







Speech Enhancement Evaluation Using Speech Recognition algorithm Yotam Elia, Supervised by Hadas Ofir, Baruch Berdugo

Introduction

Speech enhancement algorithms aims to improve the quality and intelligibility of noisy speech signals.

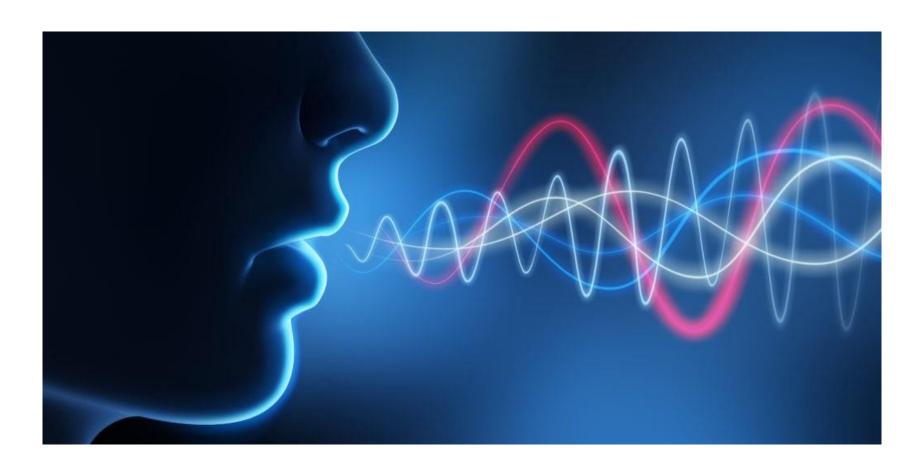
Evaluation SE Algorithm

Multi-process base algorithms give faster and more accurate solution.

Distance Estimation

• Write the difference into files, for WER calculation.

- There exist several objective measures for evaluating SE algorithm performance, but none of them gives a faithful representation of MOS.
- In this project we developed an objective measure for SE performance evaluation based on speech recognition.

