# Синтез речи

Лекция №2

- 1. Tacotron 2
- 2. Global Style Tokens (GST)
- 3. Multispeaker TTS
- 4. Multilanguage TTS
- 5. Heавторегрессивный TTS

### Данные для ТТС

#### Single speaker:

- LJSpeech
- HiFiTTS

#### Multi speaker:

- VCTK
- LibriTTS

#### Дополнительная разметка:

- Montreal Forced Aligner (MFA) hmm-gmm
- Reaper, pyreaper, CREPE
- Audacity
- Yin algorithm (плохой)

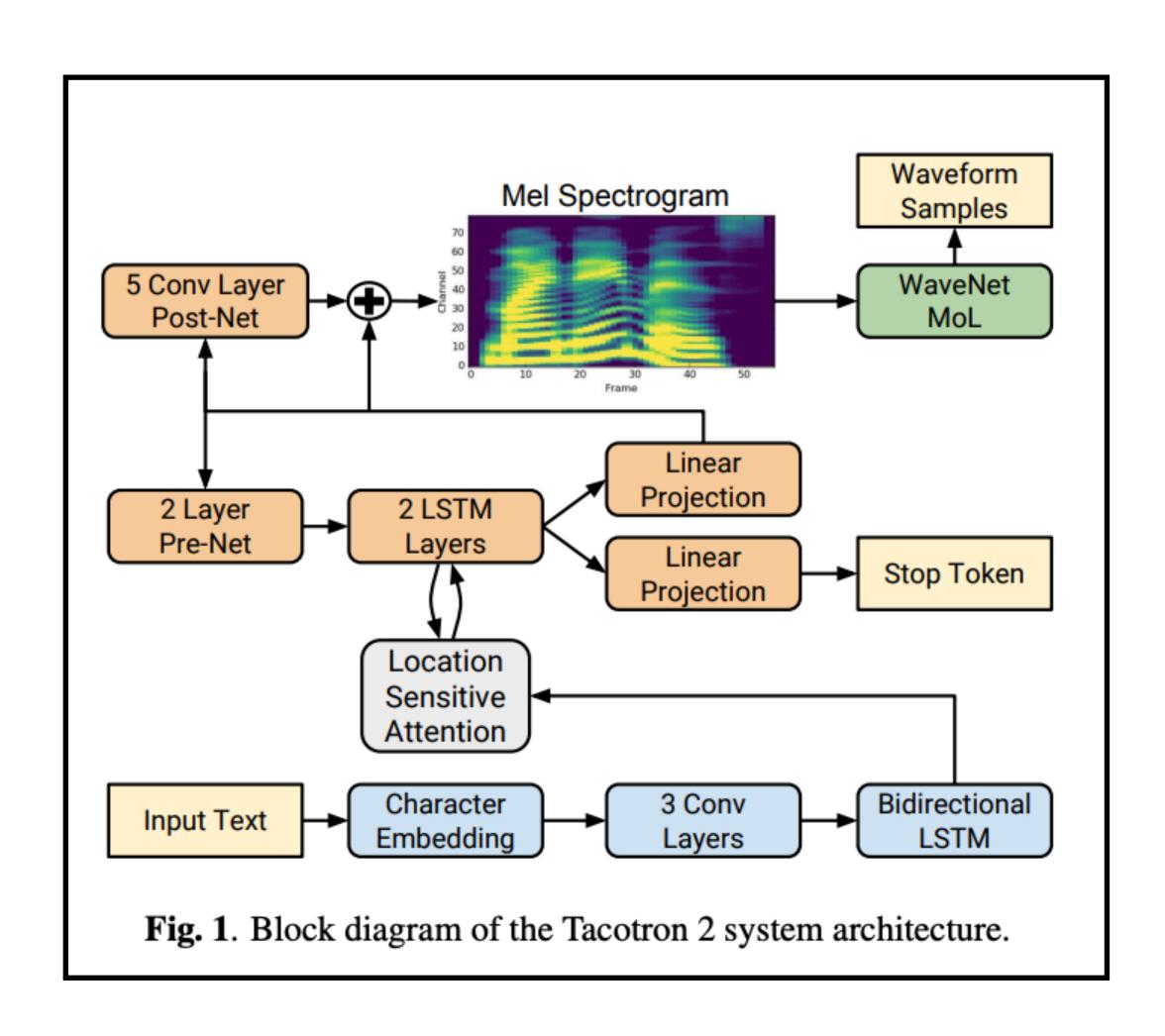
### **Tacotron 2**

seq2seq:

encoder + attention + decoder + postnet

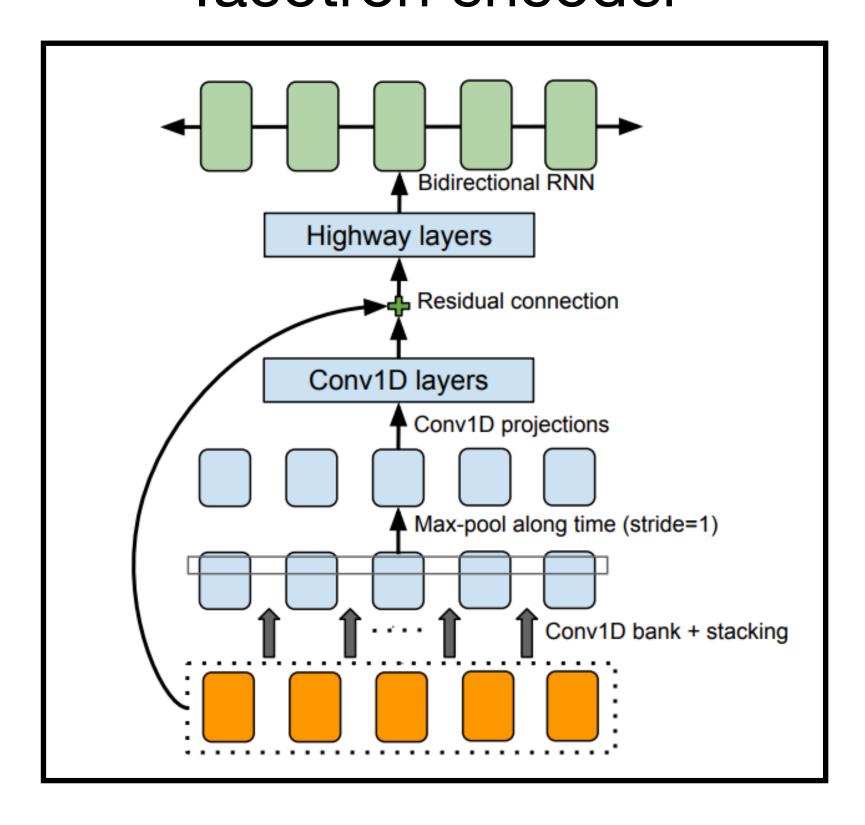
B x T x 512

+WaveNet vocoder

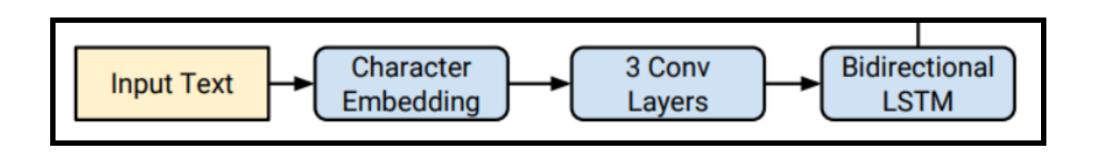


### Tacotron 2 encoder

#### Tacotron encoder



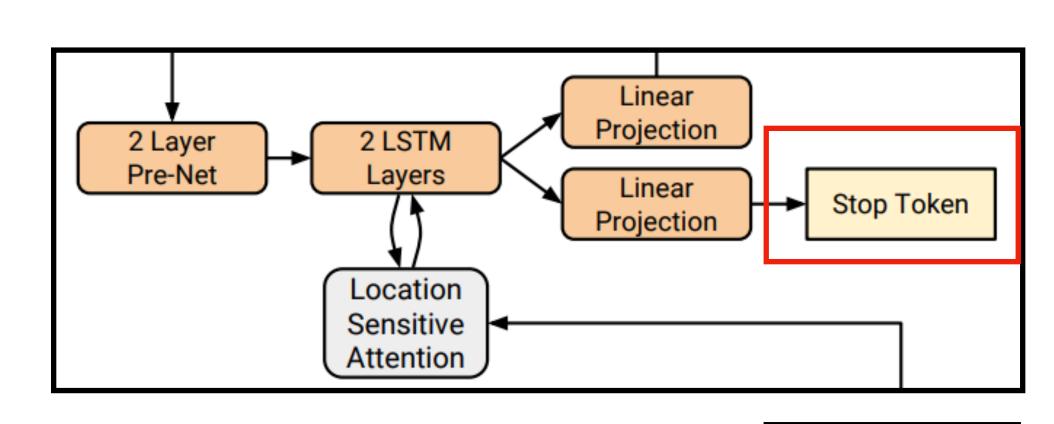
#### Tacotron 2 encoder



- нет FC после embedding слоя
- нет ResCon и Highway layers

### Tacotron 2 decoder

LSTM input = Concat(Prenet(last\_frame), context)



Teacher forcing:

encoder out

batch\_size x num\_letters x 512 -> batch\_size x 512

### **Tacotron 2 attention**

Location sensitive attention:

 $e_{i,j} = w^T anh(Ws_{i-1} + Vh_j + Uf_{i,j} + b)$ scores:

location features: |

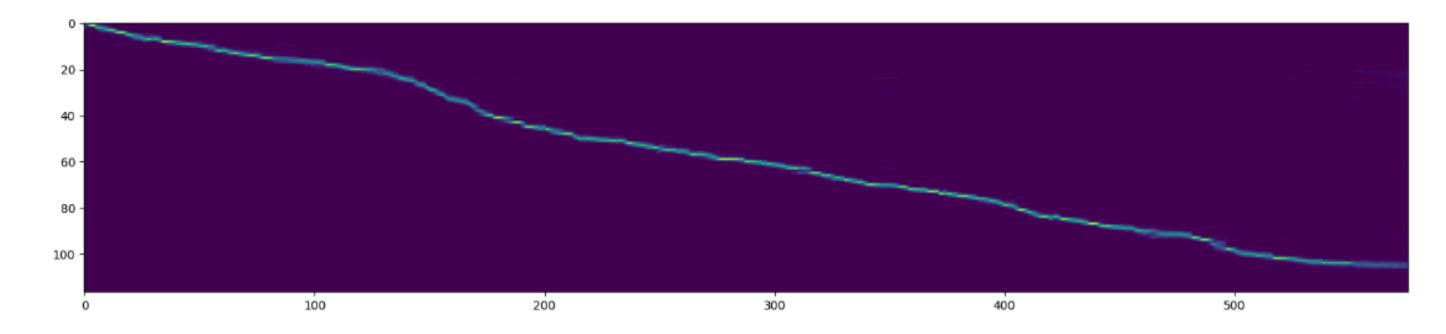
$$f_i = F * \alpha_{i-1}.$$

weights: 
$$\alpha_{ij} = \frac{\exp\left(e_{ij}\right)}{\sum_{k=1}^{T_x} \exp\left(e_{ik}\right)},$$

