

# AI Task-Oriented Dialogue System for Conversational Commerce in FinTech

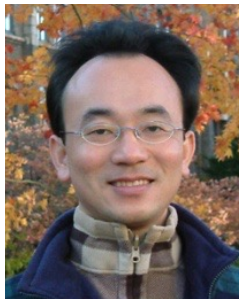
Host: Prof. Kuan-Yu Memphis Chen

Computer Science and Information Engineering, National Taiwan University of Science and Technology

Time: 14:00-15:00, May 10, 2021 (Monday)

Place: CSIE, NTUST

Address: No.43, Keelung Rd., Sec.4, Da'an Dist., Taipei, Taiwan



**Min-Yuh Day**

戴敏育

**Associate Professor**

副教授

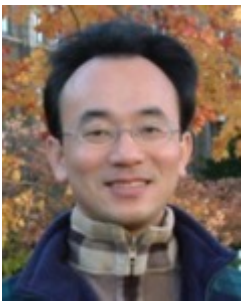
Institute of Information Management, National Taipei University

國立臺北大學 資訊管理研究所

<https://web.ntpu.edu.tw/~myday>

2021-05-10





# 戴敏育 博士

## (Min-Yuh Day, Ph.D.)

國立台北大學 資訊管理研究所 副教授  
中央研究院 資訊科學研究所 訪問學人  
國立台灣大學 資訊管理 博士

Publications Co-Chairs, IEEE/ACM International Conference on  
Advances in Social Networks Analysis and Mining (ASONAM 2013- )

Program Co-Chair, IEEE International Workshop on  
Empirical Methods for Recognizing Inference in Text (IEEE EM-RITE 2012- )

Publications Chair, The IEEE International Conference on  
Information Reuse and Integration (IEEE IRI)



# Outline

- Artificial Intelligence
- FinTech
- Conversational Commerce
- Task Oriented Dialogue System

# AIWISFIN

## AI Conversational Robo-Advisor (人工智慧對話式理財機器人)

First Place, InnoServe Awards 2018

InnoServe 資服創新  
競賽粉絲團  
@InnoServe.tca.org

Home  
About  
Photos  
Welcome  
發燒粉絲活動  
Welcome  
Videos  
Posts  
Community  
Info and Ads  
Create a Page

Liked Following Share ...

InnoServe 資服創新競賽粉絲團 shared a post.  
November 28 at 2:43 PM ·

《#InnoServe 競賽得獎作品系列報導七》  
理財 方式百百種卻不知道該從何著手嗎？  
來看金融結合 AI 如何讓投資變得更簡單。

28,112 Views

經濟部工業局  
November 28 at 11:37 AM ·

假如有一筆錢，您知道要怎麼投資嗎？

本作品「AIWISFIN」使用 #深度學習 預測股價漲跌、  
配置投資組合，分析 客戶需求，  
提供 #客製化 投資建議 與 #智慧對話，  
讓年輕投資者使用更方便！

- 得獎作品：AIWISFIN 人工智慧對話式理財機器人
- 獎項：玉山銀行金融科技趨勢應用組第1名
- 得獎學校：淡江大學 (資訊管理學系)
- 指導老師：戴敏育老師
- 得獎團隊：陳元致、鄧旭廷、王慶宇、邱少文
- 影片連結：<https://ppt.cc/fyc3sx>

<https://www.youtube.com/watch?v=sEhmyoTXmGk>

# 2018 The 23<sup>th</sup> International ICT Innovative Services Awards (InnoServe Awards 2018)



- Annual ICT application competition held for university and college students
- The largest and the most significant contest in Taiwan.
- More than **ten thousand teachers and students** from over **one hundred universities and colleges** have participated in the Contest.

# 2018 International ICT Innovative Services Awards (InnoServe Awards 2018)

## (2018第23屆大專校院資訊應用服務創新競賽)



第23屆 大專校院  
2018 資訊應用服務創新競賽  
International ICT Innovative Services Awards 2018

創意噴發!

InnoServe Awards

總獎金 > 200 萬

■ 報名日期: 2018/10/2(二)~  
2018/10/9(二)pm6點截止

■ 參賽對象: 大專校院學生、  
碩博士生及高中職學生

■ 決賽時間: 2018/11/3(六)  
■ 決賽地點: 國立臺灣大學  
綜合體育館

- 最新消息 ▾
- 活動訊息
- 媒體轉載
- 競賽緣起
- 競賽辦法 ▾
- 競賽報名
- 活動成果 ▾
- 產學媒合 ▾
- 媒合
- 聯絡我們

### 榮譽榜

屆別 23 ▾ 查詢

第23屆

顯示 30 ▾ 筆資料

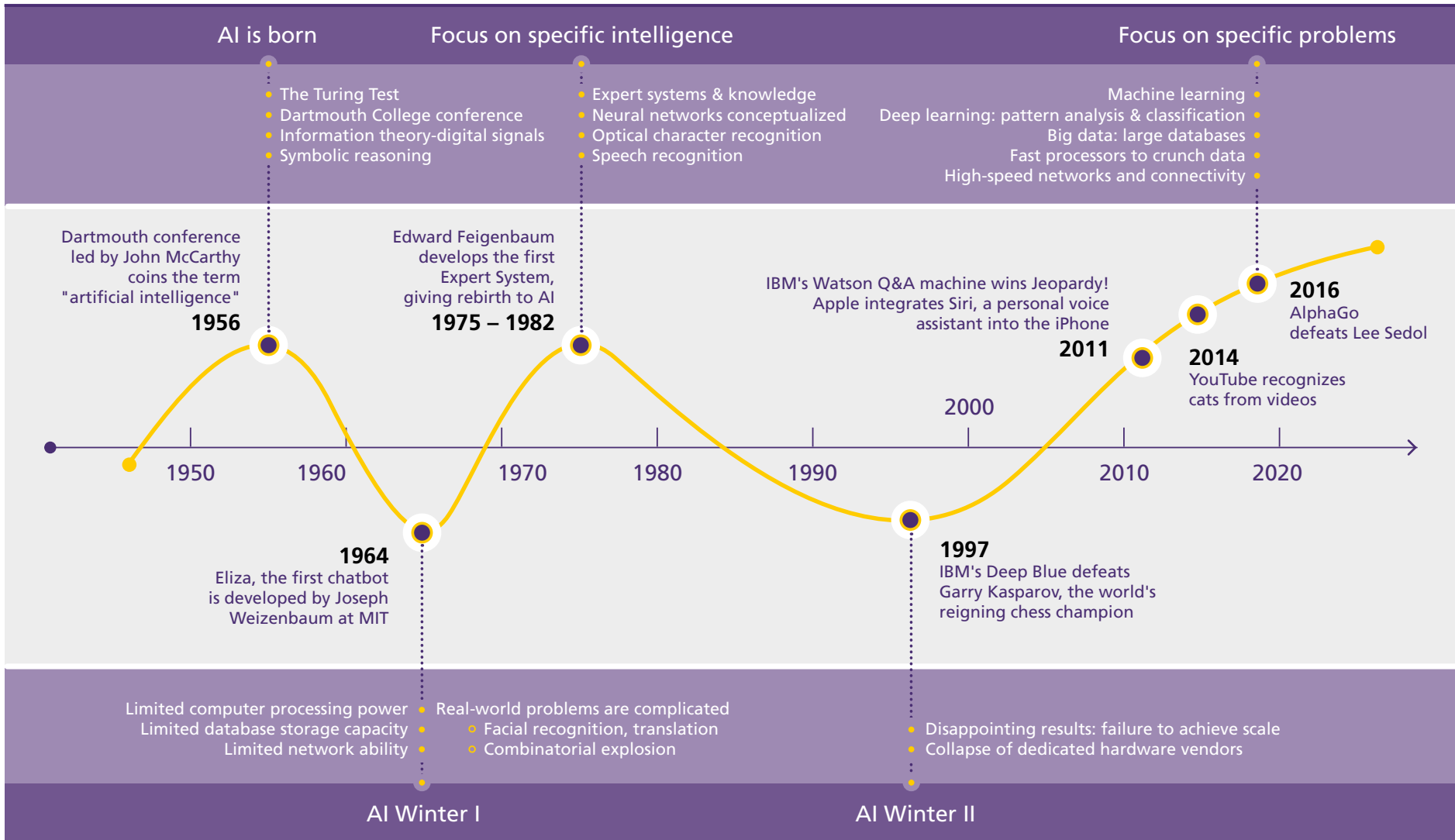
表格內全文檢索: AIWISFIN

組別	名次	組別編號	學校名稱	專題名稱	指導教授	學生
資訊應用組一	第一名	IP1-06	淡江大學	AIWISFIN 人工智慧對話式理財機器人	戴敏育老師	陳元致、鄧旭廷、王慶宇、邱少文
玉山銀行金融科技趨勢應用組	第一名	E.SUN FINTECH-01	淡江大學	AIWISFIN 人工智慧對話式理財機器人	戴敏育老師	陳元致、鄧旭廷、王慶宇、邱少文

<https://innoserve.tca.org.tw/award.aspx>

**AI**

# The Rise of AI





# **Definition of Artificial Intelligence (A.I.)**

# Artificial Intelligence

**“... the science and  
engineering  
of  
making  
intelligent machines”  
(John McCarthy, 1955)**

# Artificial Intelligence

**“... technology that  
thinks and acts  
like humans”**

# Artificial Intelligence

**“... intelligence  
exhibited by machines  
or software”**

# 4 Approaches of AI

<b>Thinking Humanly</b>	<b>Thinking Rationally</b>
<b>Acting Humanly</b>	<b>Acting Rationally</b>

# 4 Approaches of AI

**2.**

**Thinking Humanly:  
The Cognitive  
Modeling Approach**

**3.**

**Thinking Rationally:  
The “Laws of Thought”  
Approach**

**1.**

**Acting Humanly:  
The Turing Test  
Approach** (1950)

**4.**

**Acting Rationally:  
The Rational Agent  
Approach**

# AI Acting Humanly: The Turing Test Approach (Alan Turing, 1950)

- Knowledge Representation
- Automated Reasoning
- Machine Learning (ML)
  - Deep Learning (DL)
- Computer Vision (Image, Video)
- Natural Language Processing (NLP)
- Robotics

# IMTKU Textual Entailment System for Recognizing Inference in Text at **NTCIR-9** RITE

Department of Information Management  
Tamkang University, Taiwan



Min-Yuh Day



Chun Tu

[myday@mail.tku.edu.tw](mailto:myday@mail.tku.edu.tw)



