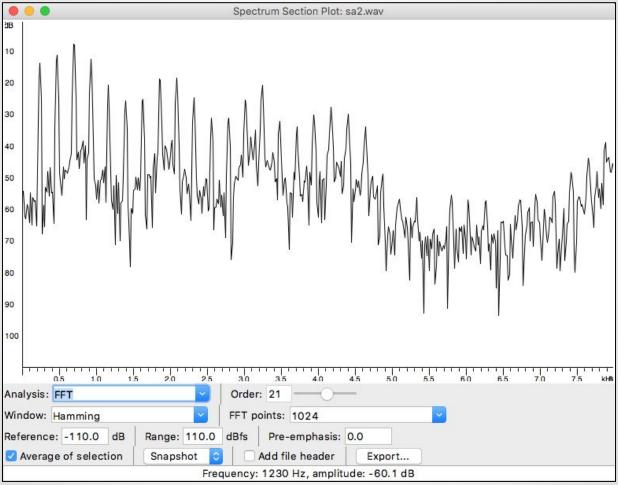


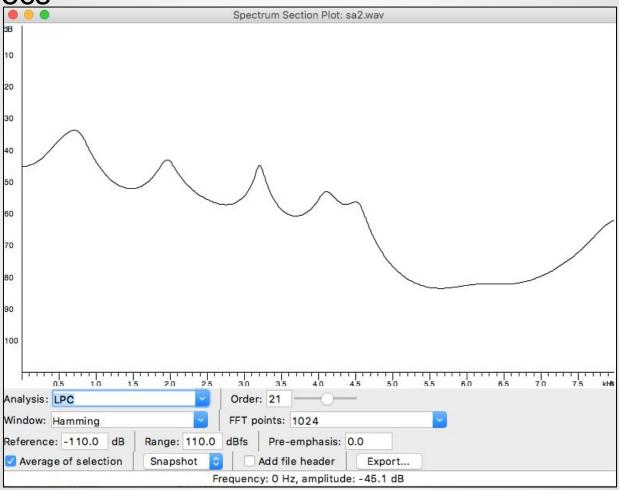
3 - 2G/3G Codecs

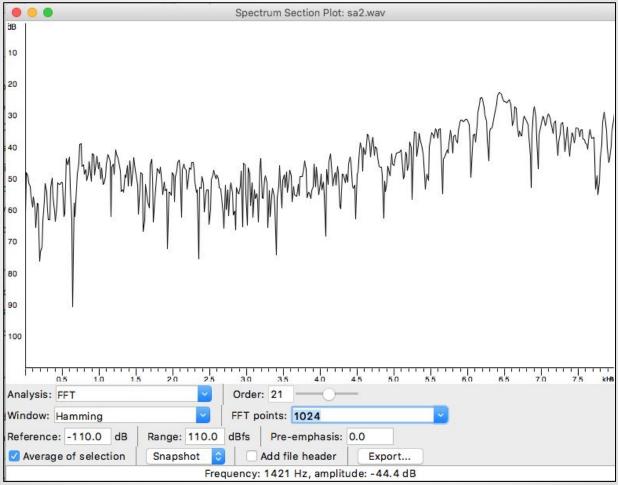
Code Exited Linear Prediction

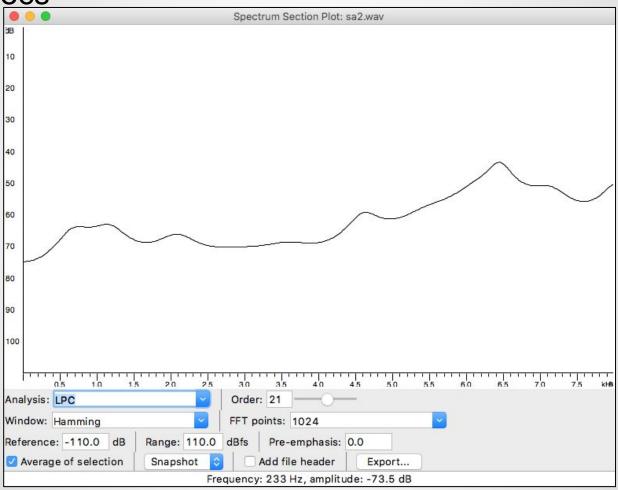














Assessing Voice Quality

SNR =10 log₁₀
$$\left\{ \frac{\sum_{n=0}^{N-1} x^{2}(n)}{\sum_{n=0}^{N-1} (x(n) - \tilde{x}(n))^{2}} \right\}$$

SEGSNR =
$$\frac{10}{K} \sum_{k=0}^{K-1} \log_{10} \left\{ \frac{\sum_{n=0}^{N-1} x^2(kN+n)}{\sum_{n=0}^{N-1} (x(kN+n) - \tilde{x}(kN+n))^2} \right\}$$