#### Bahdanau:

$$e_{i,j} = w^T anh(W s_{i-1} + V h_j + b)$$

Всем привет

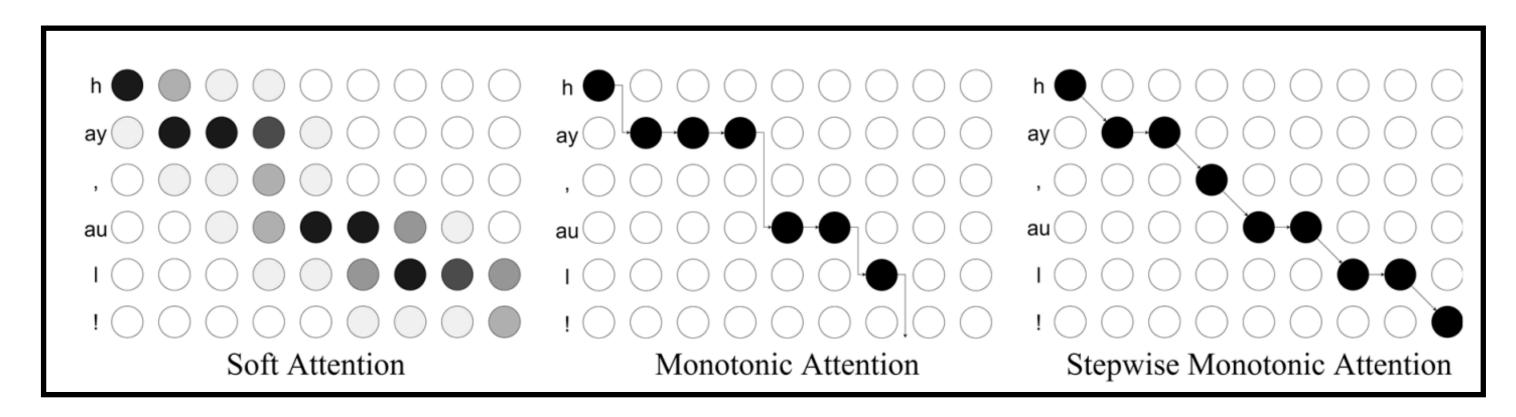


context = sum(encoder\_out \* alpha)

### Tacotron 2 другие attention механизмы

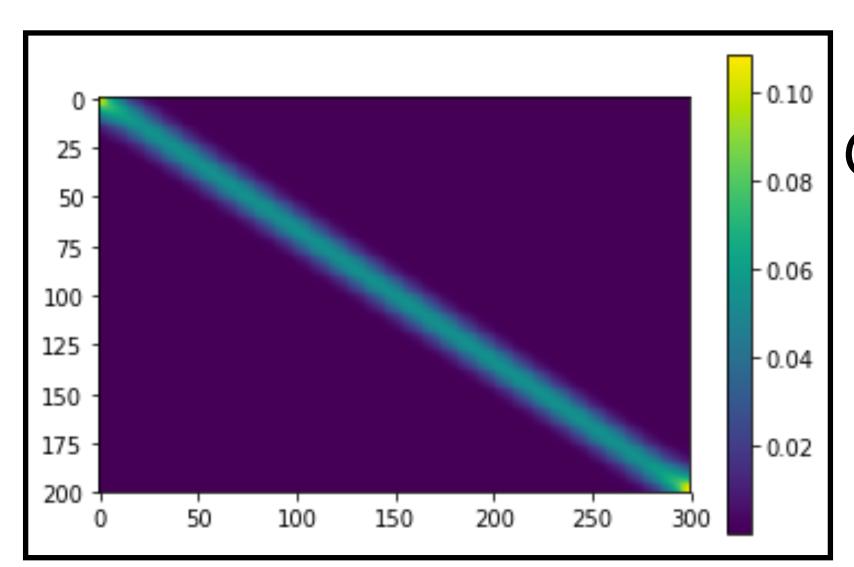
#### Решаются проблемы:

- артефакты
- сходимость
- длинные предложения



### Новые проблемы:

- энкодер учится хуже
- монотонность речи
- контекст вектор локальный



Guided attention

### Tacotron 2 inference

#### Dropout:

«In order to introduce output variation at inference time, dropout with probability 0.5 is applied only to layers in the pre-net of the autoregressive decoder»

$$D\left[\sum_{i=1}^n c_i X_i
ight] = \sum_{i=1}^n c_i^2 D[X_i] + 2\sum_{1\leqslant i < j \leqslant n} c_i c_j \operatorname{cov}(X_i, X_j),$$

$$D[X] = \mathbb{E}[X^2] - (\mathbb{E}[X])^2$$

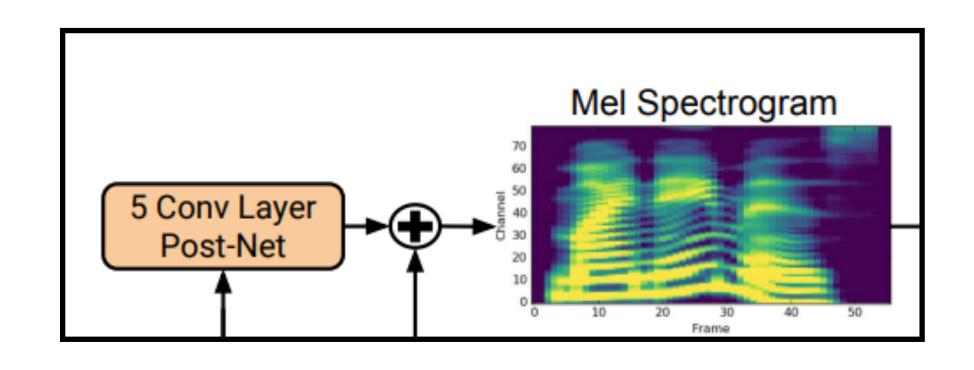
- без дропаута плохо говорит
- синтез каждый раз разный

#### Vocoder fine-tune:

1.

