

RELEVANT LINKS



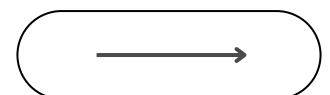
DATE

24/03/2025

CS217: ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

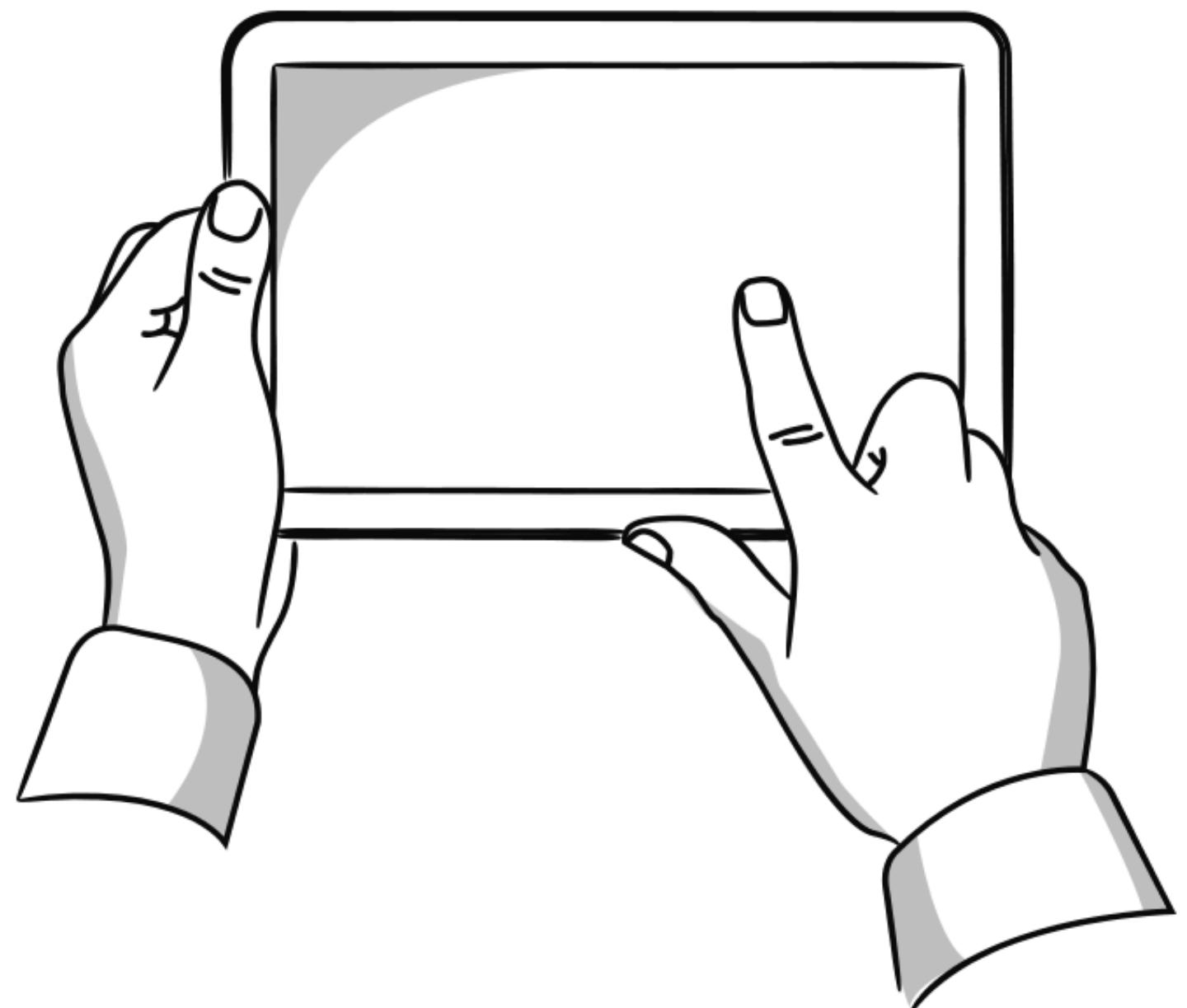
Introduction to Speech Recognition

presented by: Darshan Prabhu



Chapter 1: What is Automatic Speech Recognition?

Definition, Challenges, History and Evaluation Metrics



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What is ASR ?

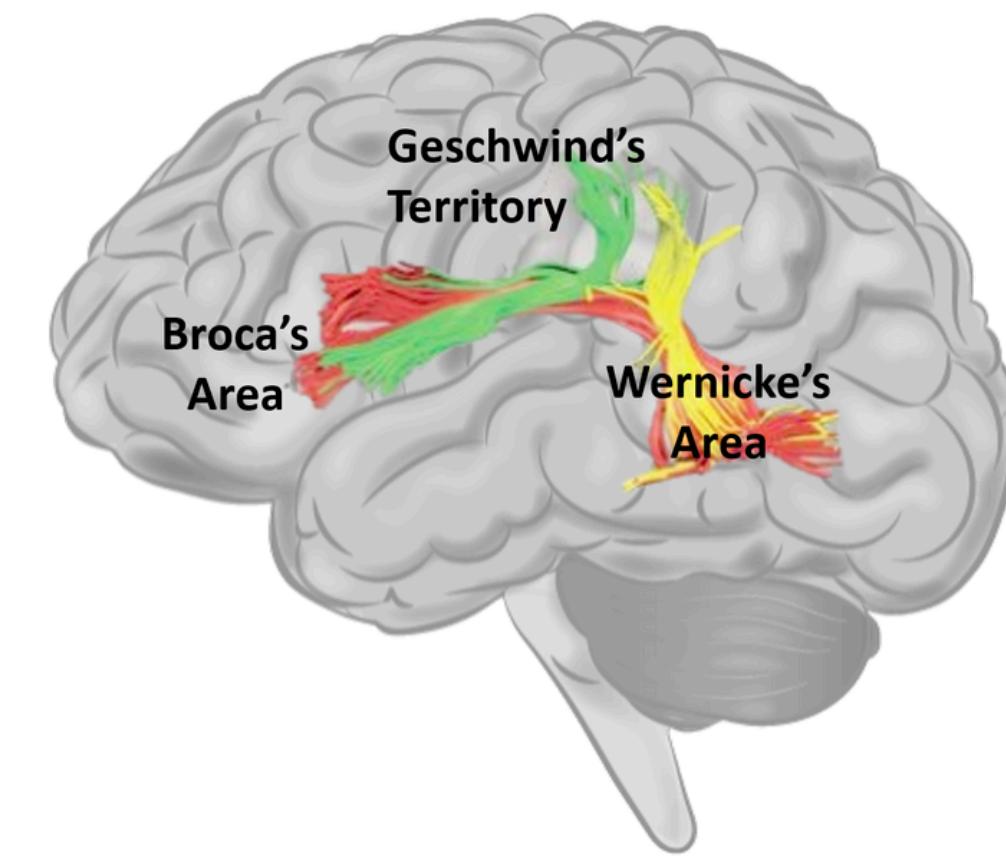


What is ASR ?





What is ASR ?