# IMTKU Textual Entailment System for Recognizing Inference in Text at **NTCIR-10** RITE-2

Department of Information Management  
Tamkang University, Taiwan



Min-Yuh Day



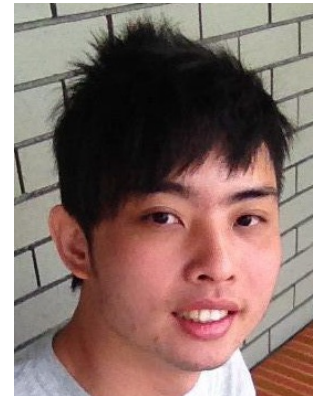
Chun Tu



Hou-Cheng Vong



Shih-Wei Wu



Shih-Jhen Huang

[myday@mail.tku.edu.tw](mailto:myday@mail.tku.edu.tw)

# IMTKU Textual Entailment System for Recognizing Inference in Text at **NTCIR-11** RITE-VAL

**Tamkang University**

淡江大學

# 2014



**Min-Yuh Day**



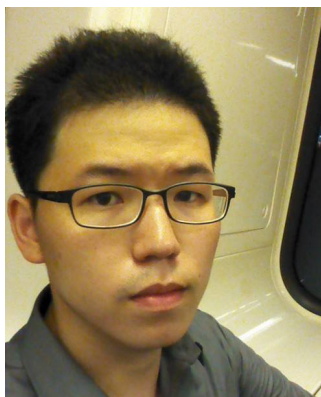
**Ya-Jung Wang**



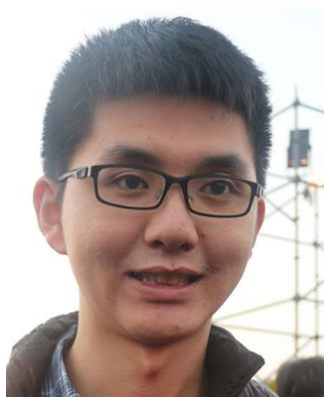
**Che-Wei Hsu**



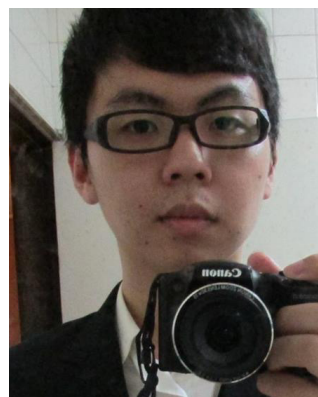
**En-Chun Tu**



**Huai-Wen Hsu**



**Yu-An Lin**



**Shang-Yu Wu**



**Yu-Hsuan Tai**



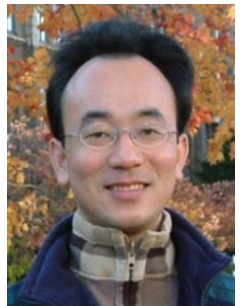
**Cheng-Chia Tsai**



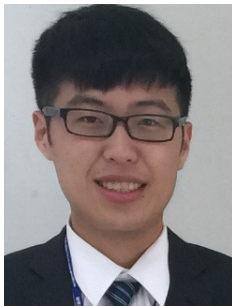
# IMTKU Question Answering System for World History Exams at **NTCIR-12** QA Lab2

Department of Information Management  
Tamkang University, Taiwan

Sagacity Technology



Min-Yuh Day



Cheng-Chia Tsai



Wei-Chun Chung



Hsiu-Yuan Chang



Tzu-Jui Sun



Yuan-Jie Tsai



Jin-Kun Lin



Cheng-Hung Lee



Yu-Ming Guo



Yue-Da Lin



Wei-Ming Chen



Yun-Da Tsai



Cheng-Jhih Han



Yi-Jing Lin



Yi-Heng Chiang



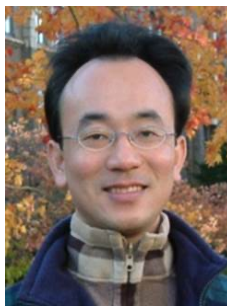
Ching-Yuan Chien

[myday@mail.tku.edu.tw](mailto:myday@mail.tku.edu.tw)

NTCIR-12 Conference, June 7-10, 2016, Tokyo, Japan

# IMTKU Question Answering System for World History Exams at **NTCIR-13** QALab-3

Department of Information Management  
Tamkang University, Taiwan



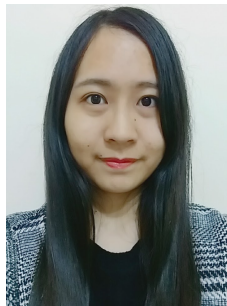
Min-Yuh Day



Chao-Yu Chen



Wanchu Huang



Shi-Ya Zheng



I-Hsuan Huang



Tz-Rung Chen



Min-Chun Kuo



Yue-Da Lin



Yi-Jing Lin

[myday@mail.tku.edu.tw](mailto:myday@mail.tku.edu.tw)



# IMTKU Emotional Dialogue System for Short Text Conversation at **NTCIR-14** STC-3 (CECG) Task

Department of Information Management  
Tamkang University, Taiwan



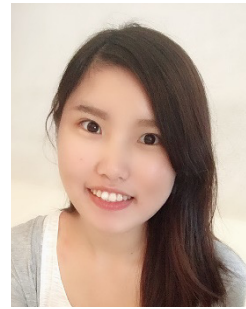
Min-Yuh Day



Chi-Sheng Hung



Yi-Jun Xie



Jhih-Yi Chen



Yu-Ling Kuo



Jian-Ting Lin

[myday@mail.tku.edu.tw](mailto:myday@mail.tku.edu.tw)

# IMTKU Multi-Turn Dialogue System Evaluation at the NTCIR-15 DialEval-1 Dialogue Quality and Nugget Detection

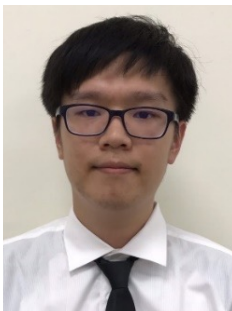
<sup>1</sup> Zeals Co., Ltd. Tokyo, Japan

<sup>2</sup> Information Management, Tamkang University, Taiwan

<sup>3</sup> Information Management, National Taipei University, Taiwan



Mike Tian-Jian Jiang<sup>1</sup>



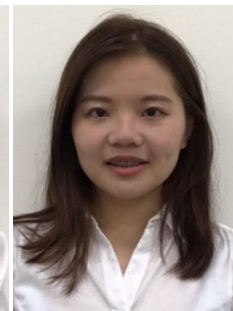
Zhao-Xian Gu<sup>2</sup>



Cheng-Jhe Chiang<sup>2</sup>



Yueh-Chia Wu<sup>2</sup>



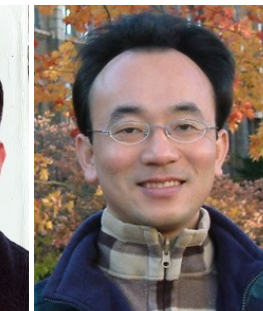
Yu-Chen Huang<sup>2</sup>



Cheng-Han Chiu<sup>2</sup>



Sheng-Ru Shaw<sup>2</sup>



Min-Yuh Day<sup>3</sup>

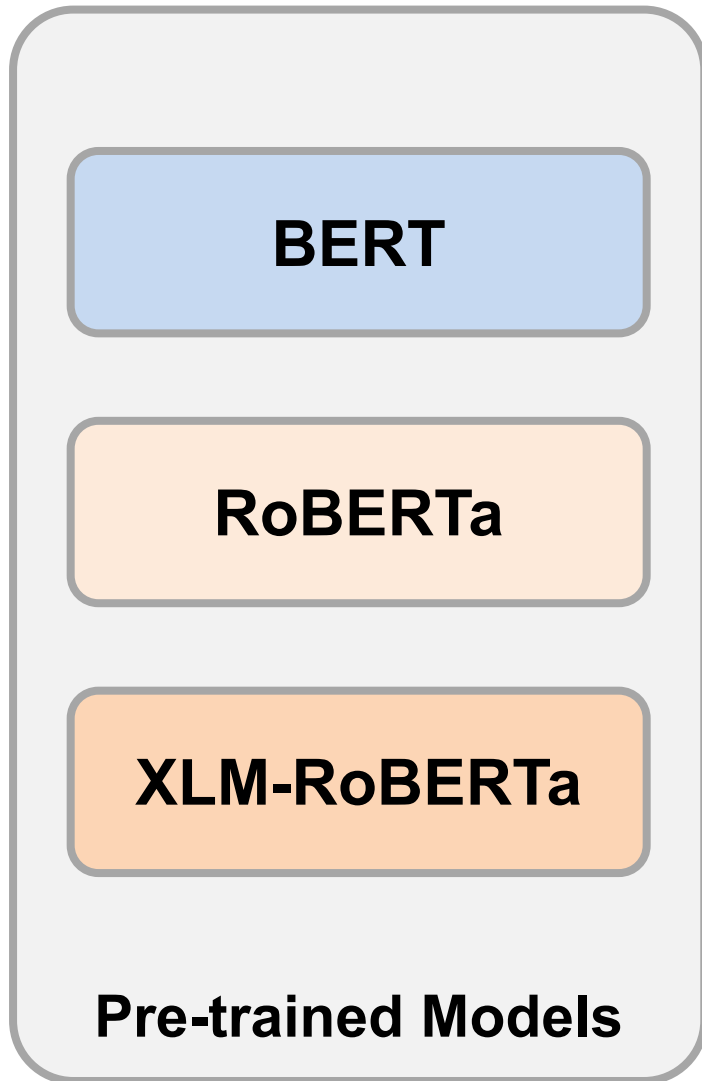
# 2020 NTCIR-15 Dialogue Evaluation (DialEval-1) Task

## Dialogue Quality (DQ) and Nugget Detection (ND)

### Chinese Dialogue Quality (S-score) Results (Zeng et al., 2020)

Run	Mean RSNOD	Run	Mean NMD
IMTKU-run2	0.1918	IMTKU-run2	0.1254
IMTKU-run1	0.1964	IMTKU-run0	0.1284
IMTKU-run0	0.1977	IMTKU-run1	0.1290
TUA1-run2	0.2024	TUA1-run2	0.1310
TUA1-run0	0.2053	TUA1-run0	0.1322
NKUST-run1	0.2057	NKUST-run1	0.1363
BL-lstm	0.2088	TUA1-run1	0.1397
WUST-run0	0.2131	BL-popularity	0.1442
RSLNV-run0	0.2141	BL-lstm	0.1455
BL-popularity	0.2288	RSLNV-run0	0.1483
TUA1-run1	0.2302	WUST-run0	0.1540
NKUST-run0	0.2653	NKUST-run0	0.2289
BL-uniform	0.2811	BL-uniform	0.2497