Training	Synthesis	
	Predicted	Ground truth
Predicted	$4.526 \pm 0.066$	$4.449 \pm 0.060$
Ground truth	$4.362 \pm 0.066$	$4.522 \pm 0.055$

2



# Просодия

Речь = кто + что + как

q-Taco

как = высокоуровневая просодия + низкоуровневая просодия

- ЭМОЦИЯ
- громкость
- скорость
- TOH

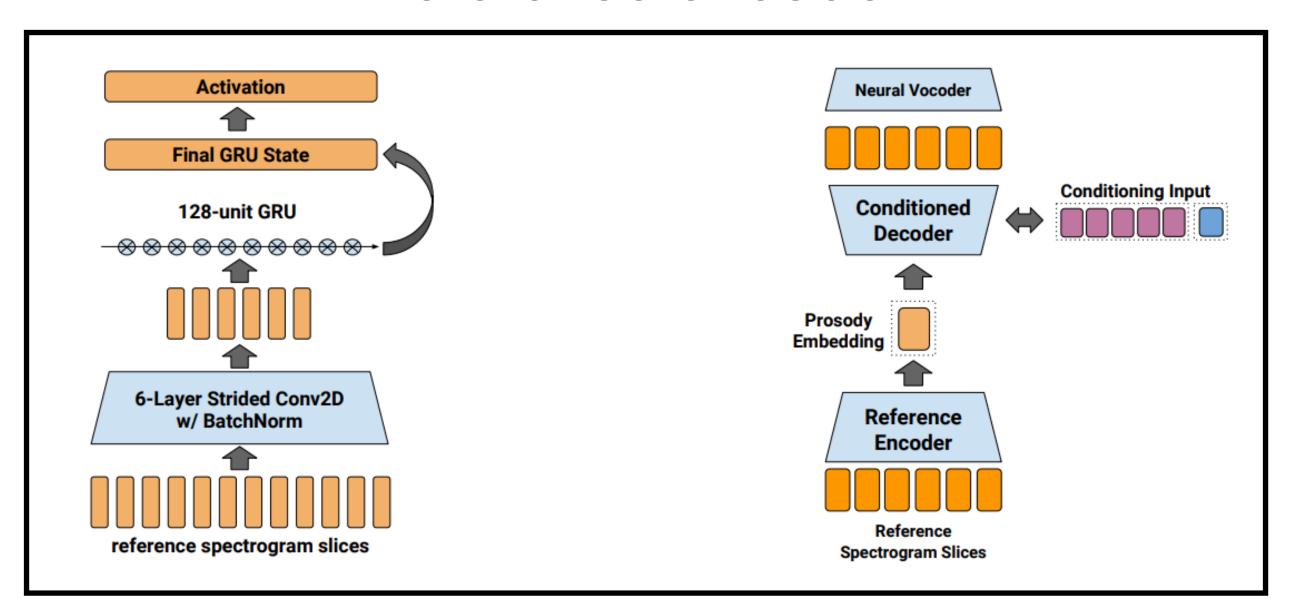
- эмфаза
- вопросы
- паузы

Будешь чай? Будешь чай?