Wernicke's area, located in the posterior segment of the superior temporal gyrus



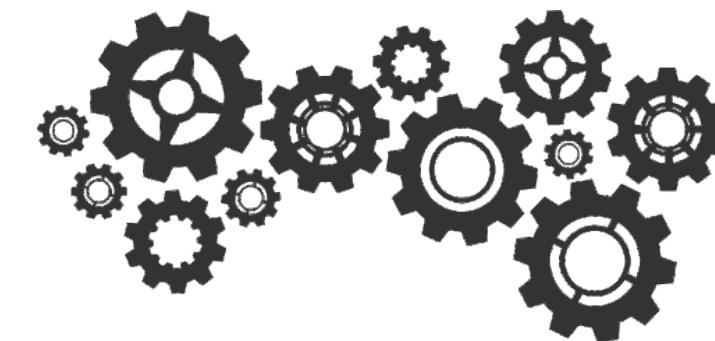
What is ASR ?

- A task of automatically converting the **speech signal** into **words**.



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WHY SO
SERIOUS?



What is ASR ?

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- The recognized words can be
 - the final output, or
 - the input to Natural Language Processing



What is ASR ?

- A task of automatically converting the **speech signal** into **words**.
- The recognized words can be
 - the **final output**, or
 - the **input** to **Natural Language Processing**
- Downstream applications of ASR
 - Spoken language understanding
 - Spoken translation
 - Intelligent video editing
 - ASR from brain signals
 - ASR for speakers with speech pathologies

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Main Challenges



Main Challenges

1

Speaker's Influence: Accent or Dialect variations, Non-native speakers, Disfluencies etc



Main Challenges

1

Speaker's Influence: Accent or Dialect variations, Non-native speakers, Disfluencies etc

2

Environmental factors: Background noise, Co-articulation, Reverberation etc



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Speaker's Influence: Accent or Dialect variations, Non-native speakers, Disfluencies etc

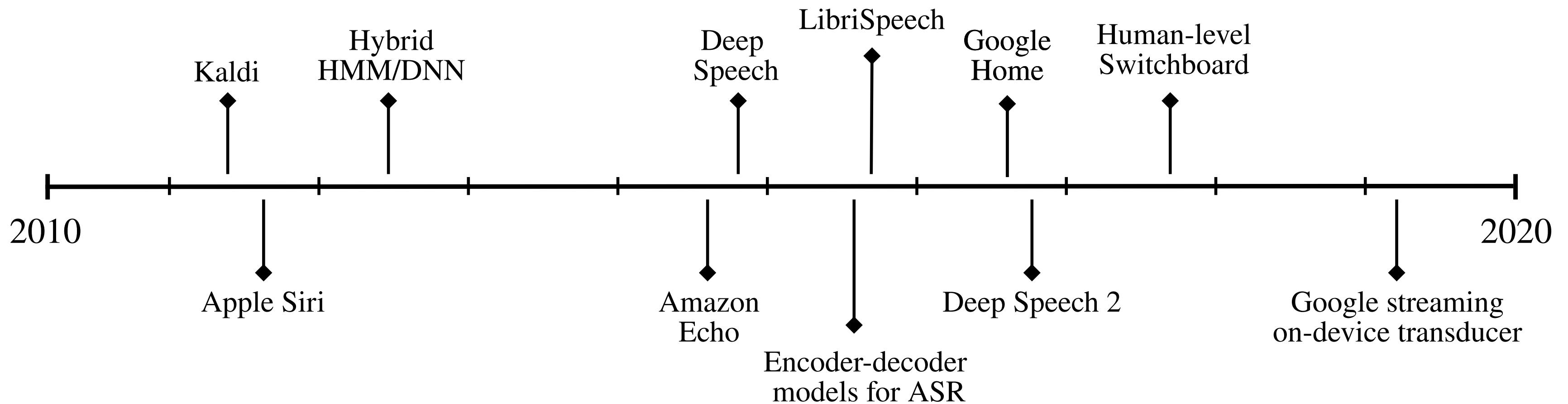
2

Environmental factors: Background noise, Co-articulation, Reverberation etc

3

Data related problems: Microphone quality, Lack of diversity etc

A historical perspective of ASR





A historical perspective of ASR (continued)



The 🎉 Open ASR Leaderboard ranks and evaluates speech recognition models on the Hugging Face Hub.
 We report the Average WER (⬇️ lower the better) and RTFx (⬆️ higher the better). Models are ranked based on their Average WER, from lowest to highest. Check the ✅ Metrics tab to understand how the models are evaluated.
 If you want results for a model that is not listed here, you can submit a request for it to be included 📩.

The leaderboard currently focuses on English speech recognition, and will be expanded to multilingual evaluation in later versions.

Leaderboard Metrics Request a model here!

model	Average WER ⬇️	RTFx ⬆️	AMI	Earnings22	Gigaspeech	LS Clean	LS Other	SPGISpeech	Tedium
microsoft/Phi-4-multimodal-instruct	6.14	62.12	11.45	10.5	9.77	1.67	3.82	3.11	2.89
nvidia/canary-1b-flash	6.35	1045.75	13.11	12.77	9.85	1.48	2.87	1.95	3.12
nvidia/canary-1b	6.5	235.34	13.9	12.19	10.12	1.48	2.93	2.06	3.56
nyrahealth/CrisperWhisper	6.67	84.05	8.71	12.89	10.24	1.82	4	2.7	3.2
nvidia/parakeet-tdt-1.1b	7.01	2390.61	15.87	14.49	9.52	1.4	2.6	3.16	3.59
nvidia/parakeet-rnnt-1.1b	7.12	2053.15	17.01	13.94	9.89	1.45	2.5	2.93	3.83
nvidia/canary-180m-flash	7.12	1233.58	14.86	12.33	10.51	1.73	4.35	2.26	3.13
efficient-speech/lite-whisper-large-v3-acc	7.23	117.8	16.1	11.04	10.1	2	3.91	2.89	3.71
nvidia/parakeet-ctc-1.1b	7.4	2728.52	15.67	13.75	10.28	1.83	3.51	4.02	3.57
efficient-speech/lite-whisper-large-v3	7.43	115.83	16.9	11.55	10.26	2.1	4.4	2.85	3.73
openai/whisper-large-v3	7.44	145.51	15.95	11.29	10.02	2.01	3.91	2.94	3.86
nvidia/parakeet-tdt_ctc-110m	7.49	5345.14	15.89	12.37	10.52	2.4	5.22	2.54	4.07



Evaluation Metrics

Reference: I want to go to the cse office

Prediction: I want to go **see a** office



Evaluation Metrics

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Method 1: Sentence Error Rate

An entire sentence is either correct or not.

- 100% error rate in the case above.

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Evaluation Metrics

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Prediction: I want to go **see a** office

Method 1: Sentence Error Rate

An entire sentence is either correct or not.

- 100% error rate in the case above.
- **Problem:** Too strict. Need to consider some measure of local correctness.



Evaluation Metrics

Reference: I want to go to the cse office

Prediction: I want to go **see a** *** office

Method 2: Word Error Rate (WER)

insertion errors = 0, # substitution errors = 2, # of deletion errors = 1 → Edit distance = 3

Word Error Rate (%): Edit distance (=3) / # reference words (=8) * 100 = 37.5 %

- Calculated using Levenshtein distance.



