

RACV 2021



Transformer促进范式迁移

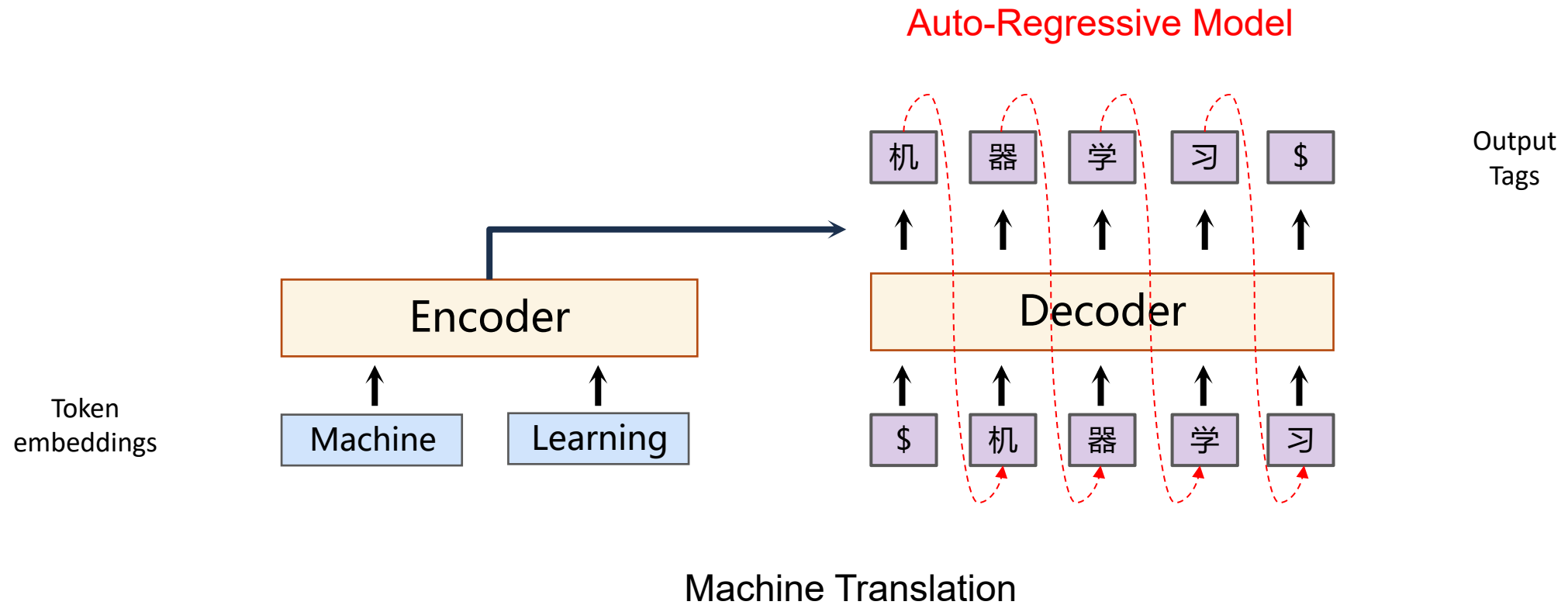
邱锡鹏

复旦大学

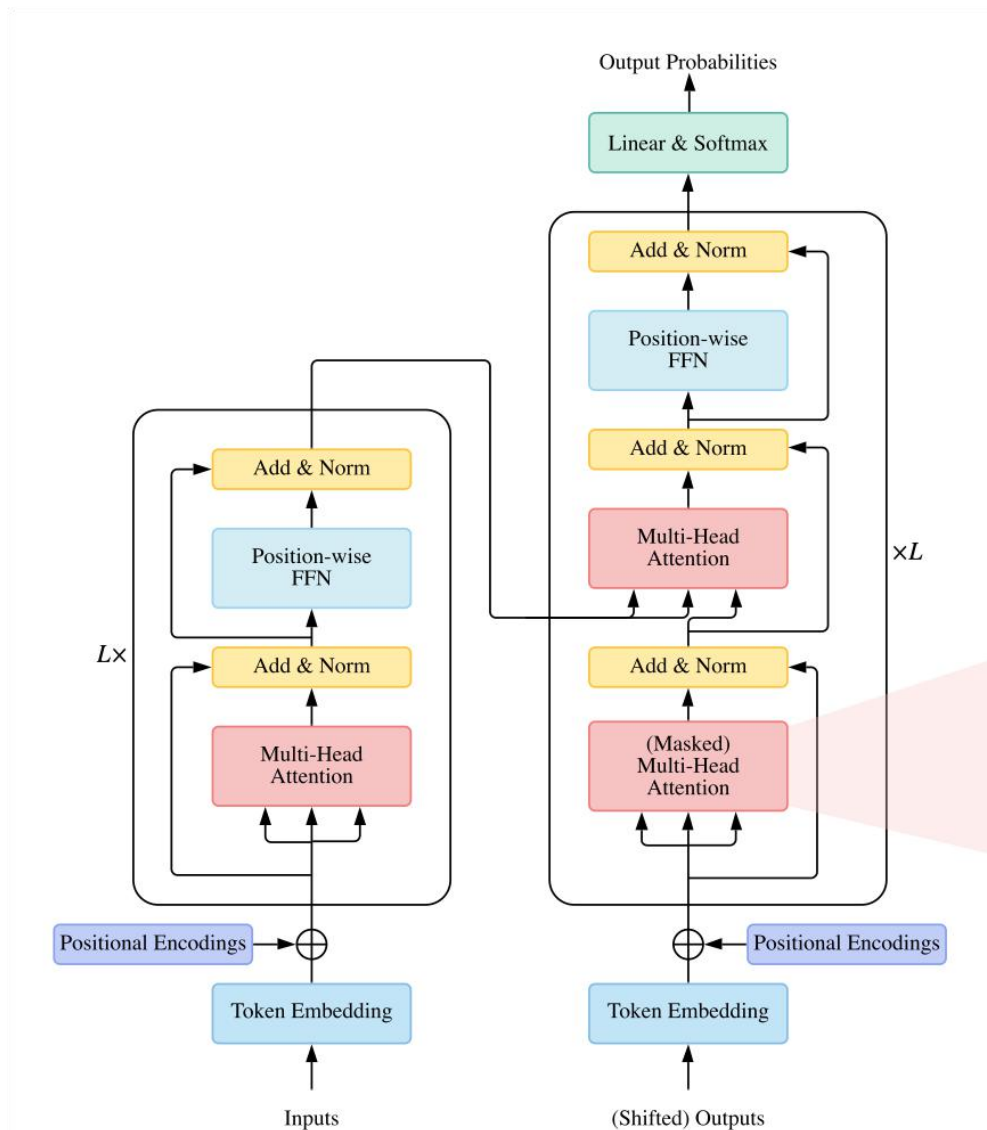
<https://xpqiu.github.io/>

16/10/2021

Sequence-to-Sequence (Seq2Seq) In NLP

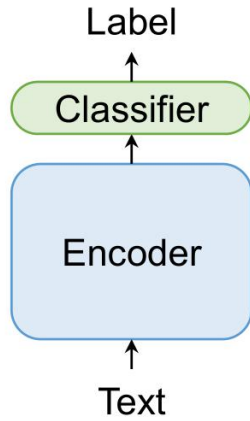


Transformer

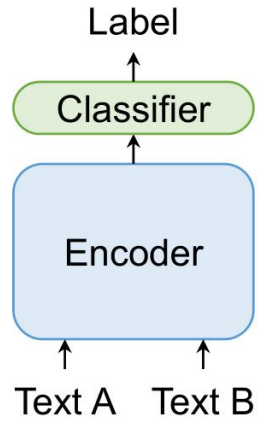


Decoder part is neglected in CV.