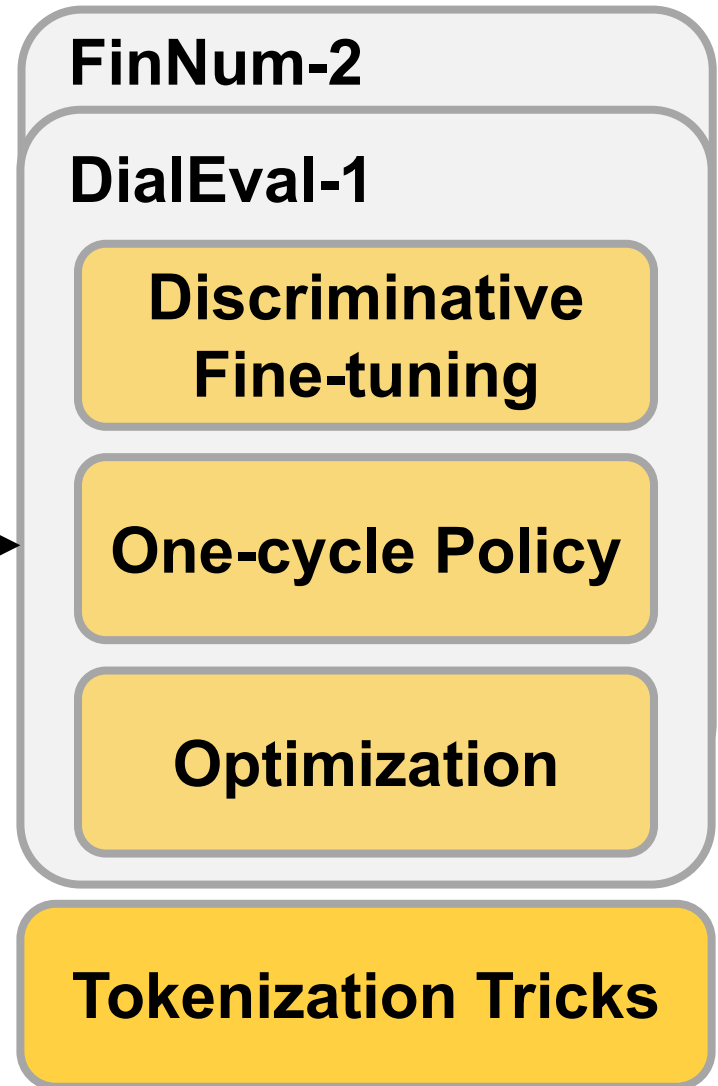
## Transformer-based Models Selection



Transfer Learning



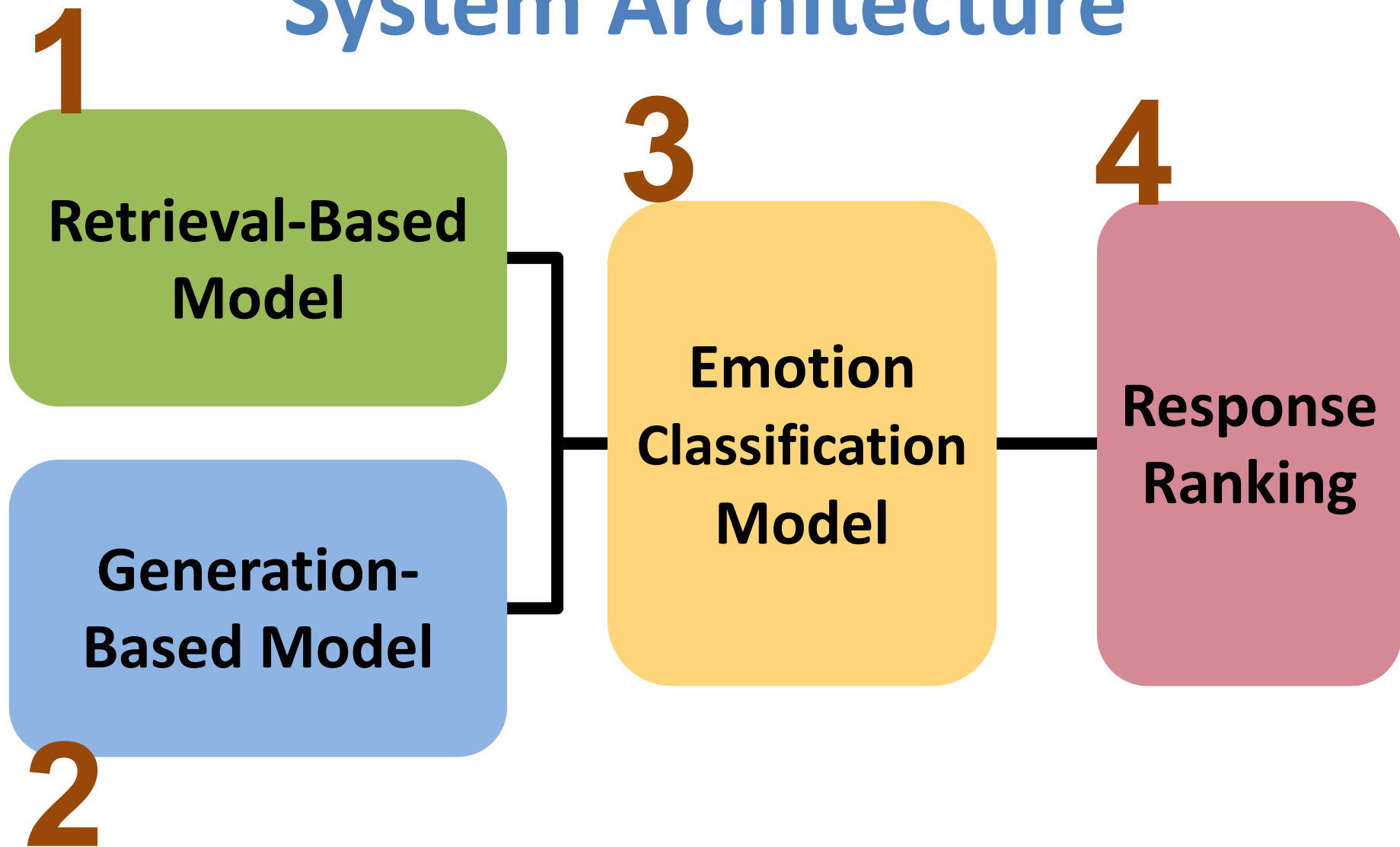
## Fine-tuning Techniques







# IMTKU Emotional Dialogue System Architecture





**Short Text Conversation Task  
(STC-3)  
Chinese Emotional Conversation  
Generation (CECG) Subtask**

# NTCIR Short Text Conversation

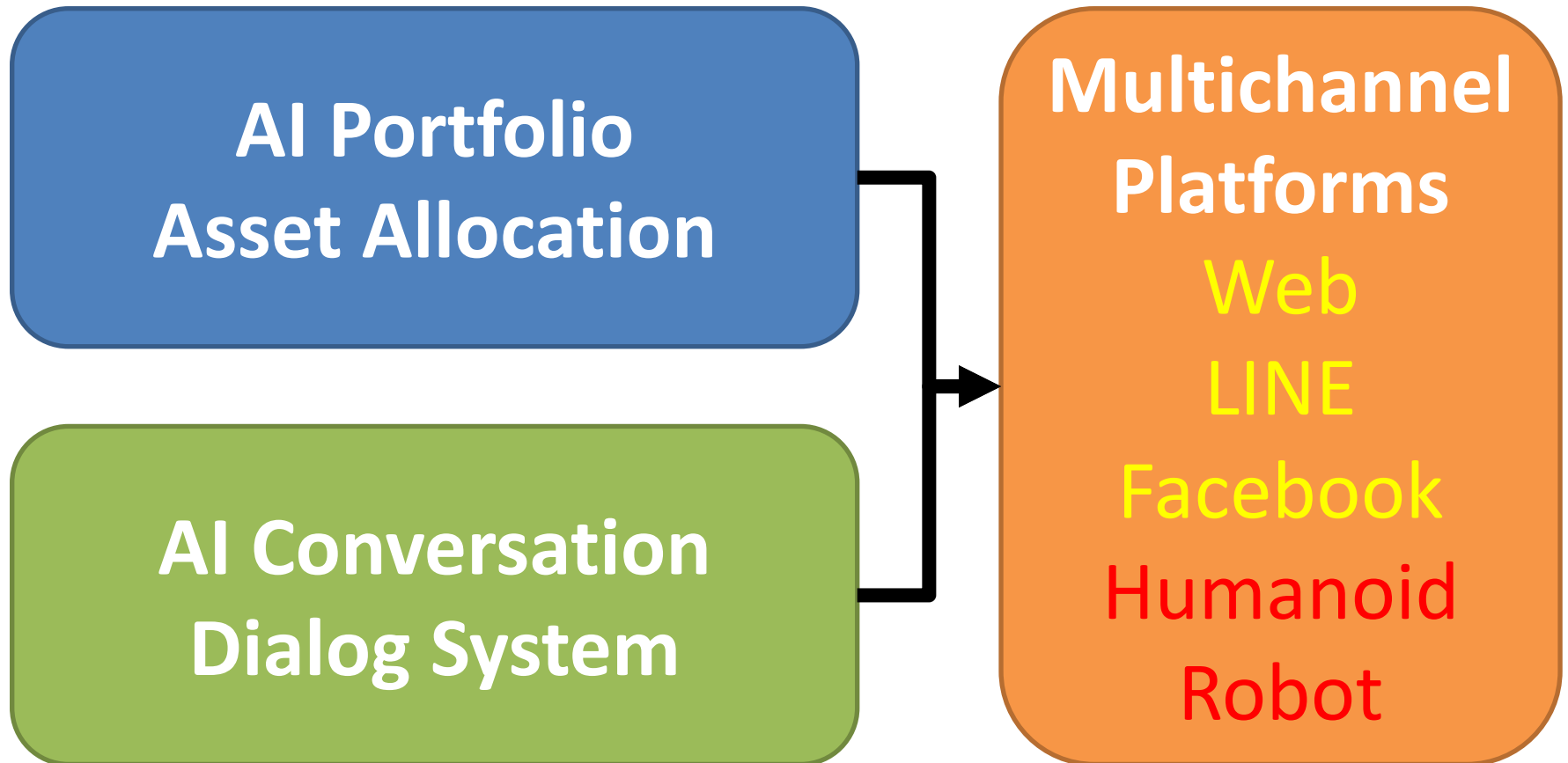
## STC-1, STC-2, STC-3

	Japanese	Chinese	English	
NTCIR-12 STC-1 22 active participants	Twitter, Retrieval	Weibo, Retrieval		Single-turn, Non task-oriented
NTCIR-13 STC-2 27 active participants	Yahoo! News, Retrieval+ Generation	Weibo, Retrieval+ Generation		
NTCIR-14 STC-3		Weibo, Generation for given emotion categories		Multi-turn, task-oriented (helpdesk)
Chinese Emotional Conversation Generation (CECG) subtask		Weibo+English translations, distribution estimation for subjective annotations		
Dialogue Quality (DQ) and Nugget Detection (ND) subtasks				