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- Calculated using Levenshtein distance.
- **Problem:** How to handle languages that do not have word boundaries? (Ex: Japanese)



Evaluation Metrics

Reference: hello world

Prediction: heldoo world

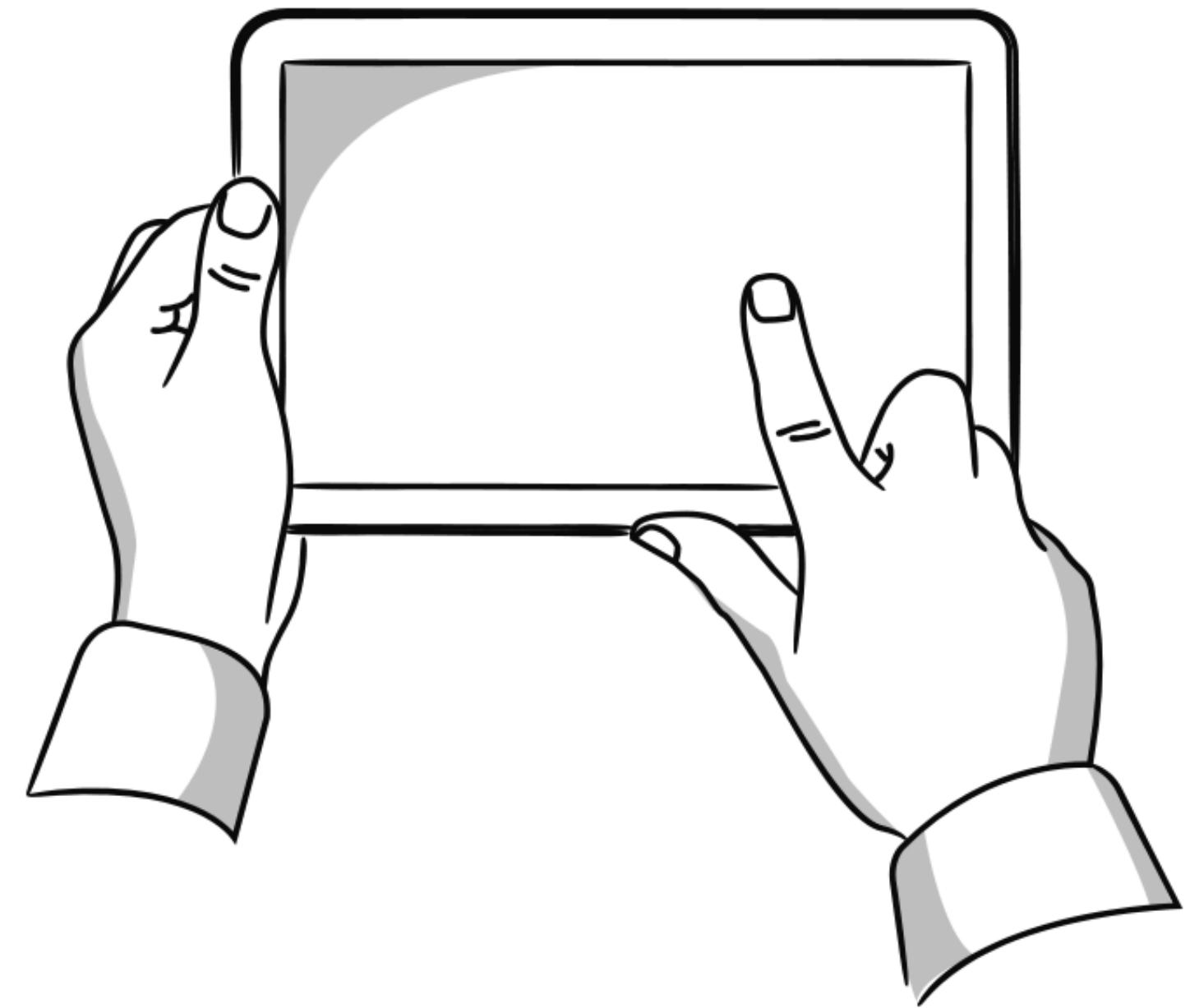
Method 3: Character Error Rate (CER)

insertion errors = 1, # substitution errors = 1, # of deletion errors = 0 → Edit distance = 2

Character Error Rate (%): $\text{Edit distance} (=2) / \# \text{ reference chars} (=10) * 100 = 20\%$

Chapter 2: Connectionist Temporal Classification

[Background](#), [Problem Setting](#) and [Formulations](#)



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Mathematical Formulation of ASR

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$$\mathbf{O} = \{O_1, O_2, \dots, O_T\}$$



a sequence of **acoustic features** corresponding to a **speech signal**

$$O_i \in \mathbb{R}^d$$



d-dimensional acoustic feature vector and **T** is the length of the sequence

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d-dimensional acoustic feature vector and **T** is the length of the sequence

$$\mathbf{W}^* = \arg \max_{\mathbf{W}} Pr(\mathbf{W} \mid \mathbf{O})$$

$$\mathbf{W} = \{W_1, W_2, \dots, W_M\}$$



a sequence of **words** and **M** is the length of this sequence



Training of ASR

$$\sum_{(\mathbf{O}, \mathbf{y}) \in \mathcal{D}} -\log Pr(\mathbf{y} | \mathbf{O})$$



Training of ASR

$$\sum_{(\mathbf{O}, \mathbf{y}) \in \mathcal{D}} -\log Pr(\mathbf{y} | \mathbf{O})$$

Dataset 

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Input sequence Output sequence



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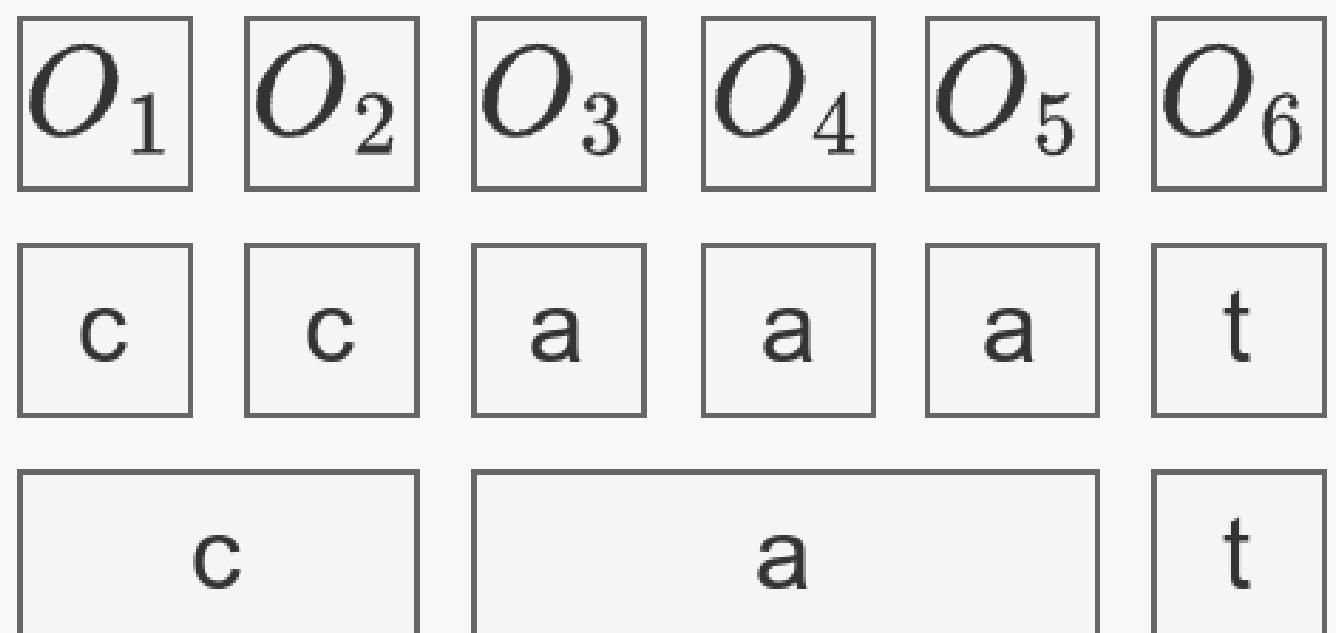
The diagram illustrates the summation term in the training equation. It shows a large summation symbol (\sum) followed by a condition $(\mathbf{O}, \mathbf{y}) \in \mathcal{D}$. Below the summation, there are two arrows pointing upwards from the labels "Input sequence" and "Output sequence" to the variables \mathbf{O} and \mathbf{y} respectively. A third arrow points upwards from the word "Dataset" to the symbol \mathcal{D} .