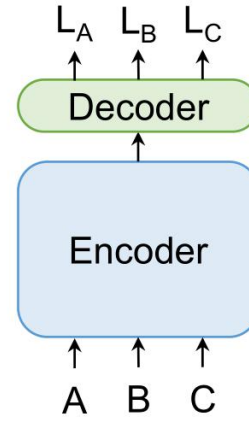
Seven Main Paradigms In NLP



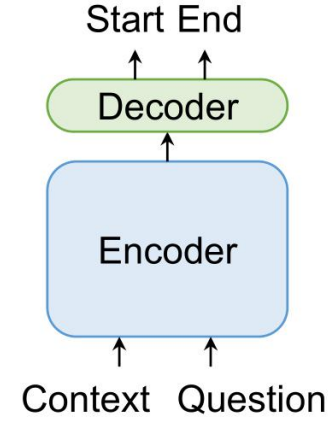
(a) Class



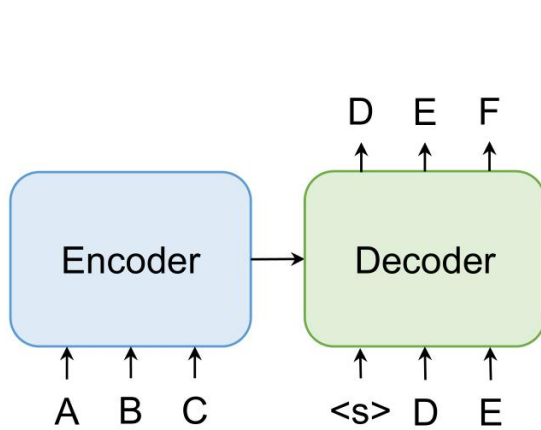
(b) Matching



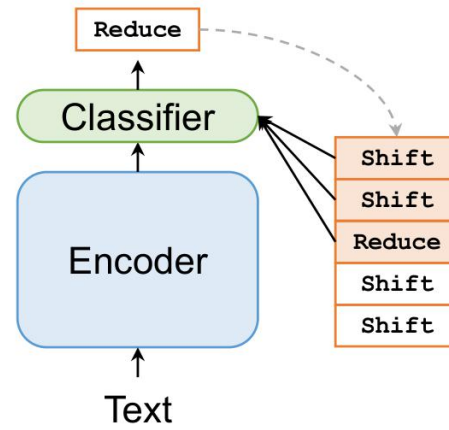
(c) SeqLab



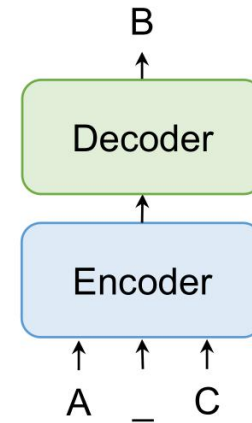
(d) MRC



(e) Seq2Seq

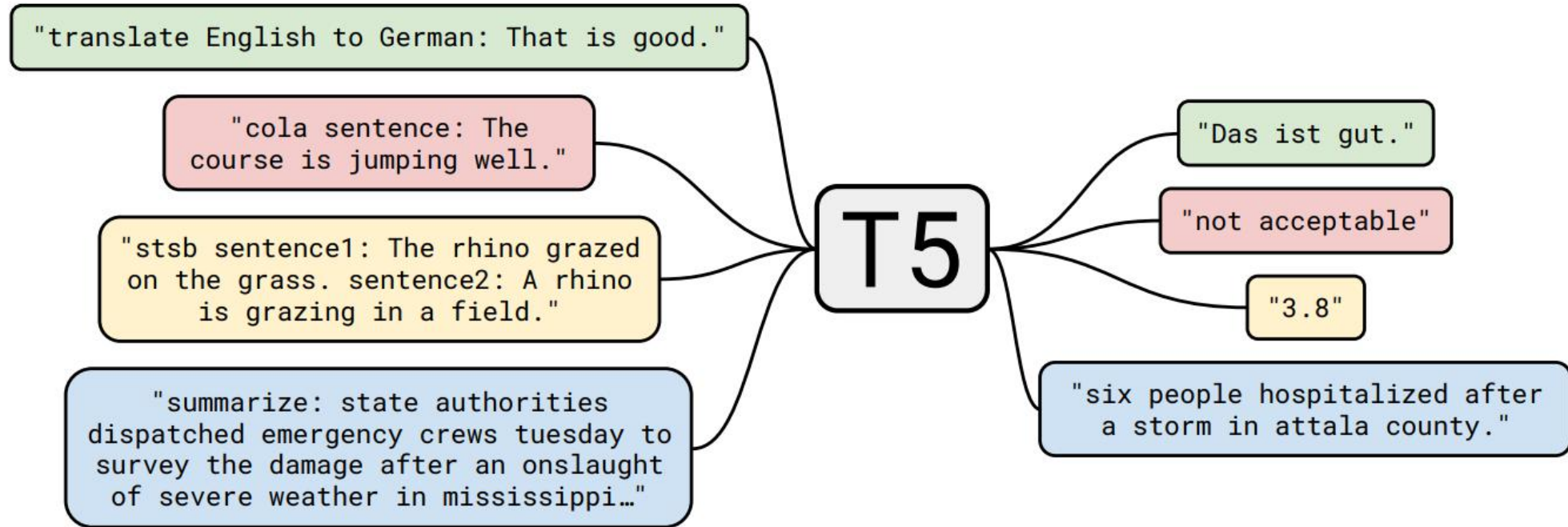


(f) Seq2ASeq

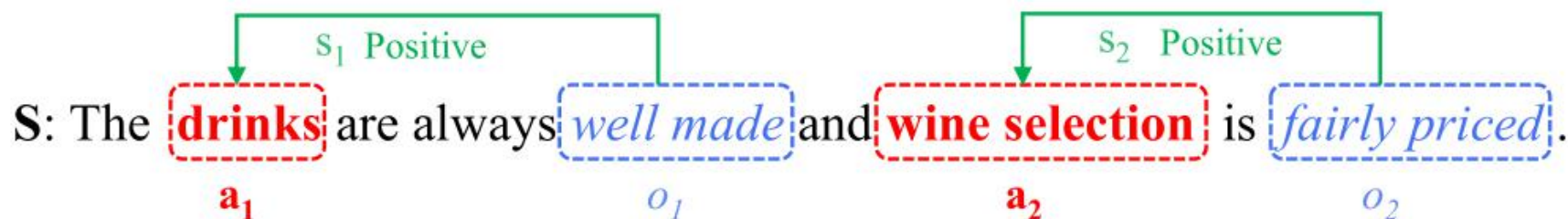


(g) (M)LM

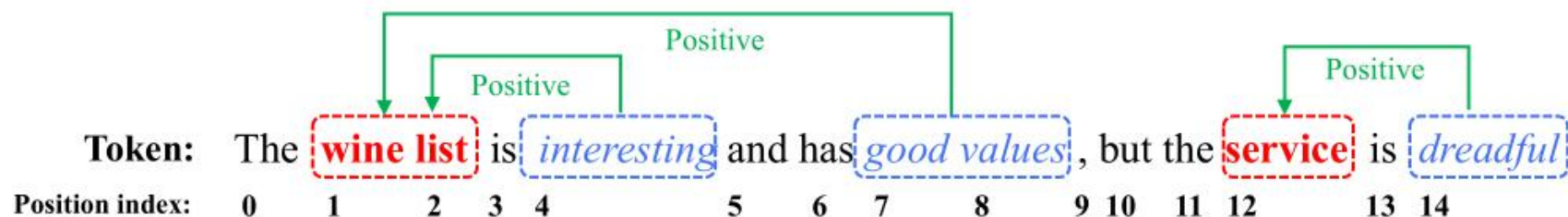
Text-to-text framework



T5: Text-to-Text Transfer Transformer



| Subtask | Input | Output | Task Type |
|--|------------------------|------------------------------------|-----------------------------|
| Aspect Term Extraction(AE) | S | a_1, a_2 | Extraction |
| Opinion Term Extraction(OE) | S | o_1, o_2 | Extraction |
| Aspect-level Sentiment Classification(ALSC) | S + a_1 S + a_2 | s_1 s_2 | Classification |