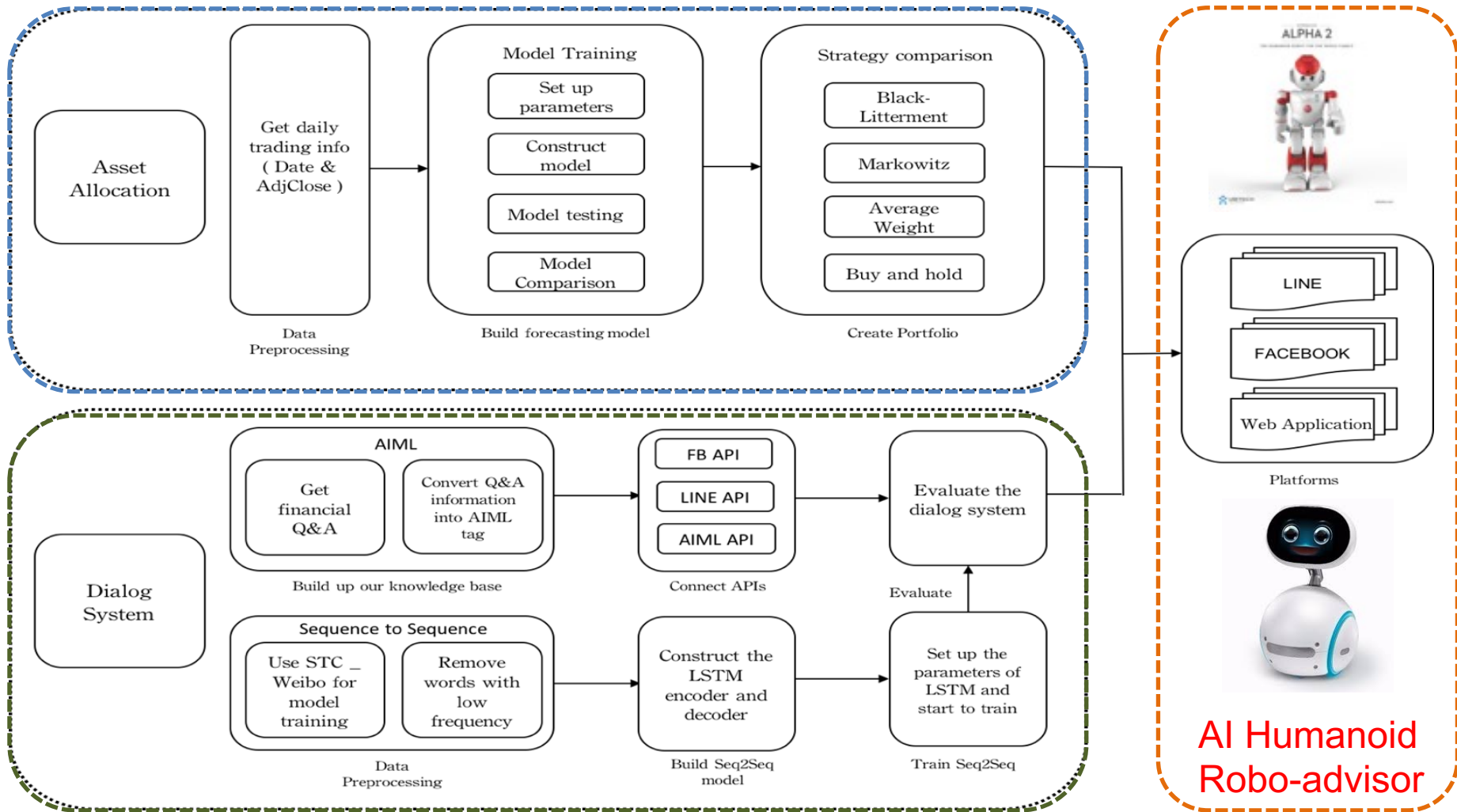
Source: <https://waseda.app.box.com/v/STC3atNTCIR-14>

# AI Humanoid Robo-Advisor

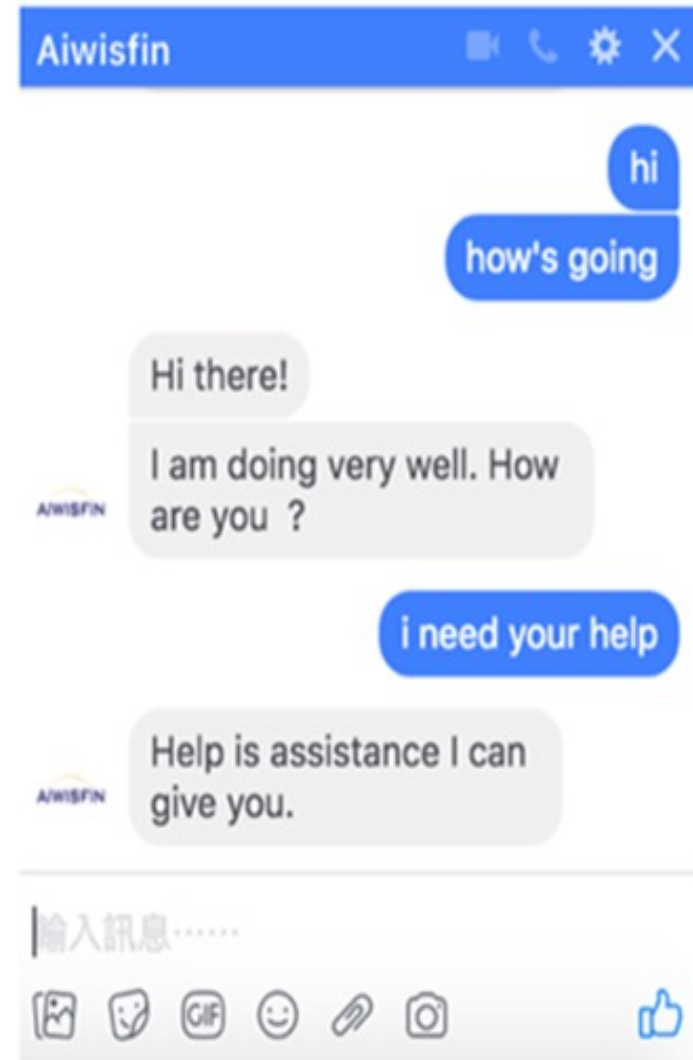
# AI Humanoid Robo-Advisor for Multi-channel Conversational Commerce



# System Architecture of AI Humanoid Robo-Advisor



# Conversational Model (LINE, FB Messenger)



# Conversational Robo-Advisor Multichannel UI/UX Robots



## ALPHA 2

## ZENBO

INTRODUCING  
**ALPHA 2**  
THE HUMANOID ROBOT FOR THE WHOLE FAMILY



ubtech.com





**FinTech**

# Financial Technology

## FinTech

“providing  
financial services  
by making use of  
software and  
modern technology”