\mathbf{O} and \mathbf{y} are different length sequences.

How do we handle it?

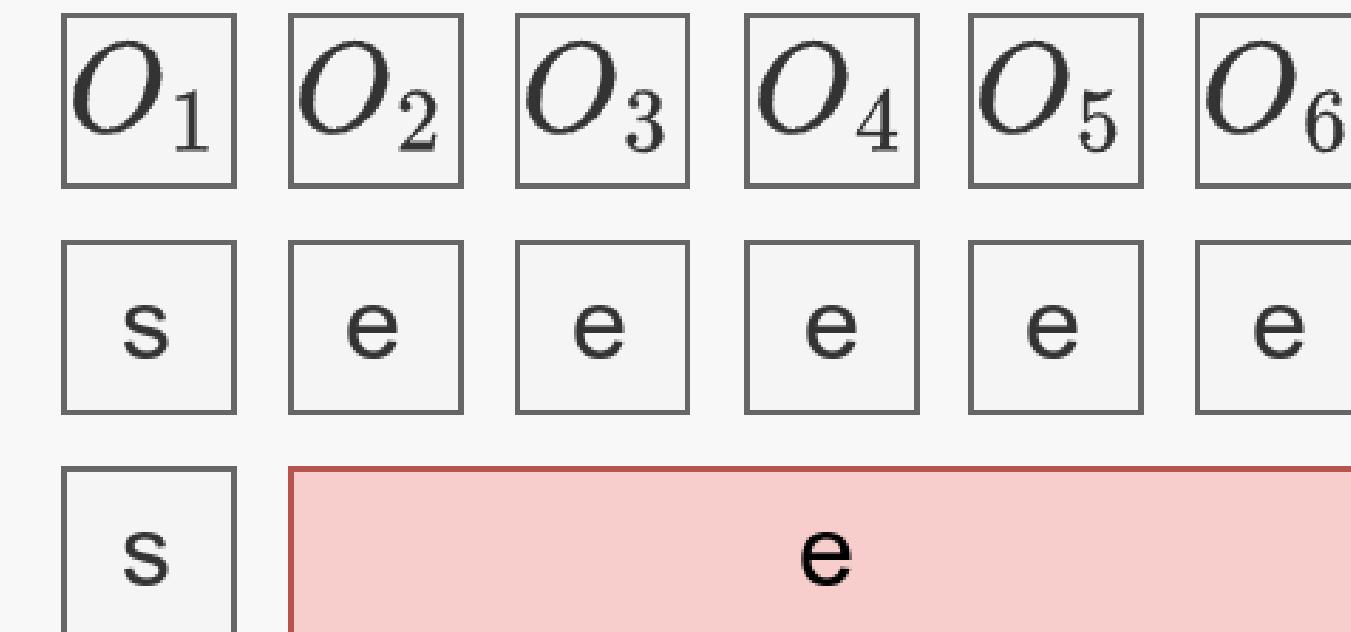
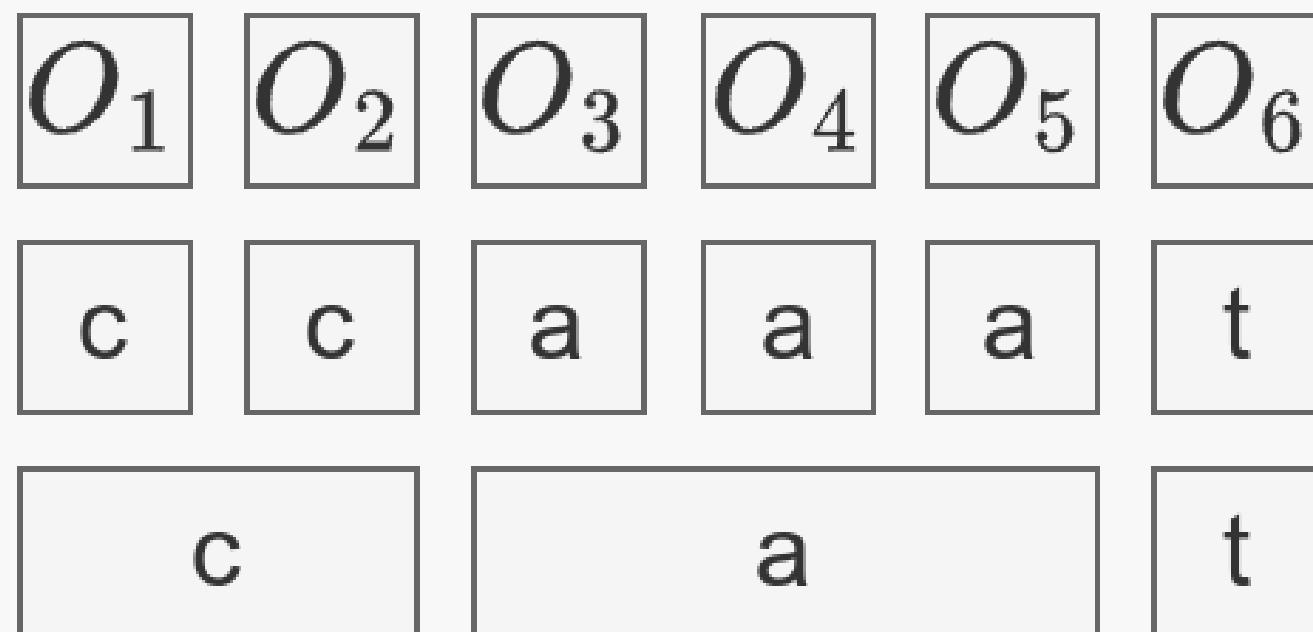
Alignment, Blank and Beta Operator

What is an alignment?