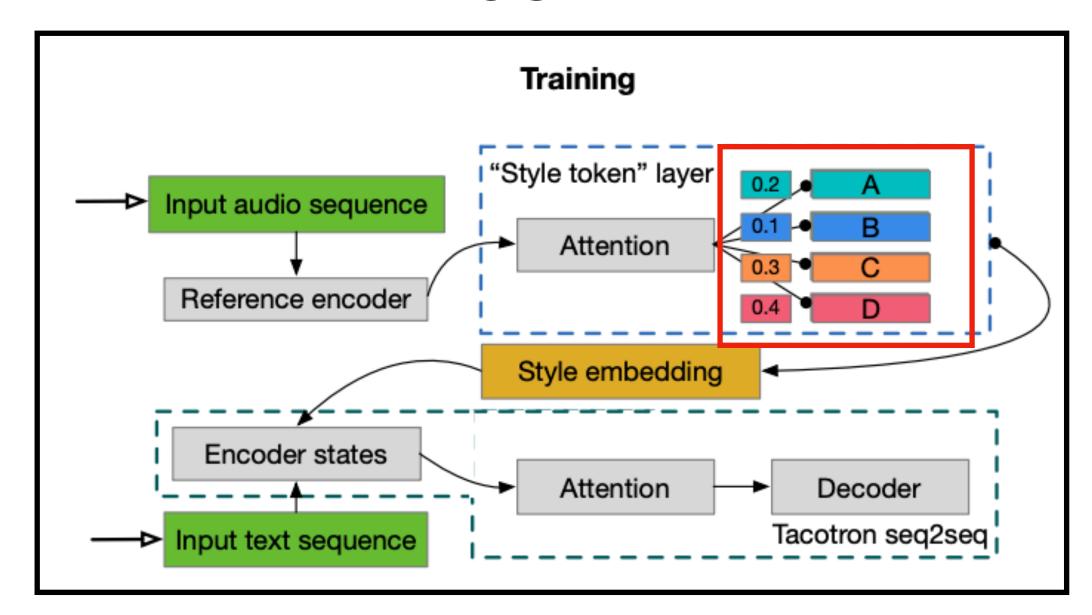
### Global Style Tokens

Blizzard challenge 2013

#### Reference encoder:



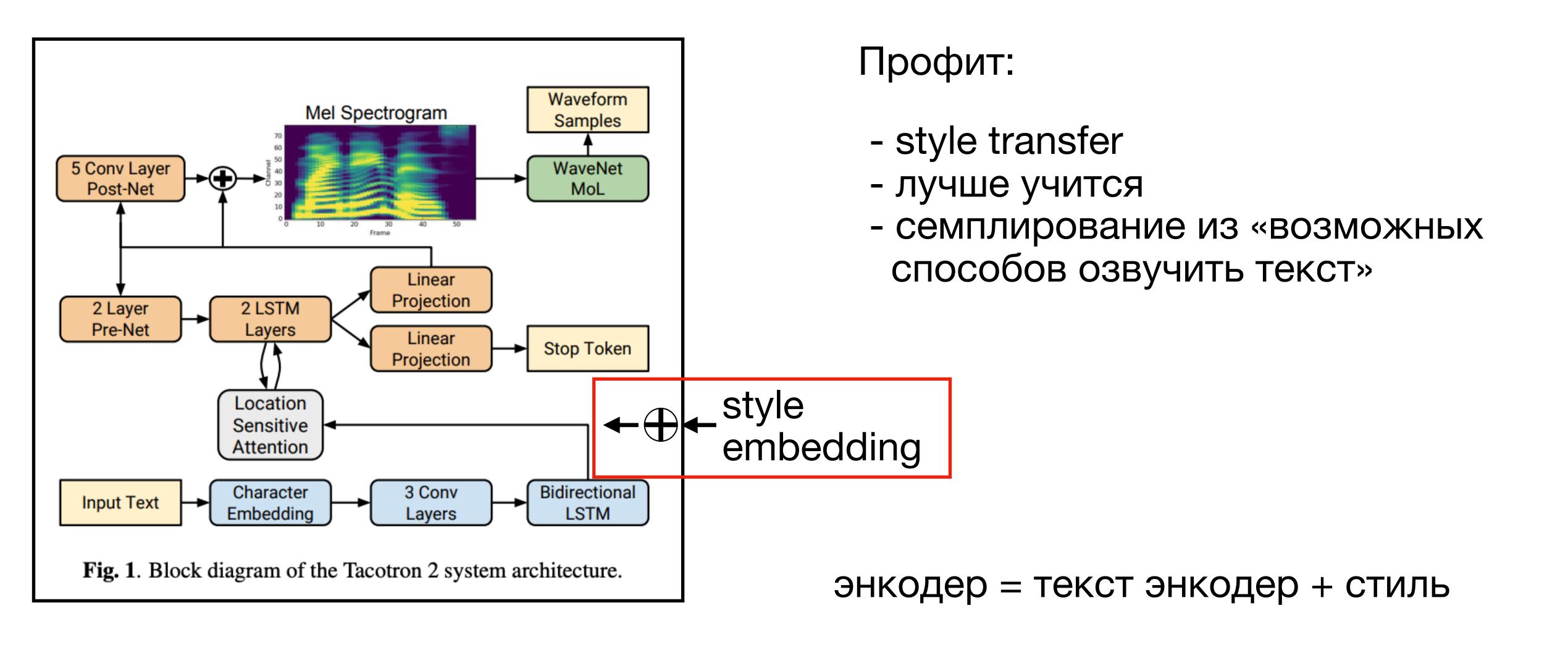
GST:



ground truth spec -> style vector

ground truth spec -> mixture of styles

### Global Style Tokens

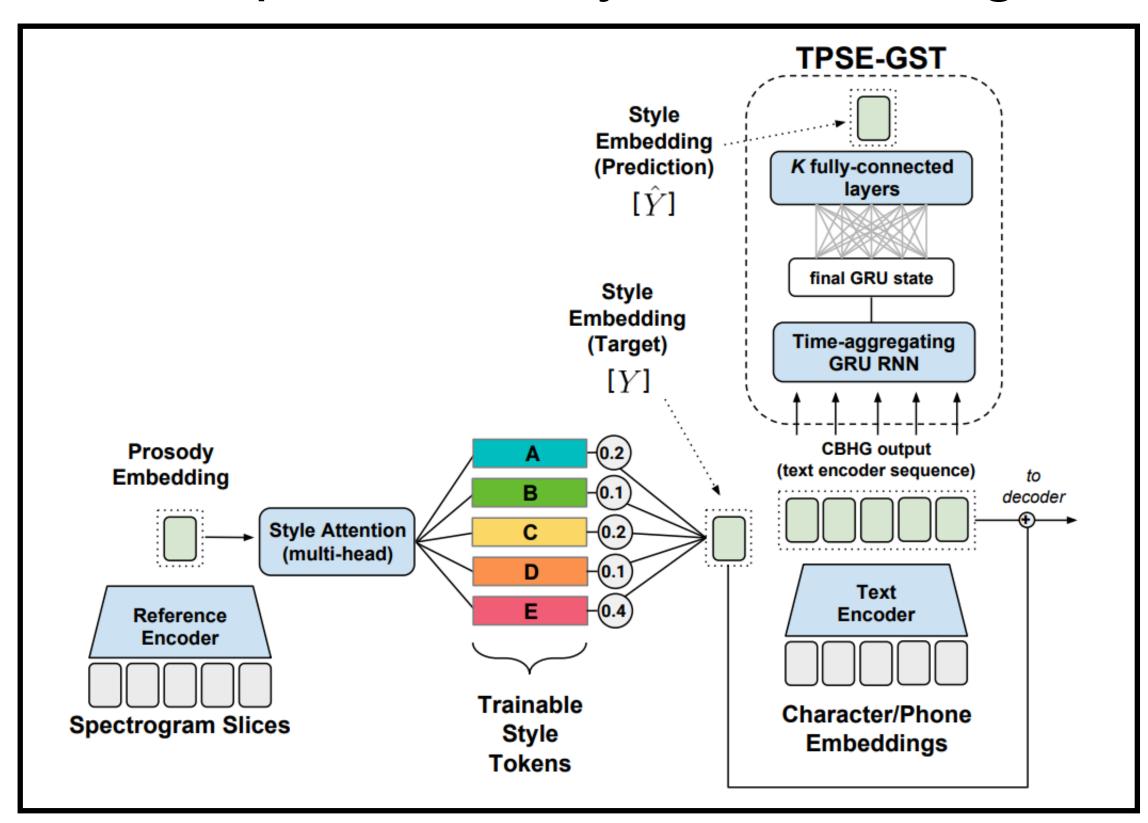


### Global Style Tokens

#### Проблемы:

- не воспроизводится :)
- стиль выучивает длину, громкость и тон
- не интерпретируется
- неоткуда брать референс

