

KubeEdge ASR Offloading

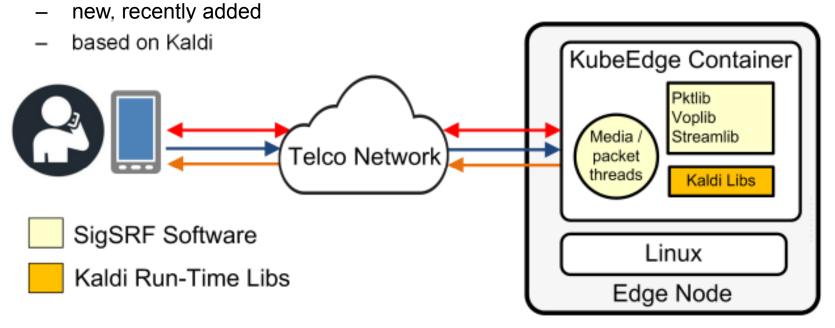
Signalogic, Aug 2020

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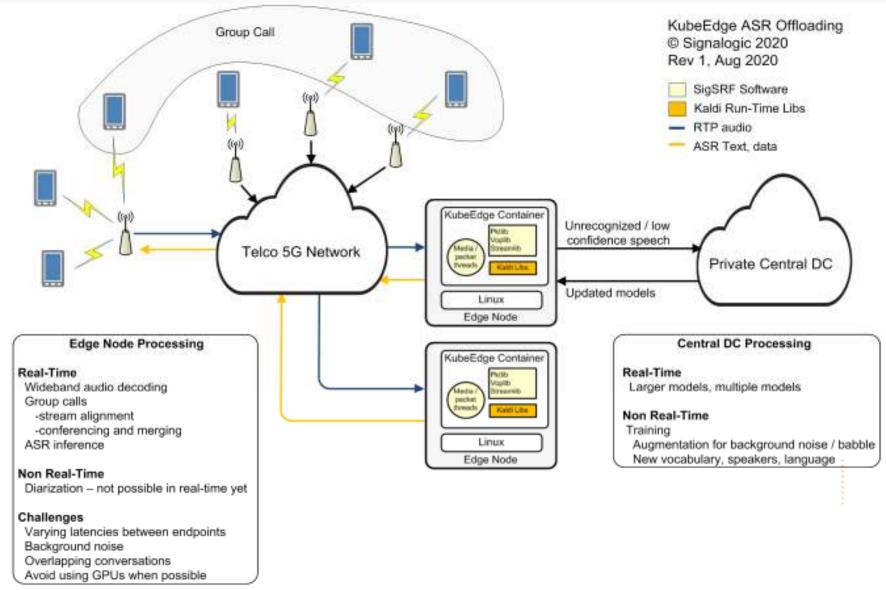
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Overview

- Signalogic is building an ASR offloading demo for KubeEdge
- SigSRF packet + media processing software
 - SigSRF is deployed by telecoms, LEAs, and analytics customers worldwide
 - includes a robust packet interface, including wideband audio codecs, jitter buffer and packet loss handling, stream alignment, etc
- ASR



ASR Offloading



Demo Capability

• ASR based on Kaldi's mini-librispeech model

- subset of librispeech model, which has 200k word vocabulary (English)
- trained with fewer hours, producing a smaller model easier to use for development and testing
- demo uses pre-trained x-vectors and i-vectors no training required

• SigSRF packet + media software

- codecs AMR, AMR-WB, EVS, G729, G726
- RFCs child streams (8108), DTMF (4733), 7198, others
- concurrent sessions 8 (demo subset of 512)
- packet handling jitter buffer, DTX, packet loss mitigation

• Call groups (one or more endpoints)

- conferencing, merging, deduplication
- ASR is applied to call group output

Kaldi Interface

Kaldi real-time inference is called "online decoding"