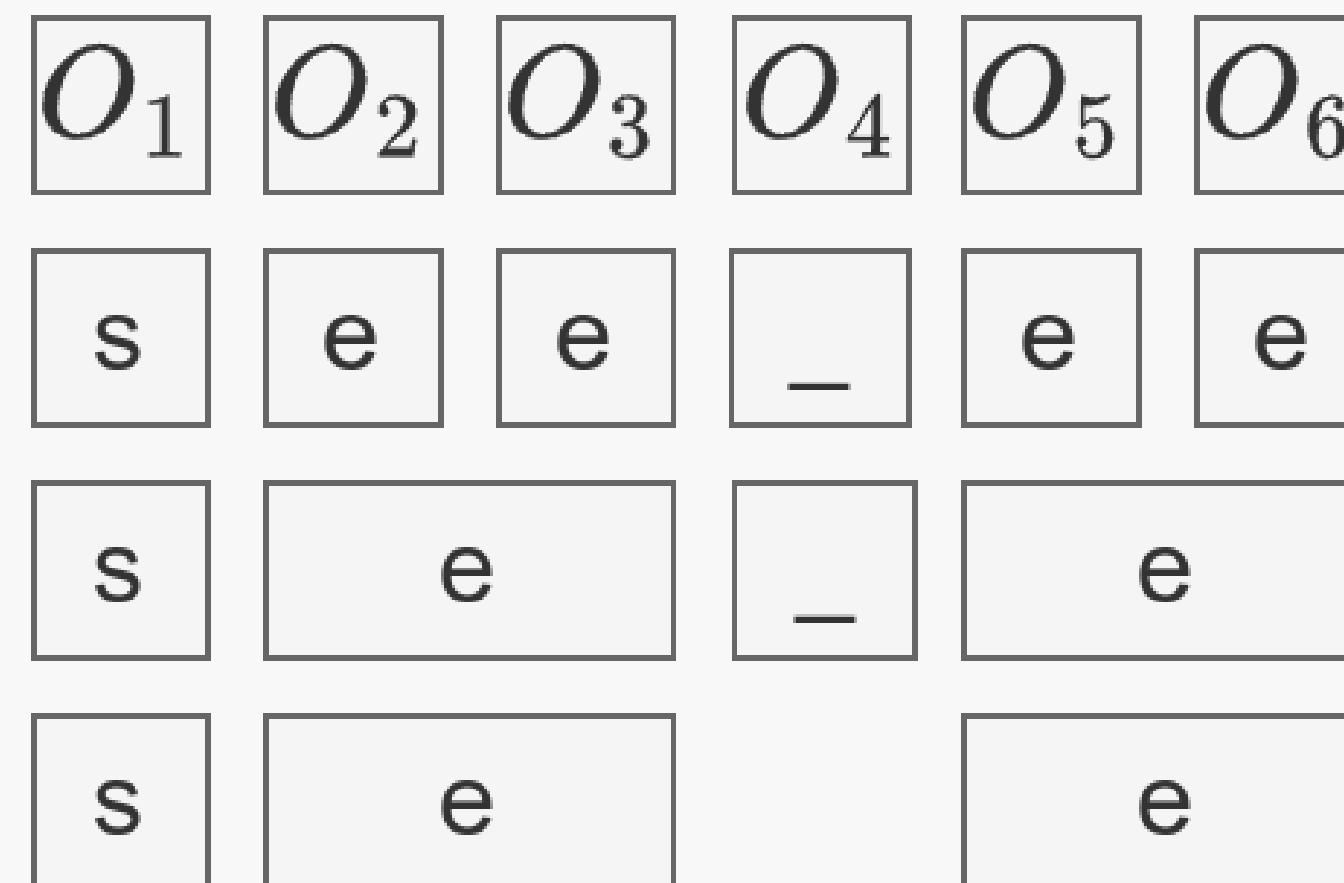
Alignment, Blank and Beta Operator

What is an alignment?



Alignment, Blank and Beta Operator

How to handle repetitions in alignments?





Alignment, Blank and Beta Operator

- **Blank symbol (_):** Added to the vocabulary. It represents “empty”.
- For a given label sequence, there can be **multiple alignments:** (x, y, z) could correspond to $(x, _, y, _, _, z)$ or $(_, x, x, _, y, z)$

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- Define a **2-step operator \mathcal{B}** that reduces a label sequence by: *first* removing repeating labels and *second* removing blanks.
 - $\mathcal{B}("x, _, y, _, _, z") = \mathcal{B}(" _, x, x, _, y, z") = "x, y, z"$

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 - $\mathcal{B}("x, _, y, _, _, z") = \mathcal{B}(" _, x, x, _, y, z") = "x, y, z"$
- $\mathcal{B}^{-1}("x, y, z") = \{"x, _, y, _, _, z", "_, x, x, _, y, z", \dots\}$ is the set of all T-length alignments that collapse to the string “ x, y, z ” on applying the operator \mathcal{B}



Rethinking the Probability

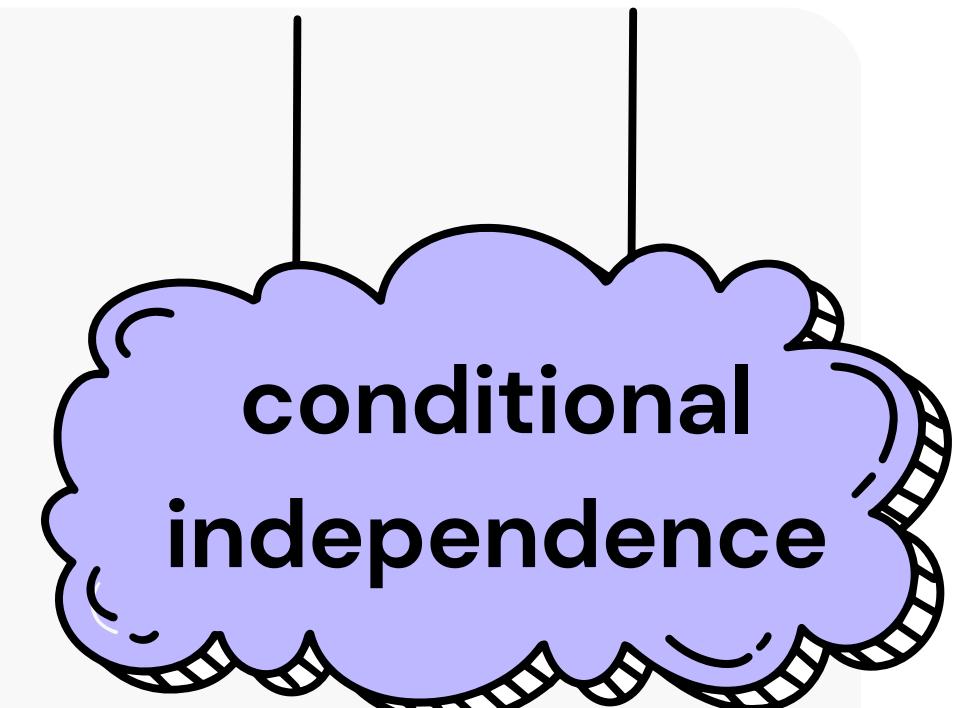
$$Pr(\mathbf{y} \mid \mathcal{O}) = \sum_{a \in \mathcal{B}^{-1}(\mathbf{y})} Pr(a \mid \mathcal{O})$$