| Aspect-oriented Opinion Extraction(AOE) | S + a_1 S + a_2 | o_1 o_2 | Extraction |
| Aspect Term Extraction and Sentiment Classification(AESC) | S | $(a_1, s_1), (a_2, s_2)$ | Extraction & Classification |
| Pair Extraction(Pair.) | S | $(a_1, o_1), (a_2, o_2)$ | Extraction |
| Triplet Extraction(Triplet.) | S | $(a_1, o_1, s_1), (a_2, o_2, s_2)$ | Extraction & Classification |



| Subtask | Target Sequence |
|-----------------|---|
| <i>AE</i> | 1, 2, 12, 12, </s> |
| <i>OE</i> | 4, 4, 7, 8, 14, 14, </s> |
| <i>ALSC</i> | <u>1</u> , <u>2</u> , POS, </s> |
| | <u>12</u> , <u>12</u> , POS, </s> |
| <i>AOE</i> | <u>1</u> , <u>2</u> , 4, 4, 7, 8, </s> |
| | <u>12</u> , <u>12</u> , 14, 14, </s> |
| <i>AESC</i> | 1, 2, POS, 12, 12, NEG, </s> |
| <i>Pair.</i> | 1, 2, 4, 4, 1, 2, 7, 8, 12, 12, 14, 14, </s> |
| <i>Triplet.</i> | 1, 2, 4, 4, POS, 1, 2, 7, 8, POS, 12, 12, 14, 14, POS, </s> |

NER



| |
|---|