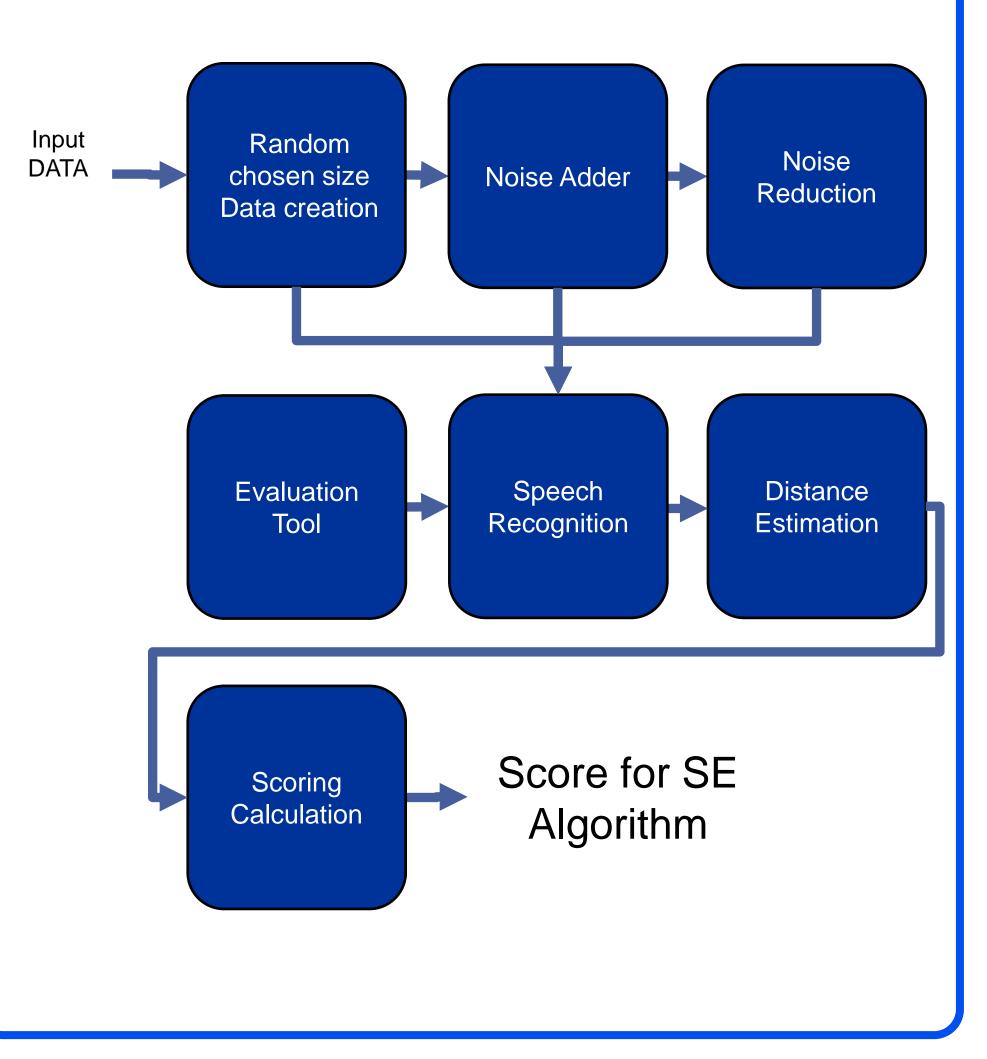


A USV at sea

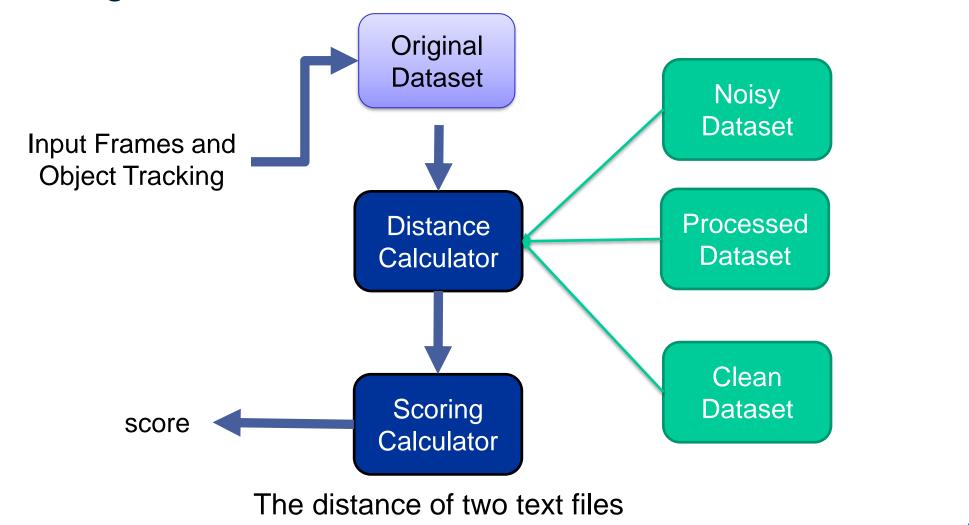
Goals

- Develop a tool for objective evaluation of Speech Enhancement algorithm's performance.
- The evaluation is based on using an Automatic

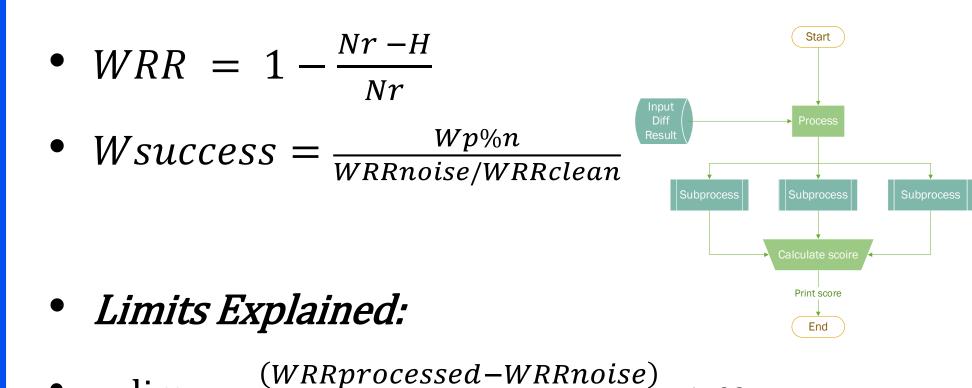
Results are compared to DNS-MOS for sanity check.



Calculate distance between output text files and original files.



Score Calculation



- Speech recognition algorithm.
- Comparison of spectral-subtruction based algorithms and to deep-learning based algorithms.