Doubt

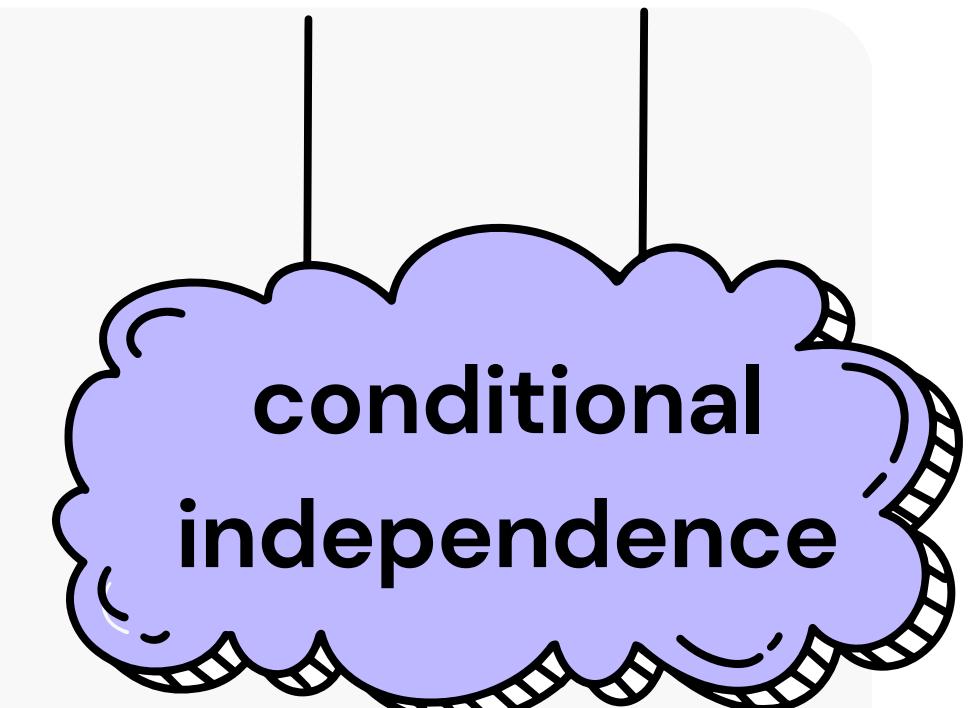
Rethinking the Probability

$$\begin{aligned} Pr(\mathbf{y} \mid \mathcal{O}) &= \sum_{a \in \mathcal{B}^{-1}(\mathbf{y})} Pr(a \mid \mathcal{O}) \\ &= \sum_{a \in \mathcal{B}^{-1}(\mathbf{y})} \prod_{t=1}^T Pr(a_t \mid \mathcal{O}) \end{aligned}$$



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marginalizes over
valid alignments

computing the probability for a
single alignment step-by-step.



Dynamic Programming

0 \emptyset

1 s

2 \emptyset

3 e

4 \emptyset

5 e

6 \emptyset

CTC
states

Dynamic Programming

0 \emptyset

1 s

2 \emptyset

3 e

4 \emptyset

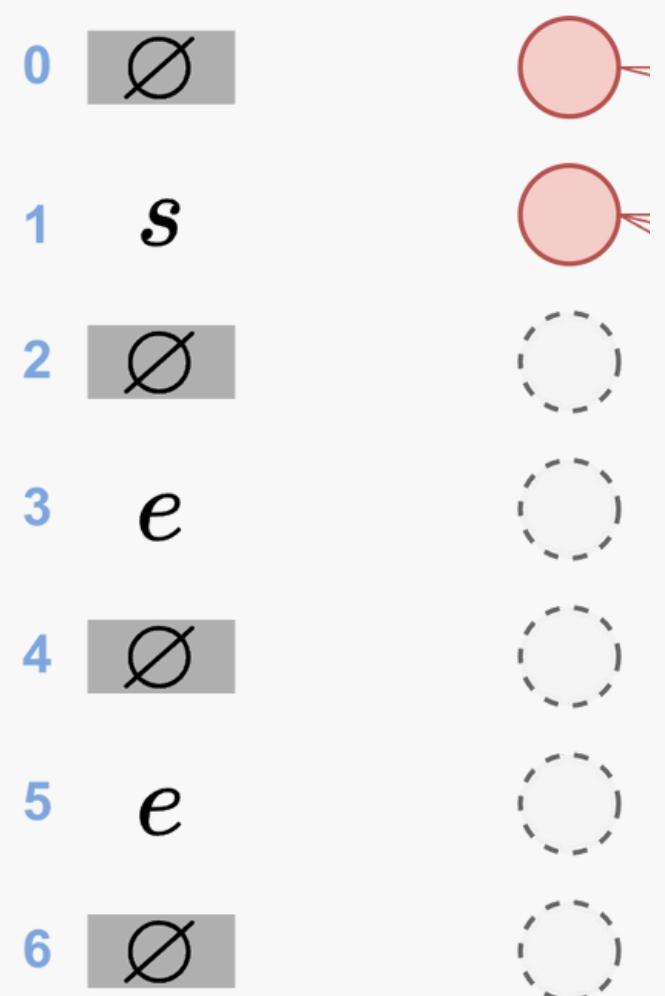
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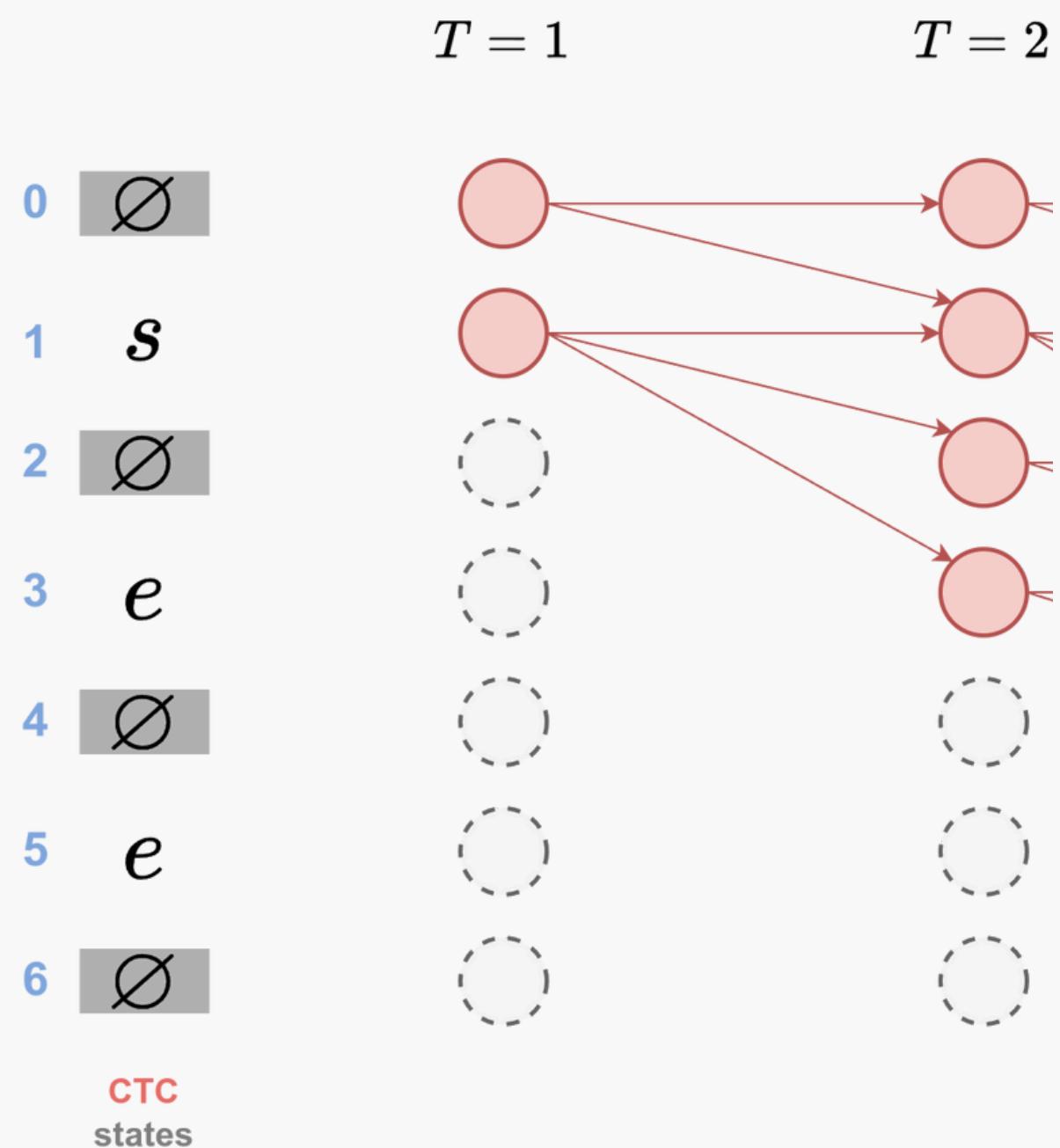
CTC
states