# Financial Services

# Financial Services



# FinTech: Financial Services Innovation

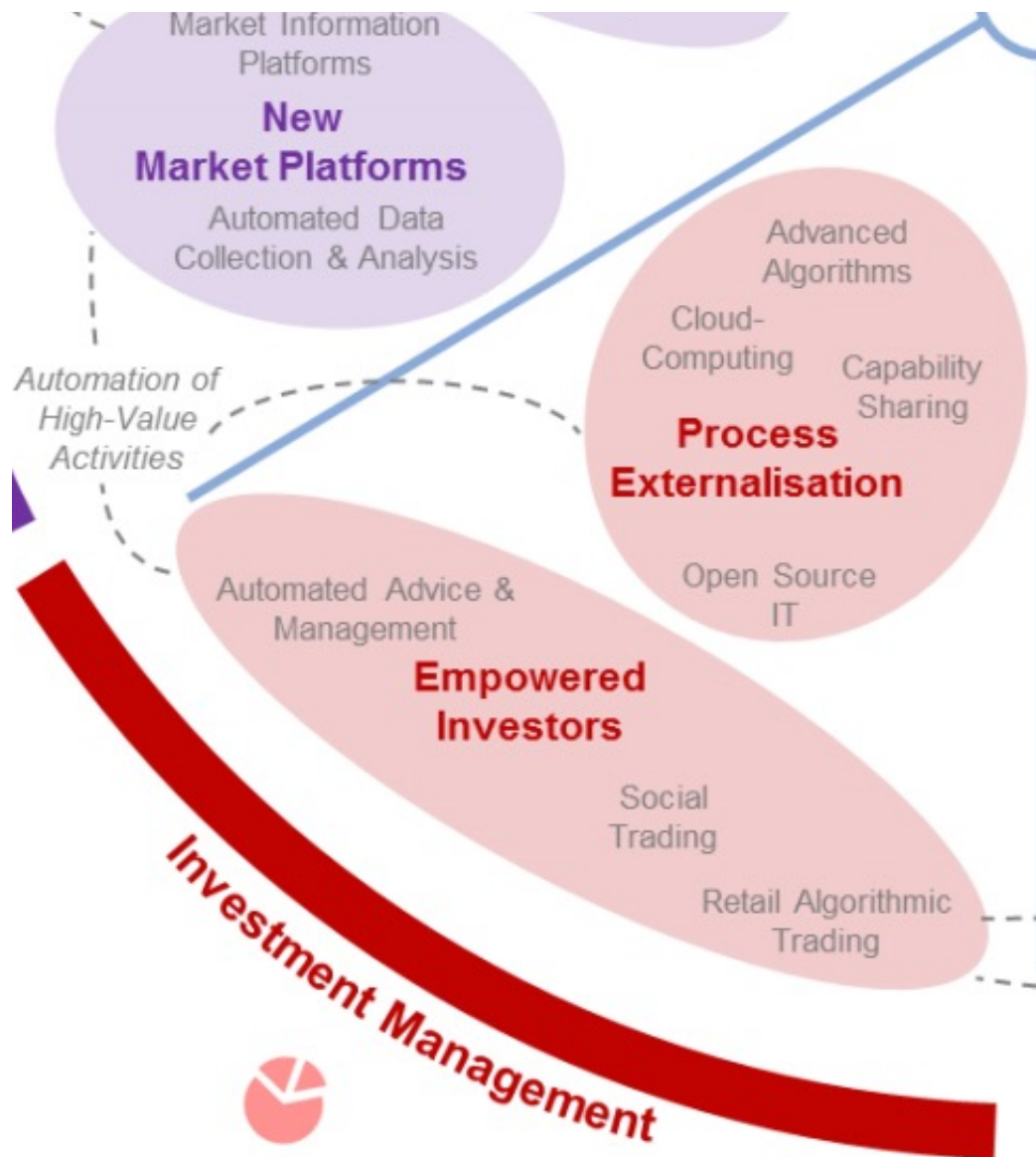


# FinTech:

## Financial Services Innovation

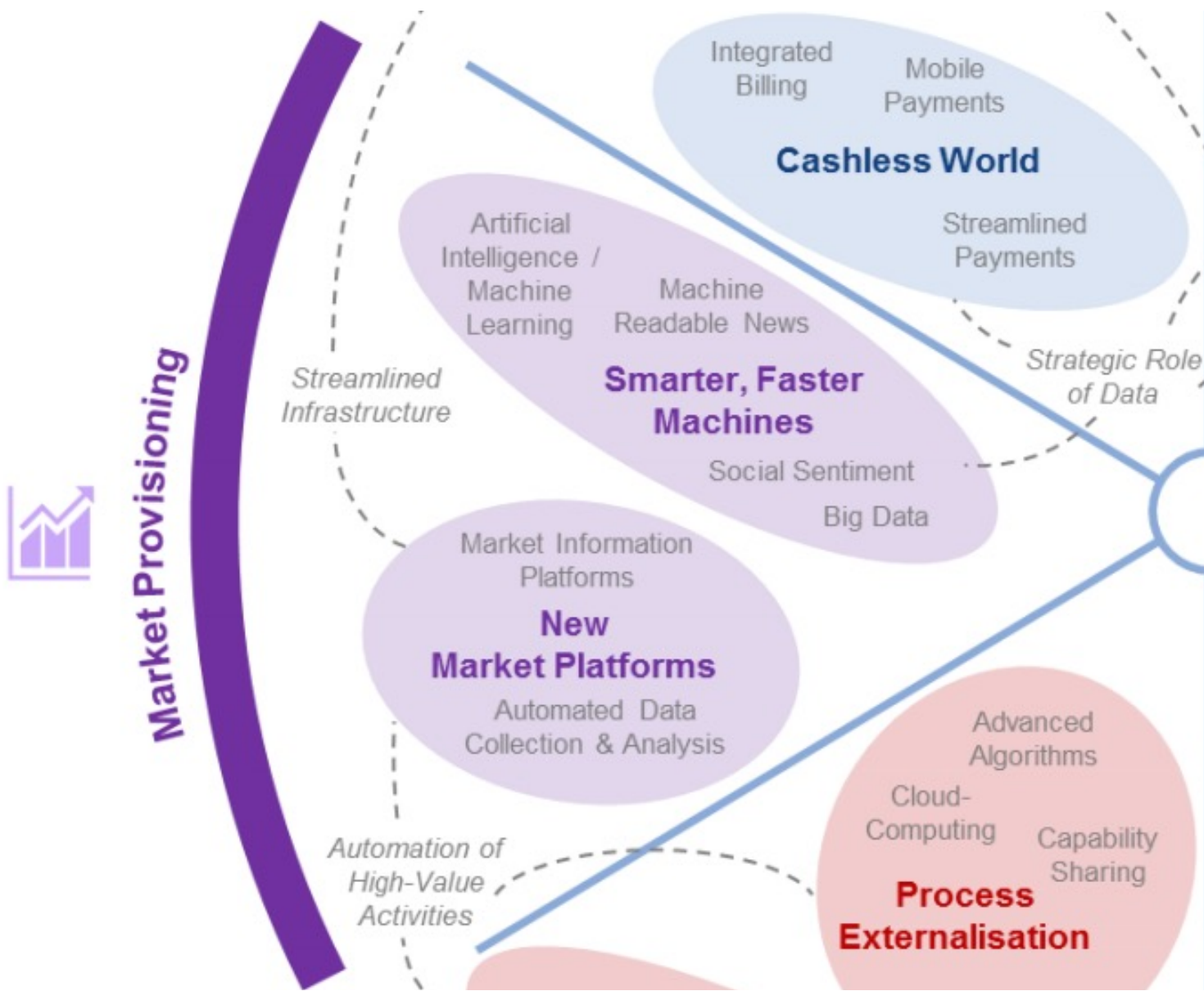
1. Payments
2. Insurance
3. Deposits & Lending
4. Capital Raising
5. Investment Management
6. Market Provisioning