| <p>S1: <u>B-Per Barack</u> <u>I-Per Obama</u> O was O born O in O the <u>B-Loc US</u> Person Location</p> <p>(a) Sequence labelling for flat NER</p> |
| <p><p>S2: The <u>Lincoln</u> Memorial Person Location</p><p>(b) Span-based classification for nested NER</p></p> |
| <p>Actions: OUT OUT SHIFT SHIFT LEFT-REDUCE COMPLETE ... S3: have much <u>muscle pain</u> and <u>fatigue</u> Disorder Disorder</p> <p>(c) Transition-based method for discontinuous NER</p> |
| <p>S1: Barack Obama <Person> US <Location> S2: The Lincoln Memorial <Location> Lincoln <Person> S2: muscle pain <Disorder> muscle fatigue <Disorder></p> <p>(d) A unified generative solution for all NER tasks</p> |

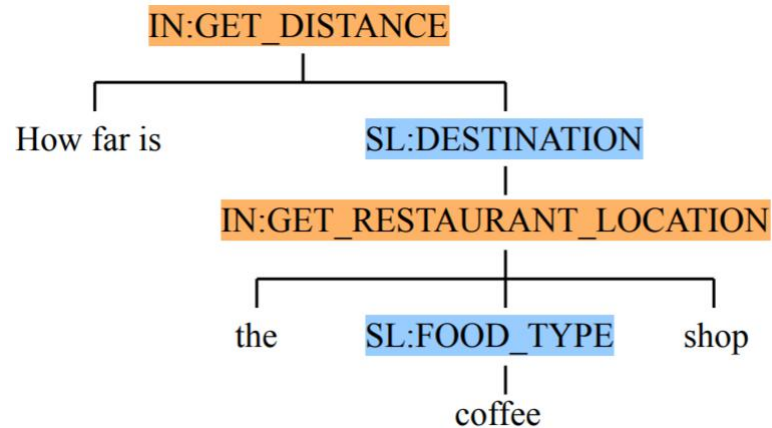
Semantic parsing as Seq2Seq



“simple” query



Source: play the song don't stop believin by journey
 Target: PlaySongIntent SongName(@ptr₃ @ptr₄ @ptr₅)
 SongName ArtistName(@ptr₇)ArtistName