If the original label sequence is of length l , then
the new sequence would have $2l+1$ labels

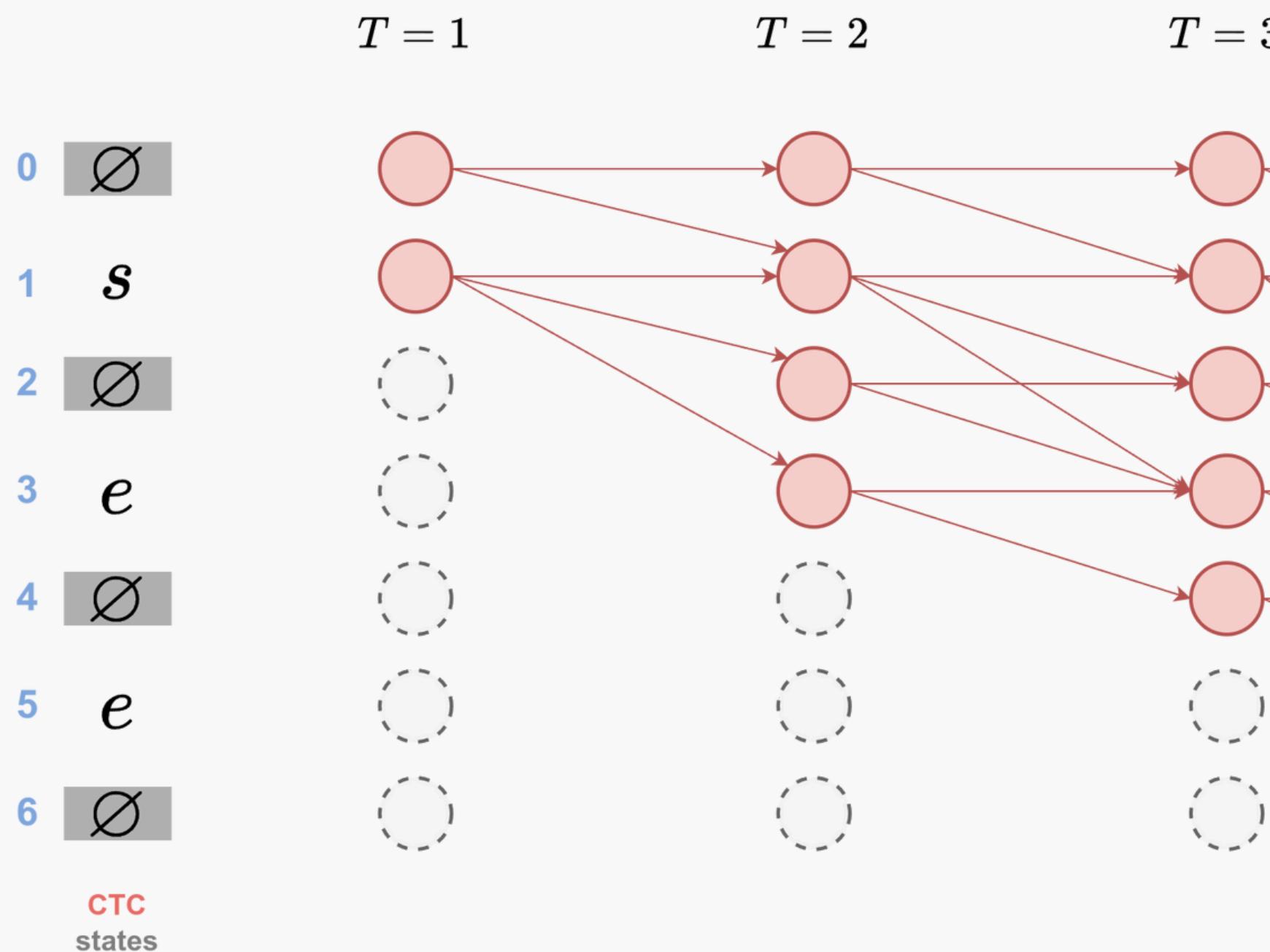
Dynamic Programming

 $T = 1$ CTC
states

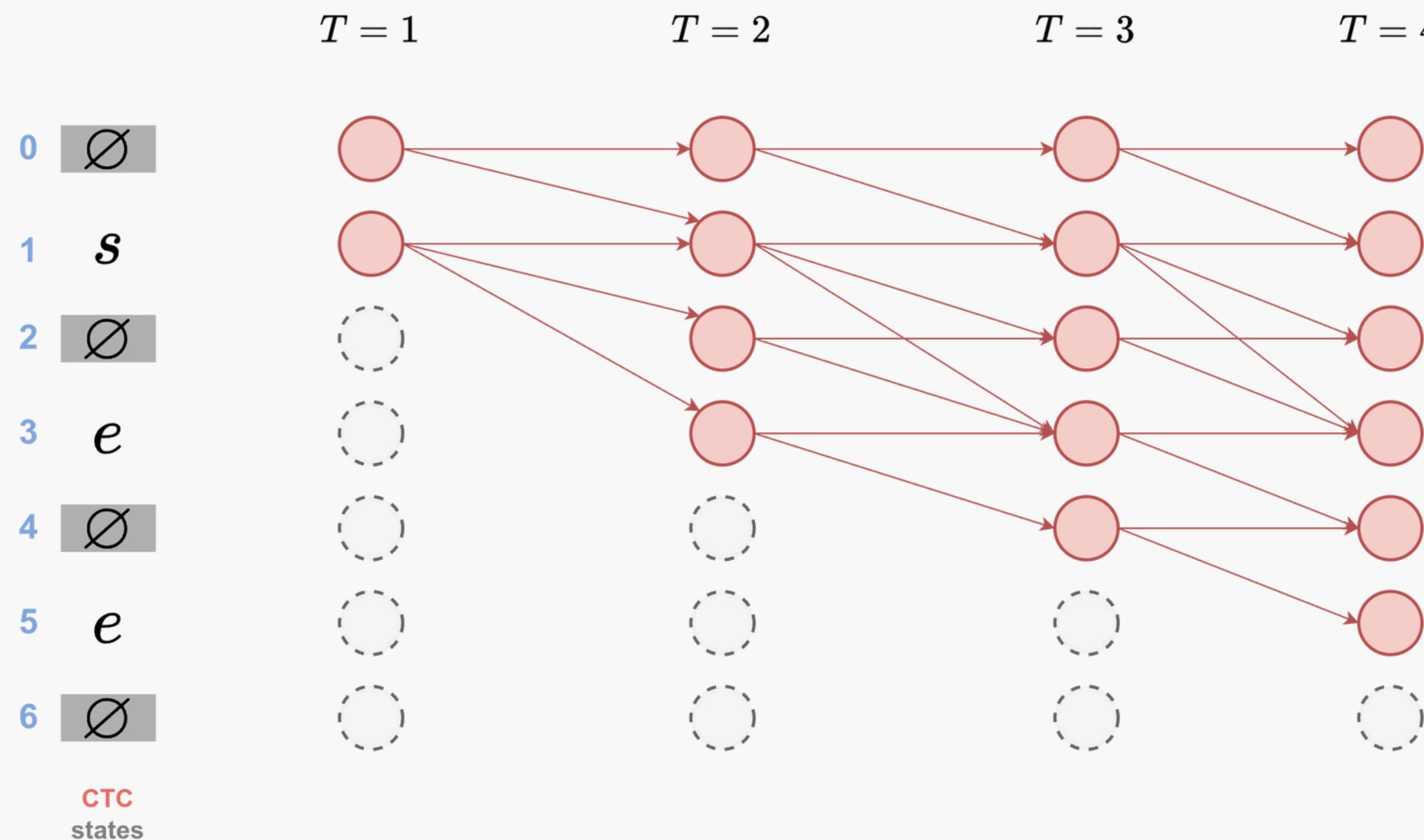
Dynamic Programming



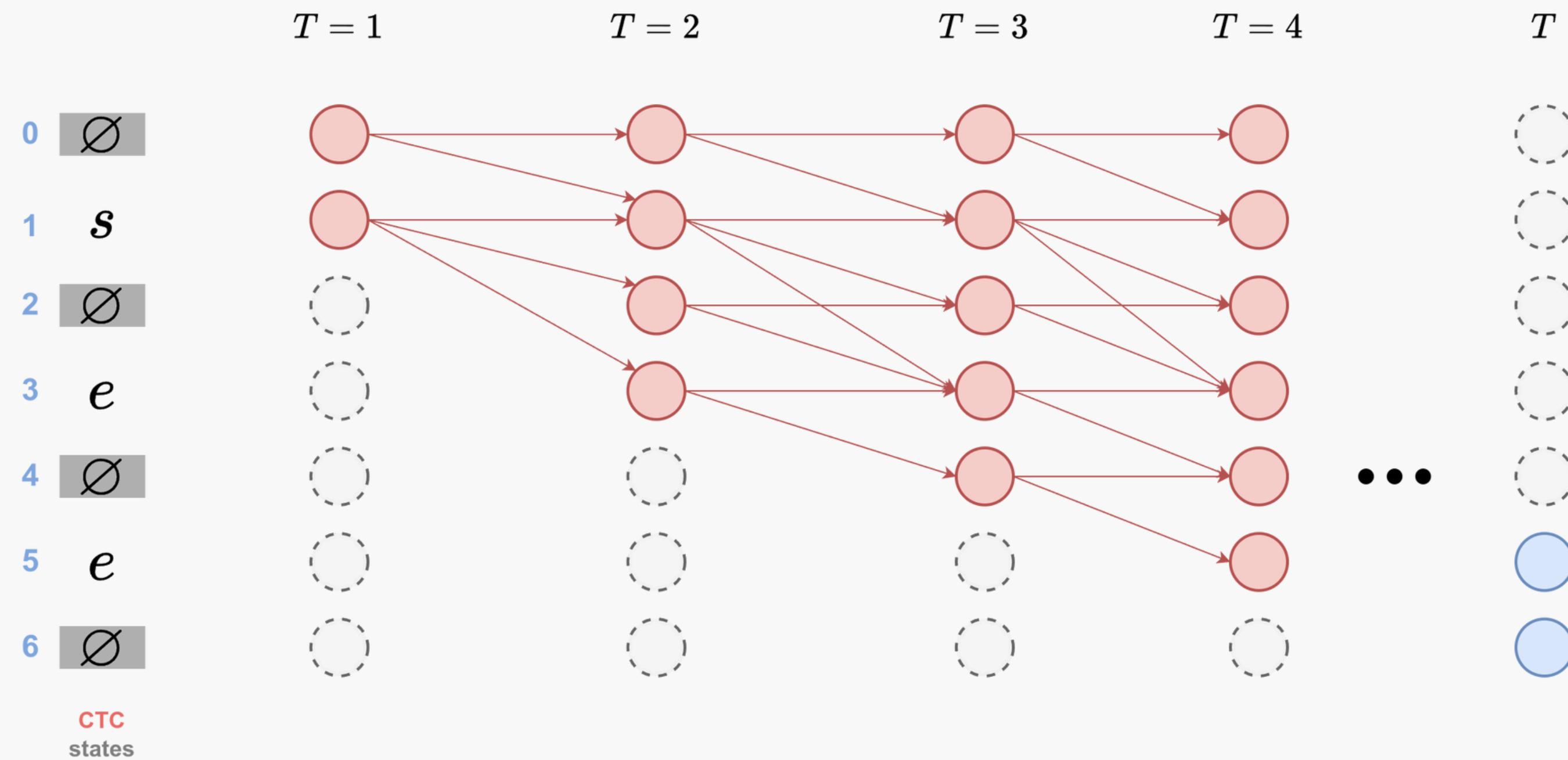
Dynamic Programming



Dynamic Programming



Dynamic Programming





CTC formulation

$$\alpha_t(j) = \sum_{i=j-2}^j \alpha_{t-1}(i) a_{ij} b_t(y'_j)$$

$$t = 1 \dots T$$

$$j = 1 \dots 2l + 1$$

$$\begin{aligned} |\mathbf{O}| &= T \\ |y| &= l \end{aligned}$$

CTC formulation

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$$y'_j = \begin{cases} y_{j/2} & \text{if } j \text{ is even} \\ \emptyset & \text{otherwise} \end{cases}$$

CTC formulation