Text predicted style embedding:



Задачи:

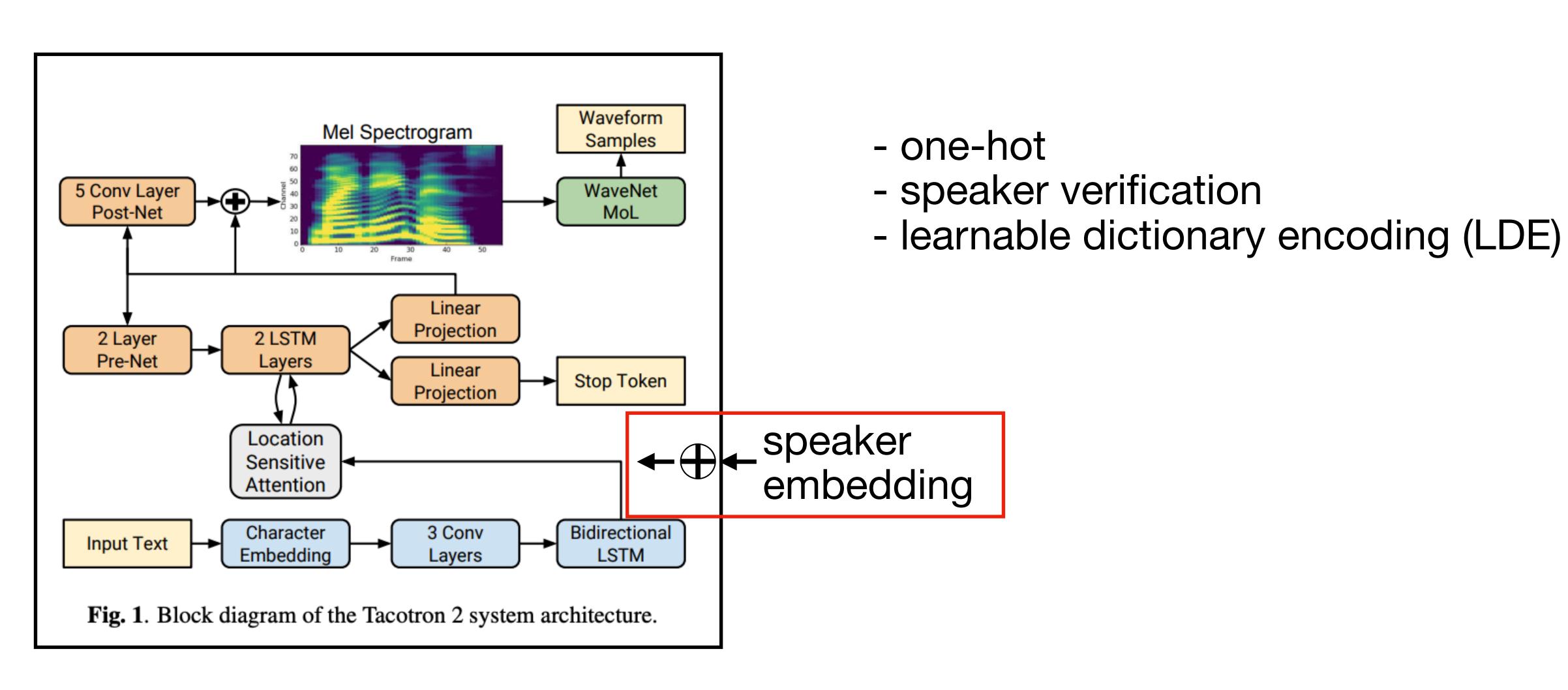
seen2seen - обычный ms tts

seen2unseen - thisvoicedoesnotexist

unseen2unseen - zero shot voice transfer

Схема решения:

конкат speaker embedding к энкодеру

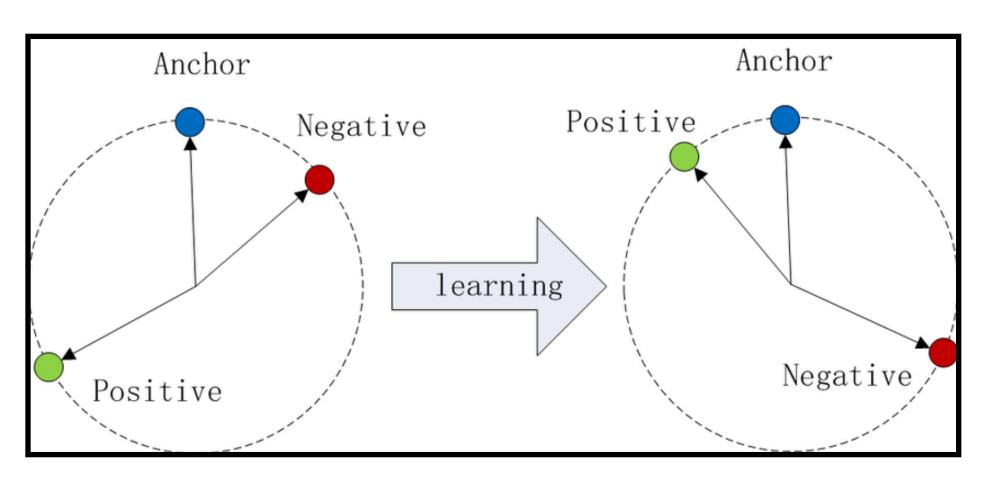


one-hot speaker embedding:

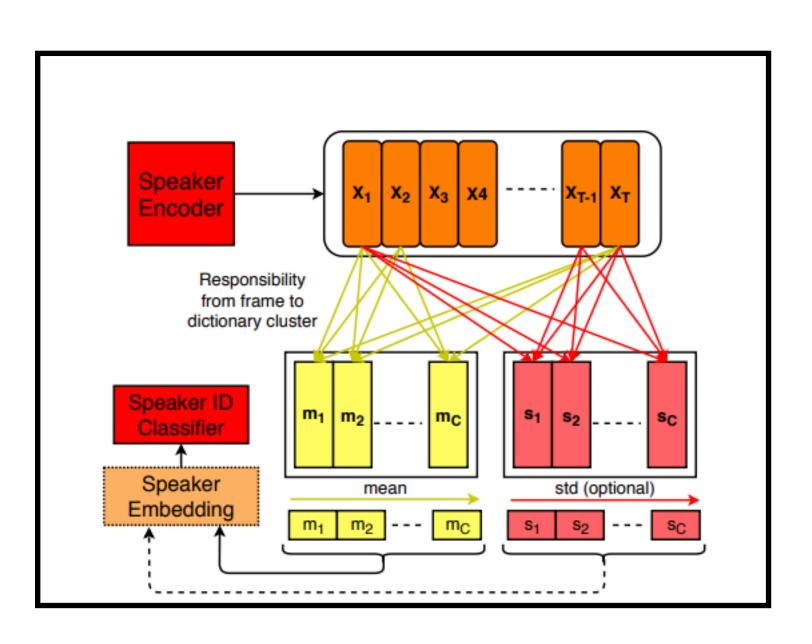
[num\_speakers x embedding\_dim]

speaker verification:

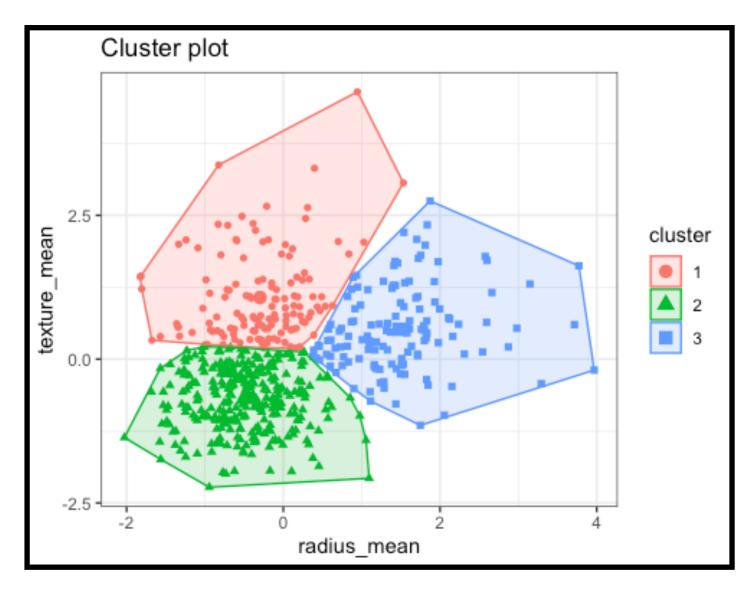
- benchmark: VoxCeleb (1, 2)
- transfer learning
- metric learning



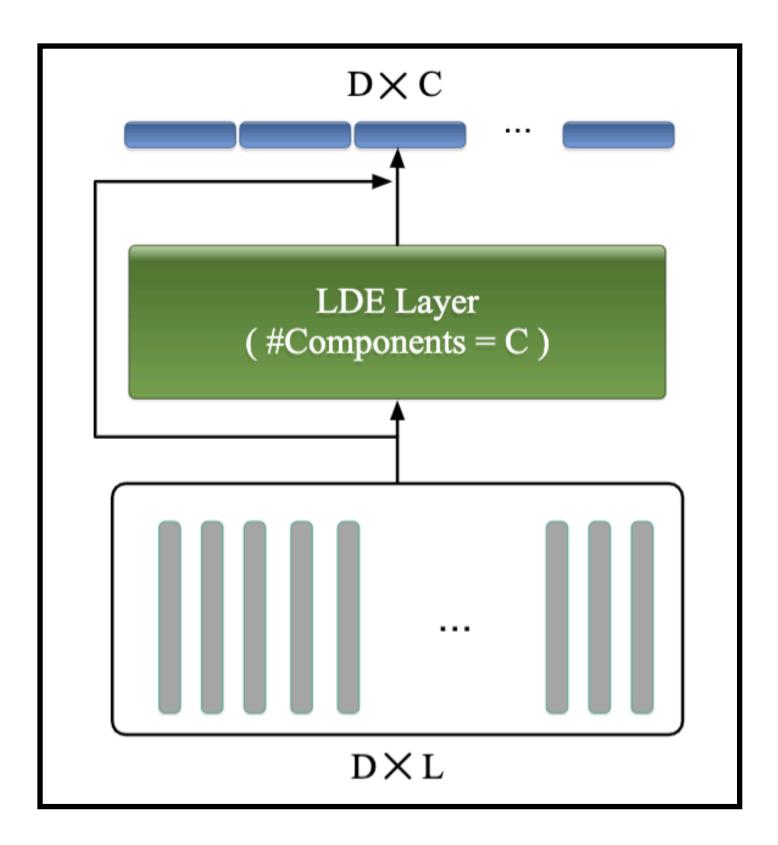
speaker embedding = mixture of encodings