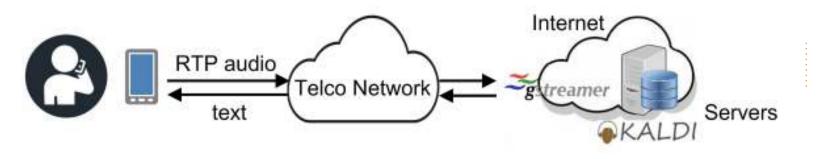
- Kaldi run-time inference expects raw 16-bit audio chunks. They recommend receiving audio as
 - raw audio over TCP/IP
 - via RTP audio packets received and decoded by GStreamer

Kaldi needs wideband audio

- for accuracy benchmarks
- training augmentation, R&D work, published results based on wideband audio

• GStreamer is weak in telecom / wideband audio support

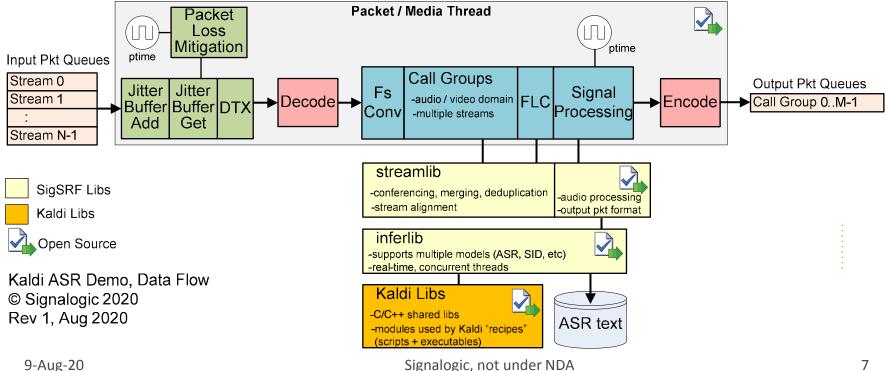
- no support for EVS, AMR-WB support is weak for concurrent threads, reliability
- lack of advanced handling for packet loss, stream alignment between multiple streams within a call, stream gaps (call waiting, music on hold), etc
- no support for RFC8108 (multiple streams from one endpoint)



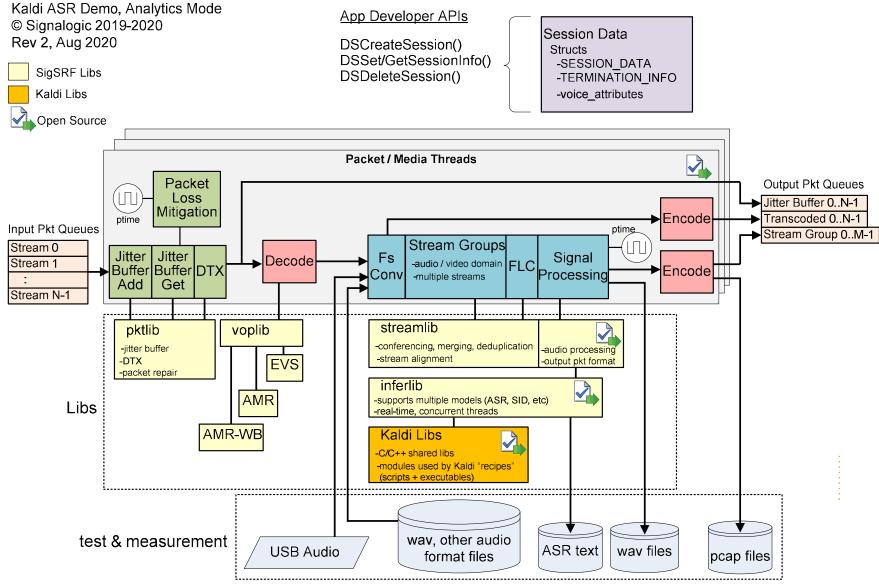
Kaldi Interface, Data Flow

SigSRF replaces GStreamer ٠

- minimum REST APIs required session create/delete/modify
- session create can be specific or give IP:port and let SigSRF auto-detect codec, bitrate, ptime, etc from RTP data flow
- packet input via UDP, pcap input for R&D, testing purposes
- inferlib ٠
 - we added an interface library that in turn interfaces to Kaldi run-time libs