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$$a_{ij} = \begin{cases} 1 & \text{if } i = j \text{ or } i = j - 1 \\ 1 & \text{if } i = j - 2 \text{ and } y'_j \neq y'_{j-2} \\ 0 & \text{otherwise} \end{cases}$$



CTC formulation

$$Pr(\mathbf{y} \mid \mathbf{O}) = \sum_{a \in \mathcal{B}^{-1}(\mathbf{y})} Pr(a \mid \mathbf{O}) = \underbrace{\alpha_T(2l)}_{\text{alignment ends with last token}} + \underbrace{\alpha_T(2l+1)}_{\text{alignment ends with blank}}$$

alignment ends with last token alignment ends with blank



Prediction with CTC

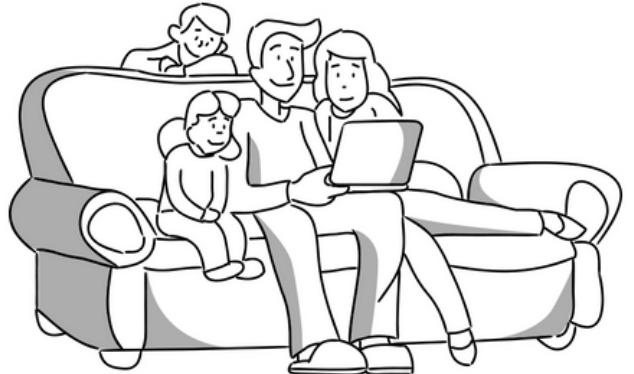
Pick the **single most probable output** at every time step

$$\arg \max_{\mathbf{y}} Pr(\mathbf{y} \mid \mathbf{O}) \approx \mathcal{B}(\arg \max_a Pr(a \mid \mathbf{O}))$$



Thank you!

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Want to know more about ASR? Check out:

CS753: Automatic Speech Recognition