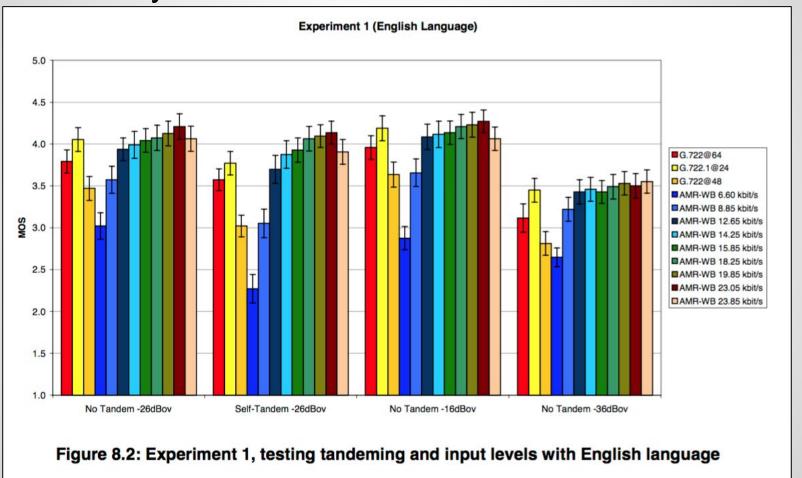
Mean opinion score (MOS)

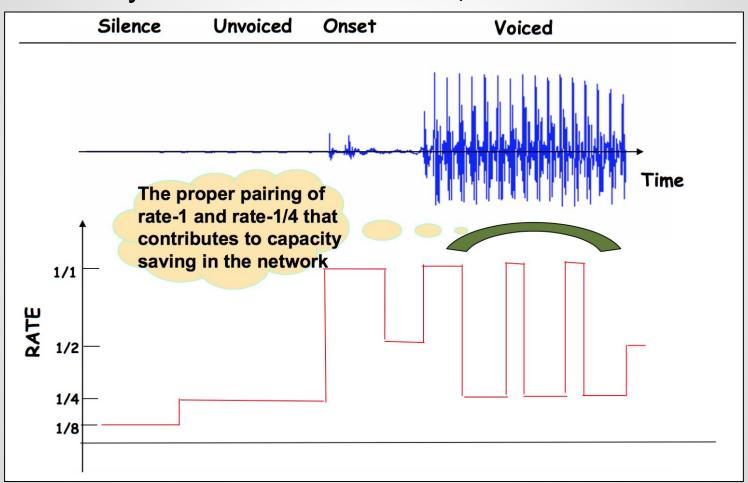
MOS	Quality	Impairment	
5	Excellent	Imperceptible	
4	Good	Perceptible but not annoying	
3	Fair	Slightly annoying	
2	Poor	Annoying	
1	Bad	Very annoying	

https://en.wikipedia.org/wiki/Mean_opinion_score

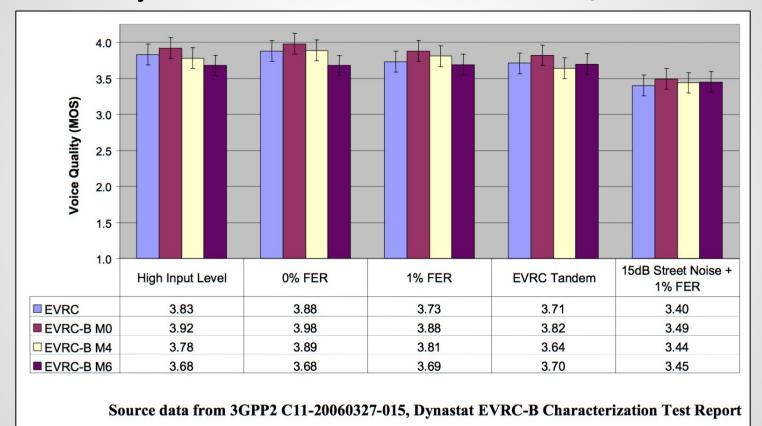
Codec \$	Data rate [kbit/s]	Mean opinion score (MOS)
G.711 (ISDN)	64	4.1
iLBC	15.2	4.14
AMR	12.2	4.14
G.729	8	3.92
G.723.1 r63	6.3	3.9
GSM EFR	12.2	3.8
G.726 ADPCM	32	3.85
G.729a	8	3.7
G.723.1 r53	5.3	3.65
G.728	16	3.61
GSM FR	12.2	3.5