# 5 FinTech: Investment Management



# 6

# FinTech: Market Provisioning





# The New Alpha: 30+ Startups Providing Alternative Data For Sophisticated Investors

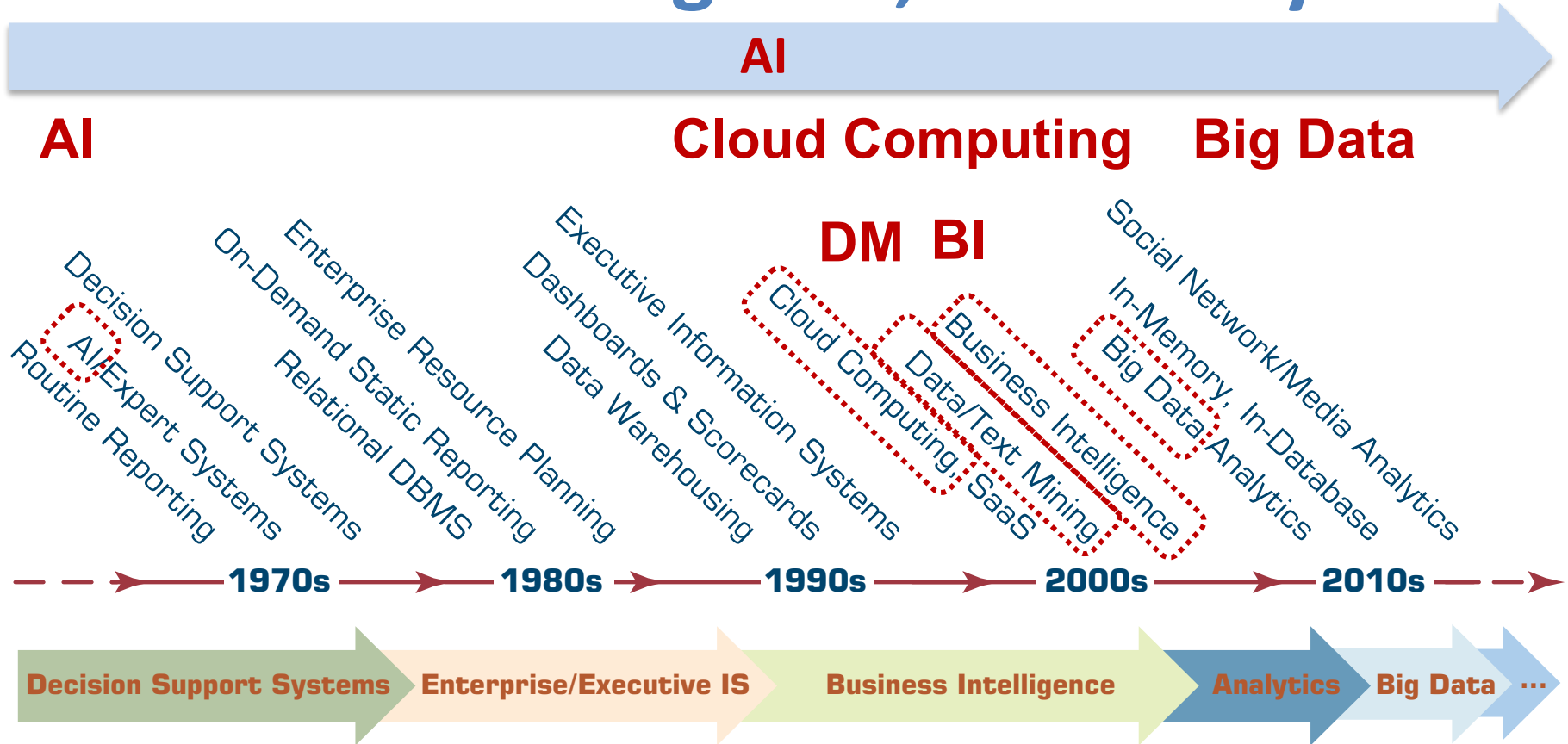
## Alternative Data Sources



**AI in**

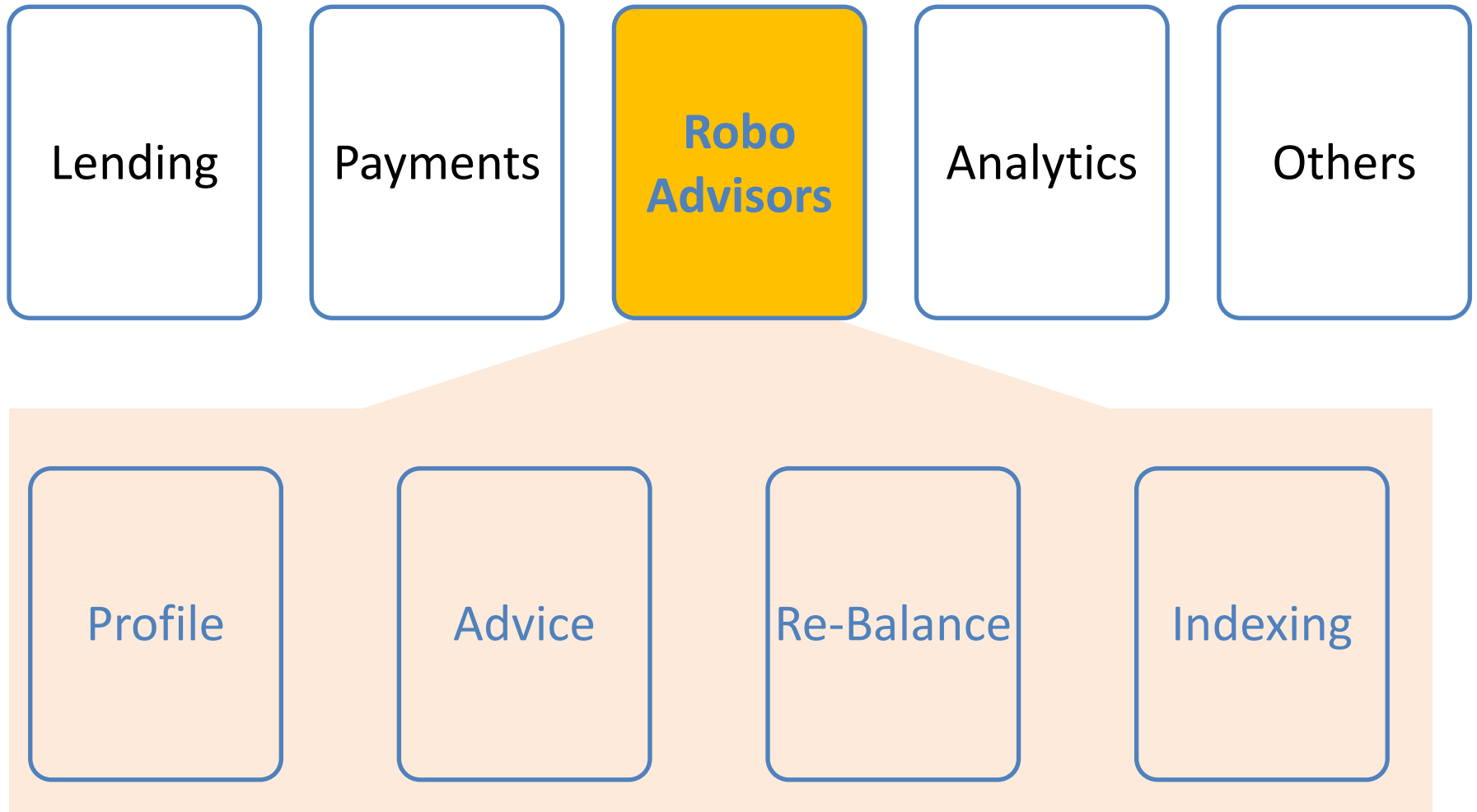
**FinTech**

# AI, Big Data, Cloud Computing Evolution of Decision Support, Business Intelligence, and Analytics



# Robo-Advisors

# FinTech high-level classification



# Wealthfront

## Financial Planning & Robo-Investing for Millennials



Plan ▾

Invest

Borrow

Expertise ▾

LOG IN

GET STARTED

WEALTHFRONT

# Meet your financial copilot

We'll build a free financial plan for the life you want and automate your investments at a low cost.

Our all-in-one solution gives you the financial expertise you need, right in your pocket. No spreadsheets, no annoying sales calls, no judgment.

GET STARTED



<https://www.wealthfront.com/>

# Betterment

## Online Financial Advisor

Betterment

Why Betterment

How it works

What we offer

Pricing

FAQs



Log in

Get started

# HELLO, INVESTOR

Betterment is an online financial advisor built for people who refuse to settle for average investing. People who demand better. People like you.

Get started

▶ Watch our video

### Right for every type of investor

#### New investor

I'm new to investing, or am looking for some guidance.

#### Hands-off investor

I invest, but don't have the time or desire to do it myself.

#### Hands-on investor

I'm a confident, hands-on investor looking for an optimal solution.

<https://www.betterment.com/>

# Financial Advisor FinTech Solutions Map

## Financial Advisor FinTech Solutions Map



Legend:   = Category Newcomer

Nerd's Eye View | kitces.com

Source: <https://www.kitces.com/fintechmap>



**From Algorithmic Trading  
to Personal Finance Bots:  
41 Startups Bringing  
AI to Fintech**

# From Algorithmic Trading To Personal Finance Bots: 41 Startups Bringing AI To Fintech

## AI in Fintech

41 Startups Bringing Artificial Intelligence To Fintech

General Purpose/ Predictive Analytics



Market Research & Sentiment Analysis



Search Engine



Quantitative Trading



Blockchain



Debt Collection



AI Assistants/Bots



Fraud Detection



Credit Scoring



Personal Banking



# Artificial Intelligence (AI) in Fintech

## General Purpose/ Predictive Analytics



## Market Research & Sentiment Analysis



## Search Engine



# Artificial Intelligence (AI) in Fintech

## Quantitative Trading



## Blockchain



## Debt Collection



## AI Assistants/Bots



## Fraud Detection



## Credit Scoring



## Personal Banking



# Conversational Commerce

**Chatbot**  
**Dialogue System**  
**Intelligent Agent**

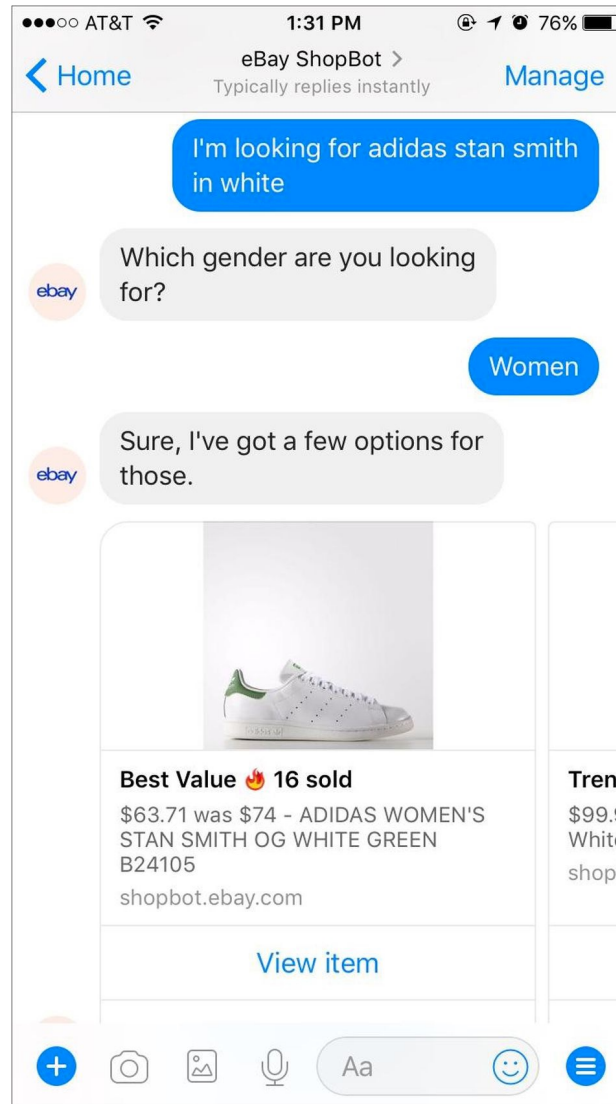
# Chatbots: Evolution of UI/UX

Paradigm	mid - 80s PC	mid - 90s Web	mid - 00s Smartphone	mid - 10s Messaging
Platform <i>Examples</i>	Desktop DOS, Windows, Mac OS	Browser Mosaic, Explorer, Chrome	Mobile OS iOS, Android	Messaging Apps WhatsApp, Messenger, Slack
Applications <i>Examples</i>	Clients Excel, PPT, Lotus	Website Yahoo, Amazon	Apps Angry Birds, Instagram	Bots Weather, Travel
UI/UX	Native Screens	Web Pages	Native Mobile Screens	Message
S/w Dev	Client-side	Server-side	Client-side	Server-side