Challenges

- The performance of SE algorithm can be evaluated based on several criterions: speech clarity, speech intelligibility and more.
- Different people prioritize those criterions differently.
- Some ASR algorithms deal with noisy environment better than others.

Speech Recognition

Speech recognition is based on 3 models.



Evaluation Tool



- Consist of speech recognition algorithm, distance estimation and scoring calculation
- To every noise kind the program create a new process to create fast solution calculation to any process.
- Give a reliable result and print it to the screen.

valuation Tool Multi Thread

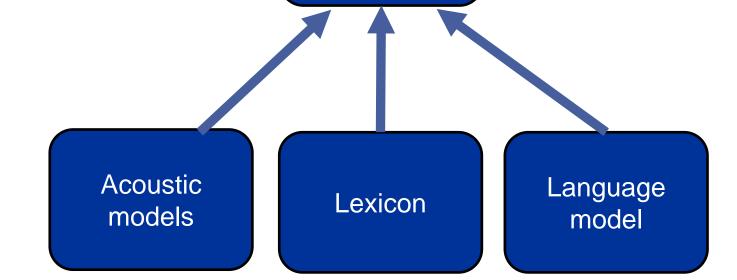
lim $\rightarrow \infty$ WRRnoise $WRRclean \rightarrow 0$ $(WRR processed - WRR noise) \in (0, -1)$ lim WRRnoise $WRR processed \rightarrow 1$

Results

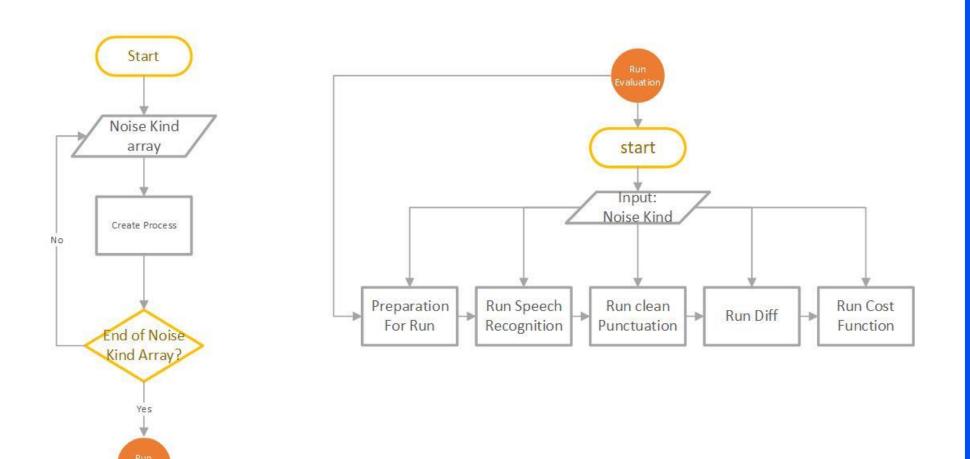
- With Sphinx Algorithm our results is inversely proportional to the DNS-MOS results.
- The results are correlated to the success of the ASR recognition results.

5DB – OMLSA algorithm 5DB – DTLN algorithm

	babble	Mos		babble	Mos
Clean Signal:	20.500%	3.70	Clean Signal:	20,500%	3,70
Noised Signal:	88,100%	2.61	Noised Signal:	93,100%	2.13
Processed Signal:	85,100%	2.71	Processed Signal:	87,500%	2.71
Algorithm Score:	3.77		Algorithm Score:	7.04	



- In our project we used either Sphinx or Baidu ASR algorithms.
 - The acoustic model of Baidu is based on deep-learning. Uses RNN architecture.
 - The acoustic model of Sphinx is based on hidden markov model.
- Note that Sphinx give more reliable results since it is more sensitive to the presence of noise.



- Block diagram showing the process creation and the evaluation procedure.
- The cost of each output is calculated by comparing the text output of the file under examination to the text output of the clean speech.

Estimated distance of OMLSA vs Estimated distance of DTLN

Sphinx algorithm

Conclusions

- Successful estimation of distance with suitable accuracy for evaluation application.
- Compare between spectral-subtruction algorithms and deep learning algorithms.
- Baidu algorithm have good immunity to noisy environment.
- Baidu algorithm works better when the speech is undistorted by the SE algorithm.