#### k-means



#### Vanilla LDE:



# Multilanguage TTS

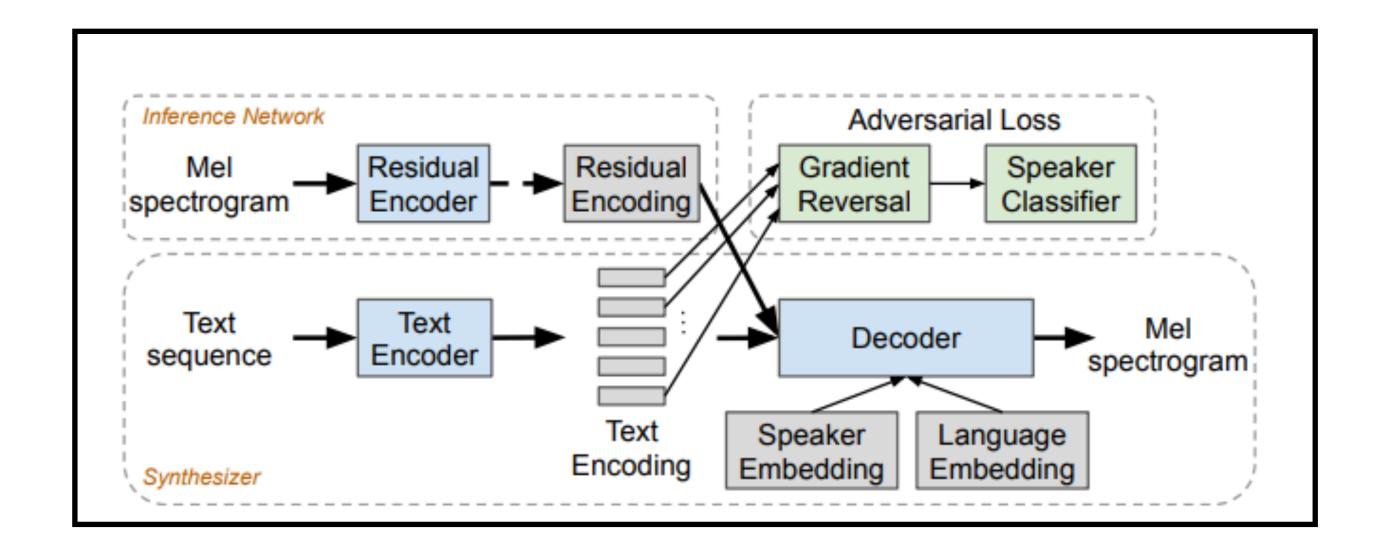
Speaking fluently in foreign language

train: inference:

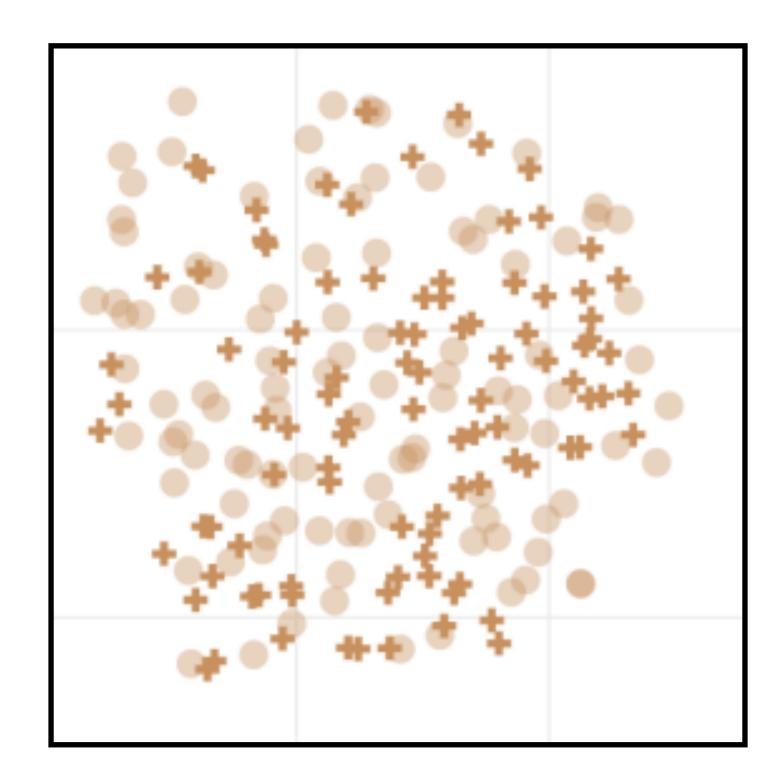
X, en

Y, ru

Y, en



«привет мир» «hello world»



# Non-autoregressive TTS

Fastspeech-family: fastpitch fastspeech-2 transformer + duration predictor

 $\mathcal{H}_{mel}$ Linear Layer Training FFT Block Add & Norm MSE Loss Positional Conv1D Duration Extractor Length Regulator Add & Norm Linear Layer Duration Predictor Conv1D + Norm FFT Block Multi-Head Attention Autoregressive Positional Conv1D + Norm Transformer TTS Phoneme Embedding Phoneme

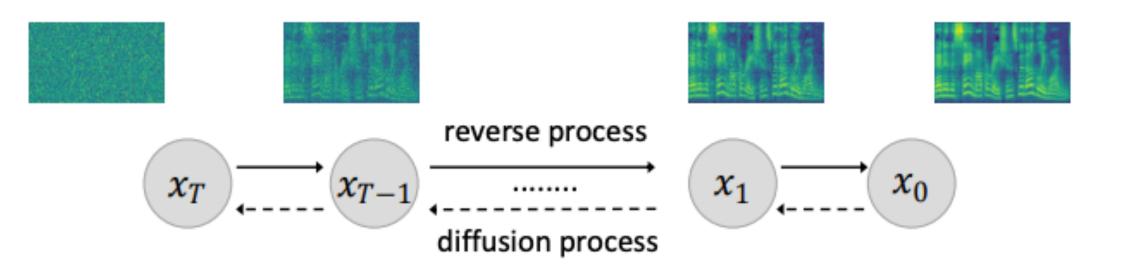
(a) Feed-Forward Transformer

(b) FFT Block

(c) Length Regulator

(d) Duration Predictor

Diffusion-based:



# Спасибо:)