Kaldi Interface, Software Architecture



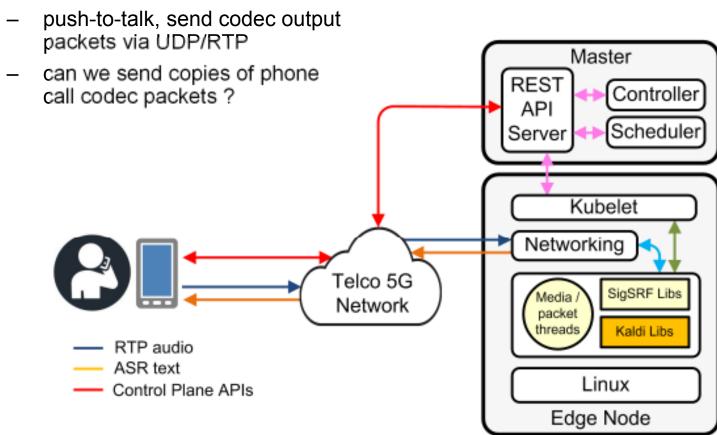
KubeEdge Integration

• SigSRF and Kaldi libs inside KubeEdge container

– minimum 4 x86 cores, 32 GB mem, 1 TB HDD

Mobile device app

creates ASR sessions with REST APIs

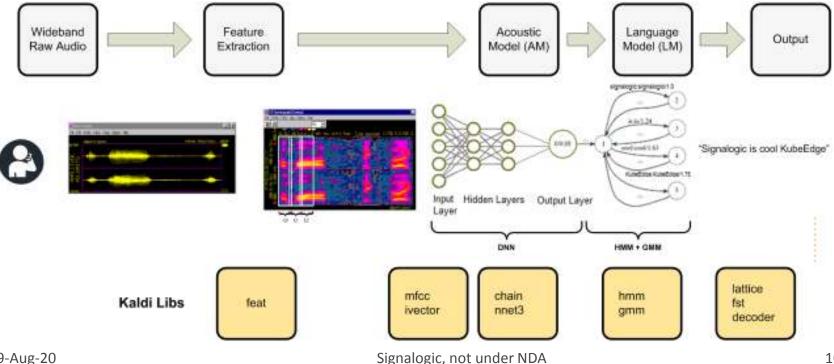


Kaldi Info: Run-Time Inference

- One end-to-end thread on one Xeon x86 core •
 - input is 16-bit raw audio, output is ASR text (plus log, stat files, etc)
 - they maintain a sweet spot of about 2x RTF (real-time factor). They don't use OpenMP, TBB, or other HPC multicore methods
 - ARM cores can be used, but support on Kaldi user groups is limited

Main Kaldi contributors are focused on state-of-the-art R&D •

- DNN and HMM architecture, improved training are priorities, not performance
- not focused on concurrent streams, high capacity, reliability, etc



Kaldi Info: Integration

Kaldi is its own framework

- main Kaldi contributors are working on PyTorch support
- partially supports TensorFlow, but main contributors no longer working on it
- no support for Caffe, MXnet, etc

• To integrate Kaldi into production applications takes effort

- developer interface is based on Linux shell scripts, so we tracked inference scripts + binaries, to find necessary APIs supported by inferlib
- if you ask questions on kaldi-asr.org about improving performance, reducing model size, concurrent threads, etc you will get general advice only

Acceleration

- GPUs are supported by Nvidia tech personnel on kaldi-asr.org
- also seems to be the case for OpenVINO (Intel)

Kaldi Info: Architecture, DNNs

Architecture

- uses "chain" models: $DNN^1 + xMM^2$
- AM (acoustic model) recognizes phonemes
- phonemes vary depending on context, so "tri-phones" are used
- LM (language model) recognizes words as tri-phone combinations

• DNN frequency domain data

- formed by sliding FFT analysis of incoming time series data. Each FFT frame output is similar to cochlea in human ears
- groups of FFT frames form images
- successive images are called "TDNN" (time delayed DNN), similar to series of CNNs³

Training

- DNNs saved as "x-vectors" and "i-vectors"
- HMM / GMMs saved as FSTs⁴