4 | Voice Quality AMR-WB MOS values for different rates





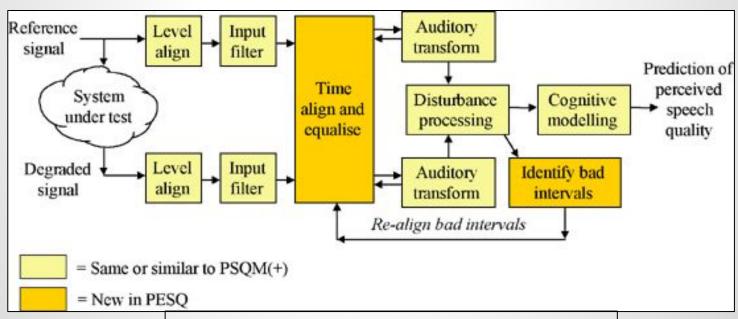
4 | Voice Quality EVRC MOS values for different average data rates



M0 = 9.3 kbps; M4 = 6.6 kbps; M6 = 5.8 kbps

Other Objective Metrics

- PESQ, Perceptual Evaluation of Speech Quality, ITU-T recommendation P.862
- POLQA, Perceptual Objective Listening Quality Assessment, ITU-T recommendation P.863



http://www.opticom.de/technology/pesq.php

3GPP2-C11-20080114-016, "Report and analysis of PESQ under-prediction of EVRC family of speech codecs", Hollywood, CA, Jan 14-18, 2008

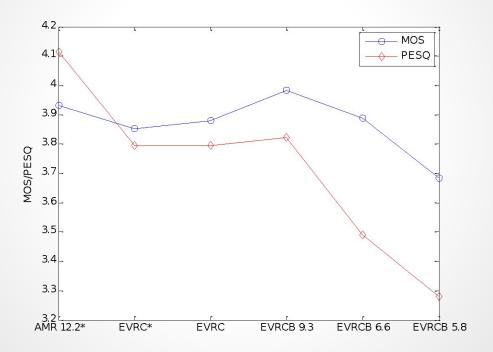
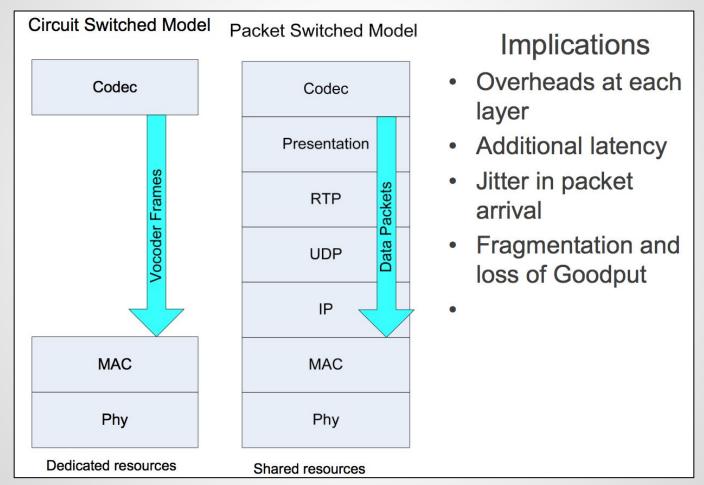


Figure 6 Comparison of PESQ and MOS for different codecs under 0% frame erasure





- RFC 3550: RTP: A Transport Protocol for Real-Time Applications
- RFC 3551: RTP Profile for Audio and Video Conferences with Minimal Control
- <u>RFC 4867</u>: RTP Payload Format and File Storage Format for the Adaptive Multi-Rate (AMR) and Adaptive Multi-Rate Wideband (AMR-WB) Audio Codecs
- RFC 5188: RTP Payload Format for the Enhanced Variable Rate Wideband Codec (EVRC-WB) and the Media Subtype Updates for EVRC-B Codec

RFC 4995 and RFC 5225: RObust Header Compression (ROHC)

- At 4—12 kbps payload traffic, we have a total of 500—1500 bytes per second
- If we transmit data at 20 msec frames, we have a total of 50 frames / second
- So, we have 10—30 bytes in each VoIP packet
- In *IPv4*, the overhead per packet is 40 bytes (IP=20; UDP=8; RTP=12)
- In IPv6, the overhead per packet is 60 bytes (IP=40; UDP=8; RTP=12)
- ROHC compresses VoIP headers to 1—3 bytes
- Robust to lossy channels, long round trip delays, out of sequence arrivals
- Takes advantage of static fields and fields with known properties (e.g. Timestamp)
- Applied to only low bandwidth links (e.g. cellular)

Further Study

- 1. **G.191 : Software tools for speech and audio coding standardization** https://www.itu.int/rec/T-REC-G.191-201003-l/en
- 2. **Digital Processing of Speech Signals,** <u>Lawrence R. Rabiner, Ronald W. Schafer https://books.google.com/books/about/Digital_Processing_of_Speech_Signals.html?</u> id=wltTAAAAMAAJ

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