**From  
E-Commerce  
to  
Conversational Commerce:  
Chatbots  
and  
Virtual Assistants**

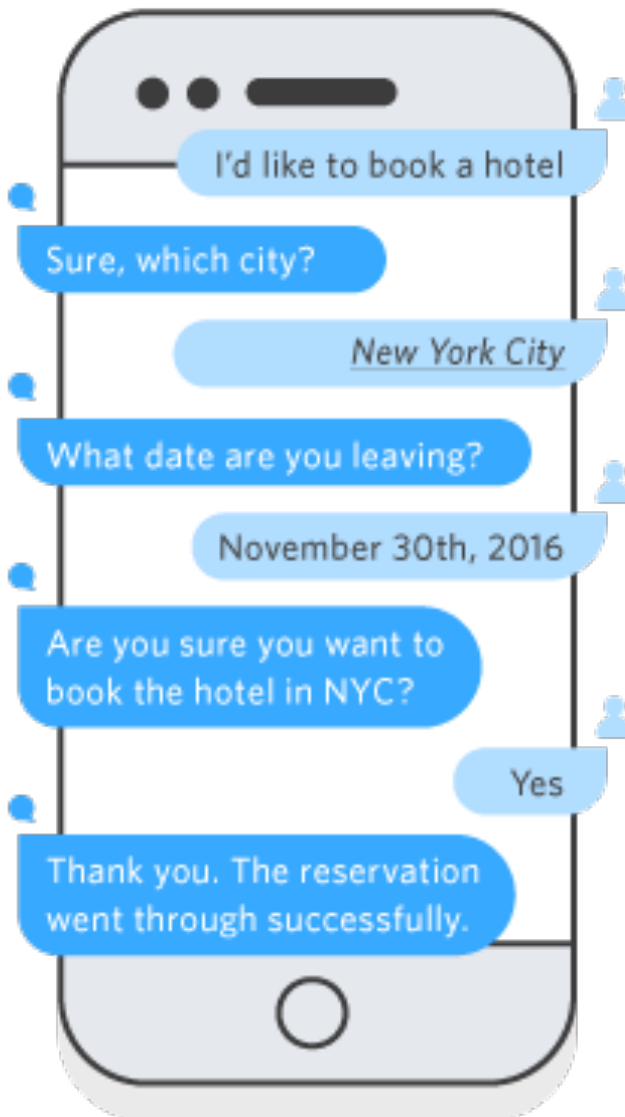


# Conversational Commerce: eBay AI Chatbots



# Hotel Chatbot

## BookHotel



### Intents

An intent performs an action in response to natural language user input

**Intent  
Detection**

### Utterances

Spoken or typed phrases that invoke your intent

### Slots

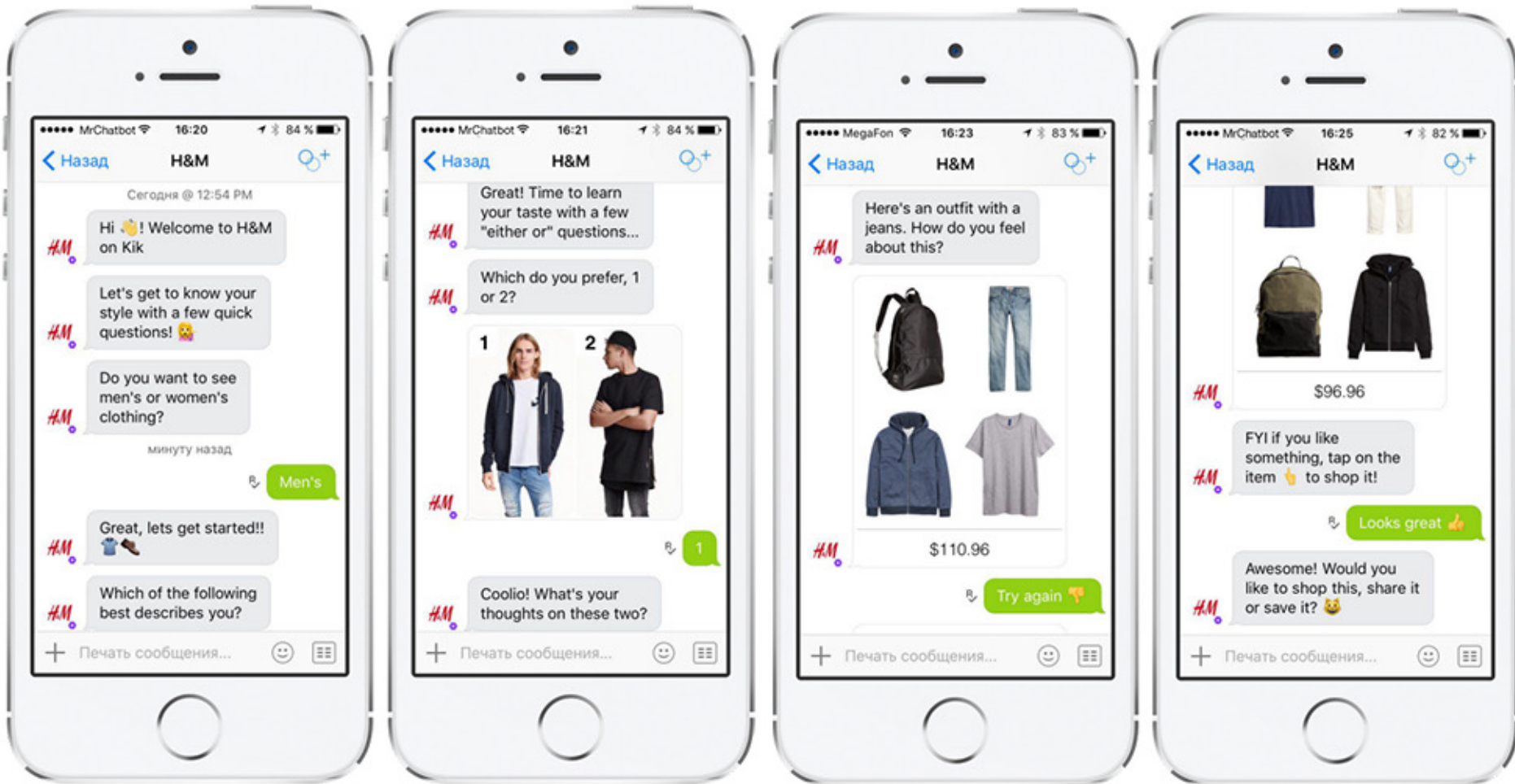
Slots are input data required to fulfill the intent

**Slot Filling**

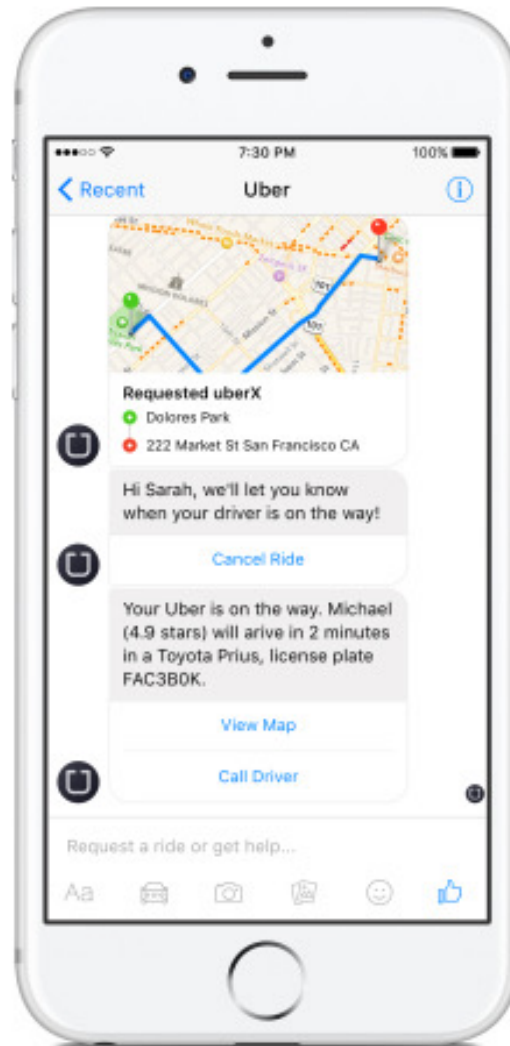
### Fulfillment

Fulfillment mechanism for your intent

# H&M's Chatbot on Kik



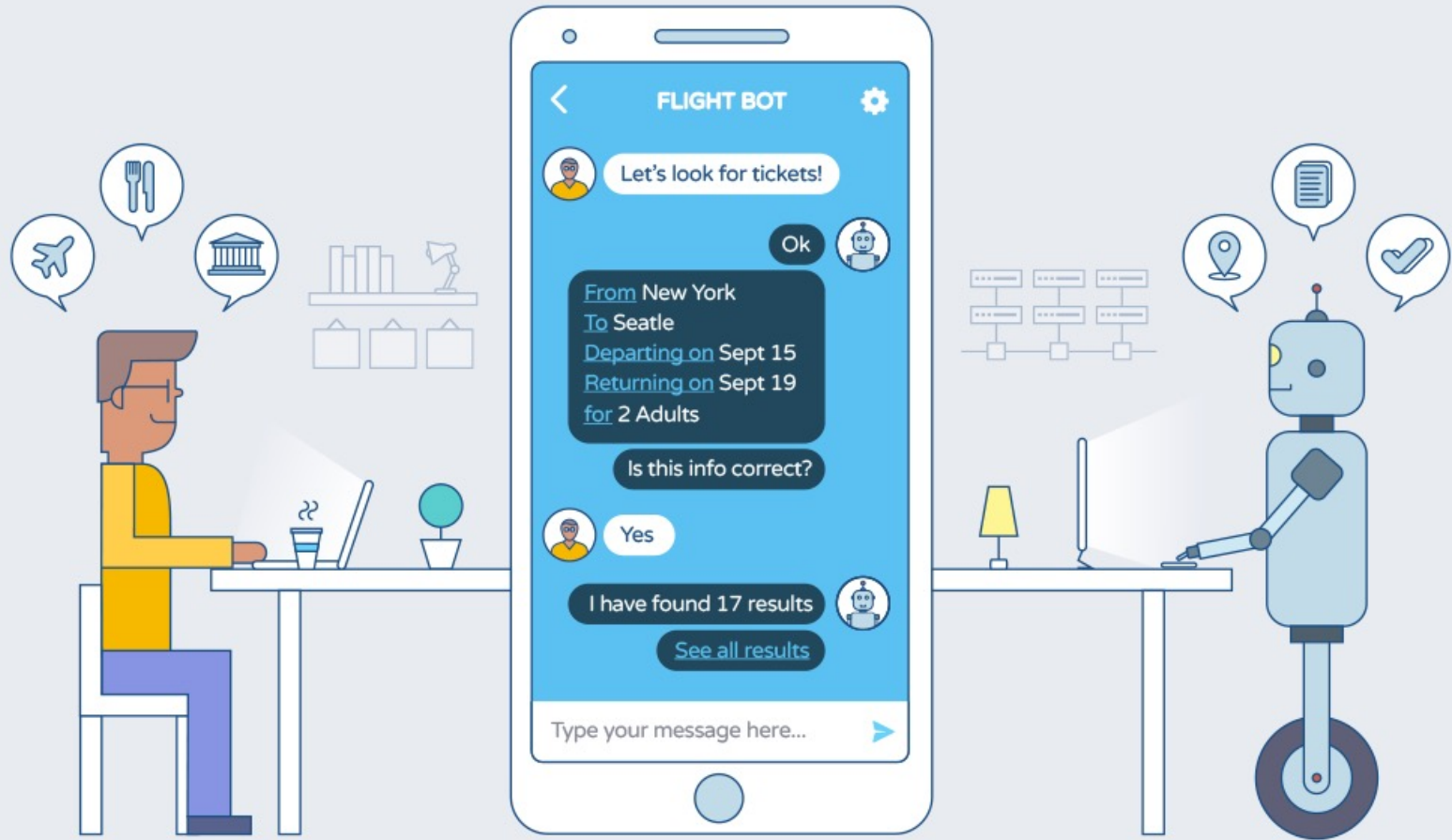
# Uber's Chatbot on Facebook's Messenger