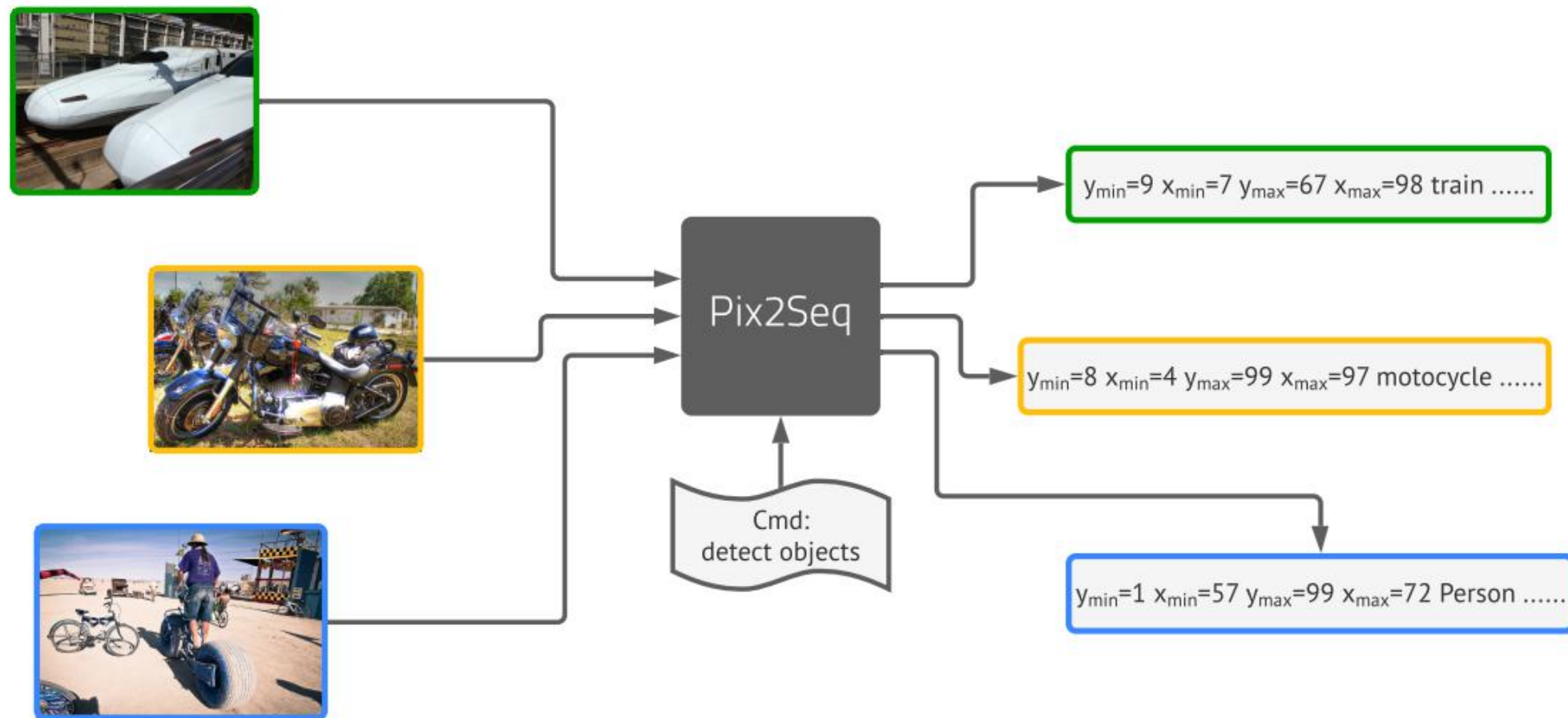


“complex” query



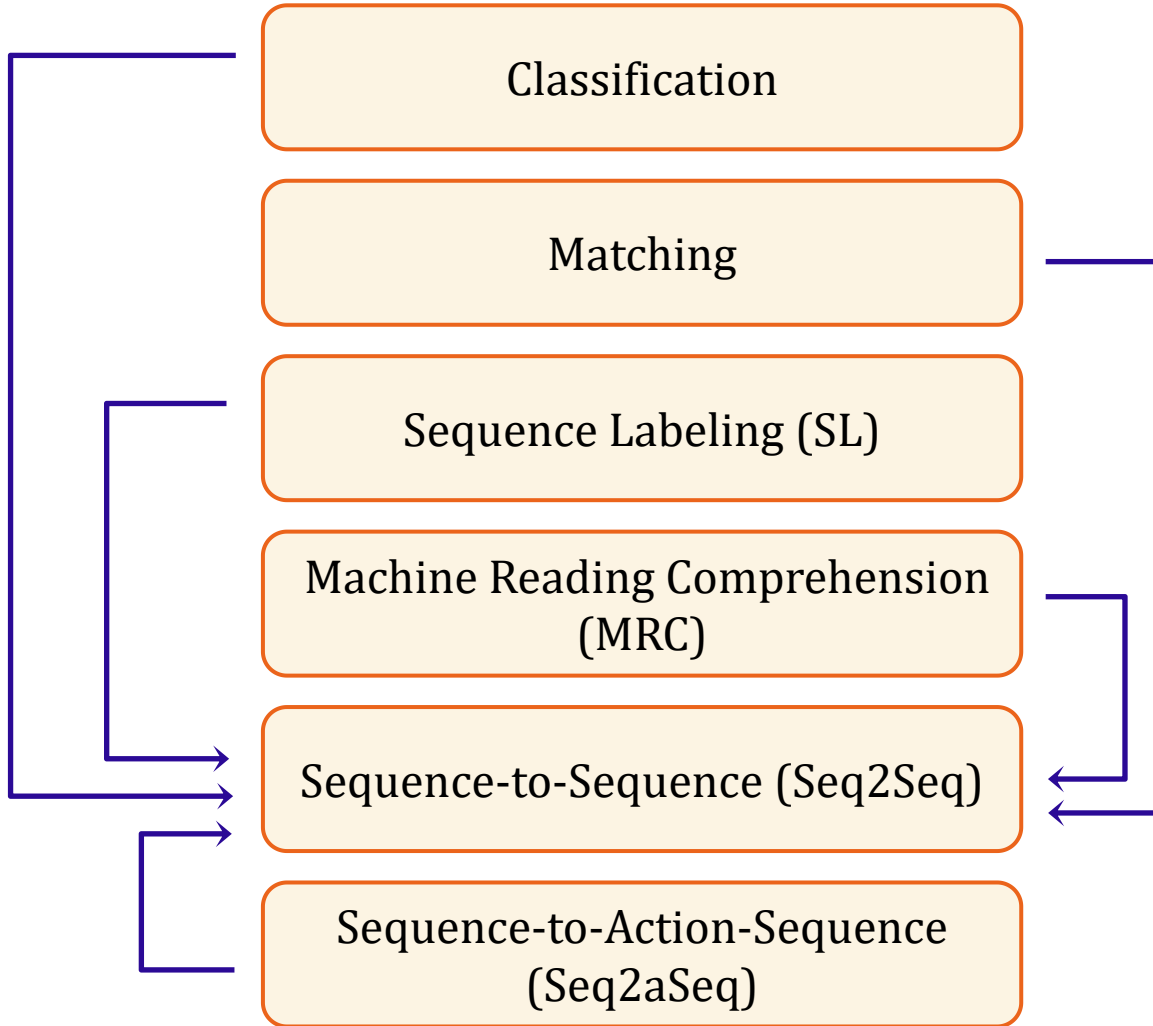
Source: How far is the coffee shop
 Target: [IN:GET_DISTANCE @ptr₀ @ptr₁ @ptr₂ [SL:DESTINATION
 [IN:GET_RESTAURANT_LOCATION @ptr₃ [SL:TYPE_FOOD
 @ptr₄ SL:TYPE_FOOD] @ptr₅ IN:GET_RESTAURANT_LOCATION]
 SL:DESTINATION] IN:GET_DISTANCE]

A recent work in CV



Pix2Seq framework for object detection

Summary



Paradigm Shift brings various NLP tasks a unified framework.
The pretrained backbone models accelerate this shift.

Seq2Seq are a powerful paradigm, which is more general and flexible.