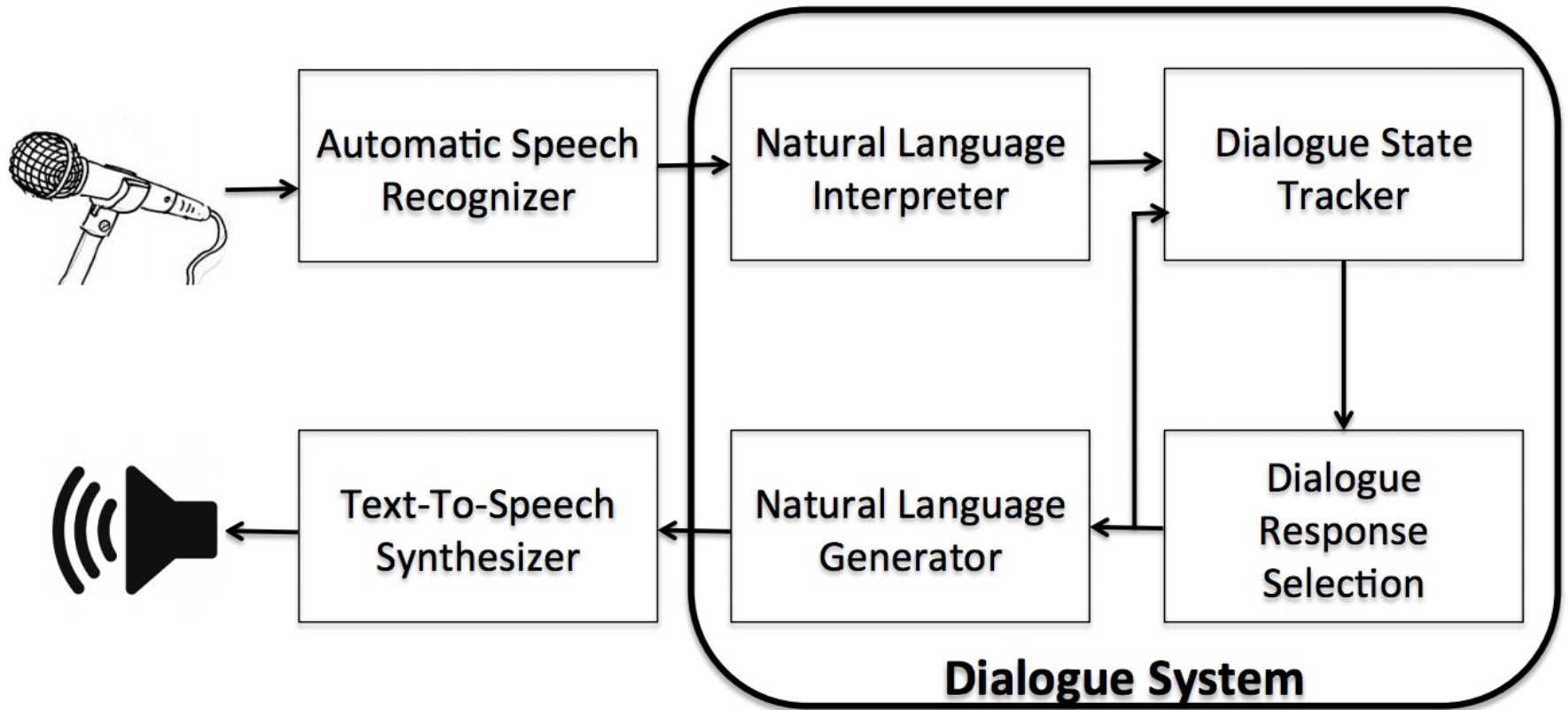
Uber's chatbot on Facebook's messenger  
- one main benefit: it loads much faster than the Uber app

Source: <http://www.guided-selling.org/from-e-commerce-to-conversational-commerce/>

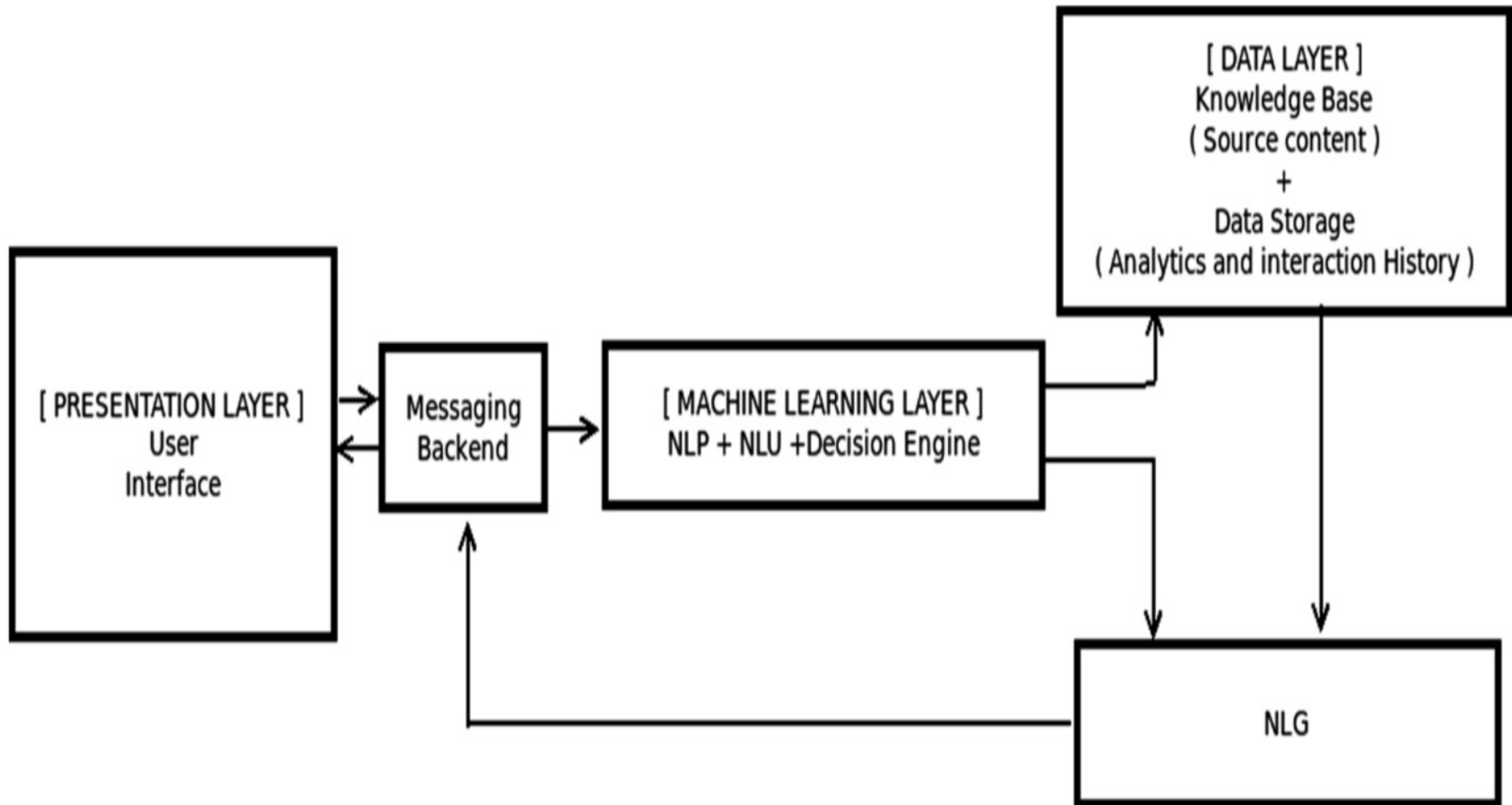
# Chatbot



# Dialogue System



# Overall Architecture of Intelligent Chatbot



# Can machines think?

**(Alan Turing ,1950)**

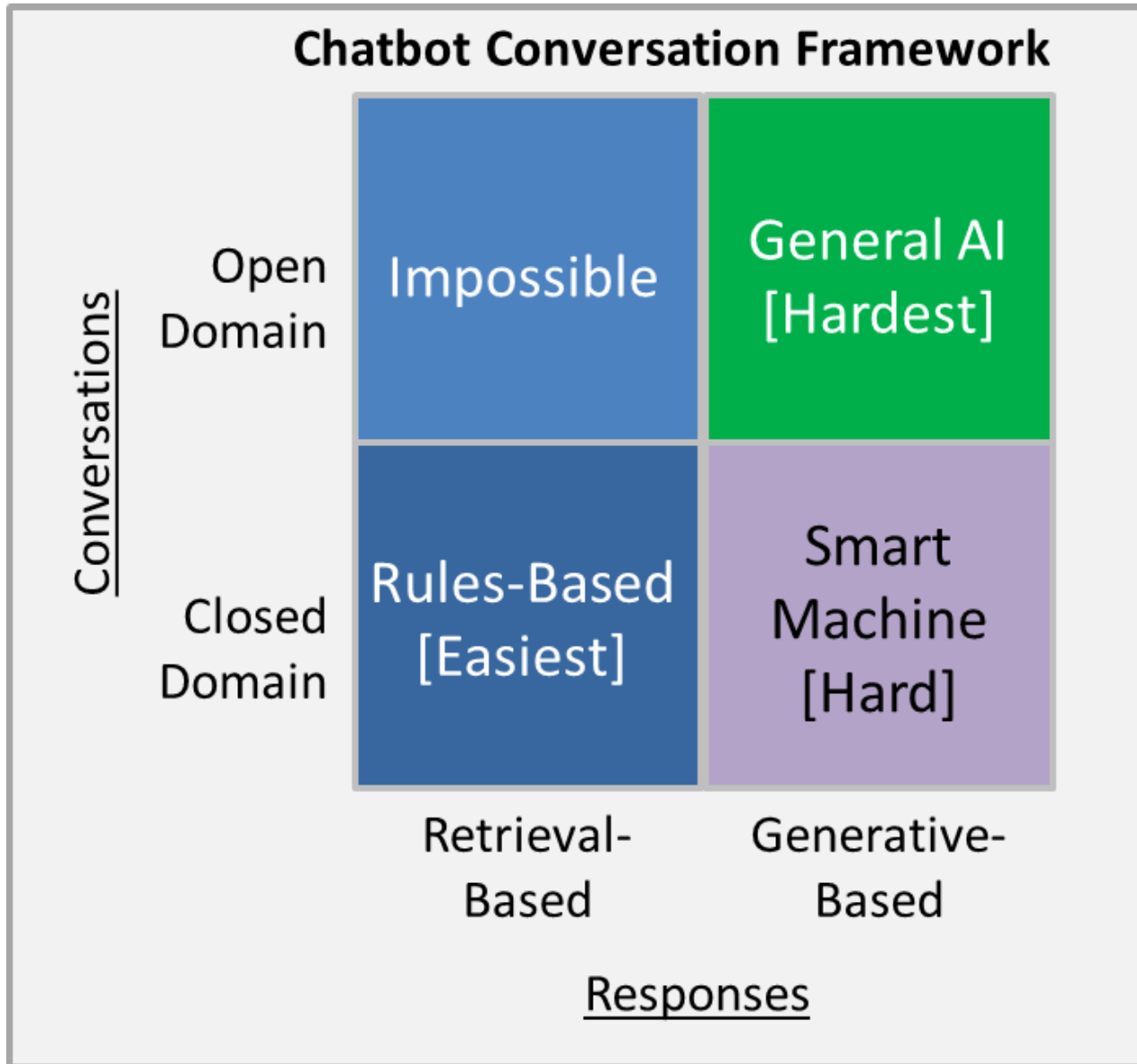
Source: Cahn, Jack. "CHATBOT: Architecture, Design, & Development."  
PhD diss., University of Pennsylvania, 2017.



# Chatbot

**“online human-computer  
dialog system  
with  
natural language.”**

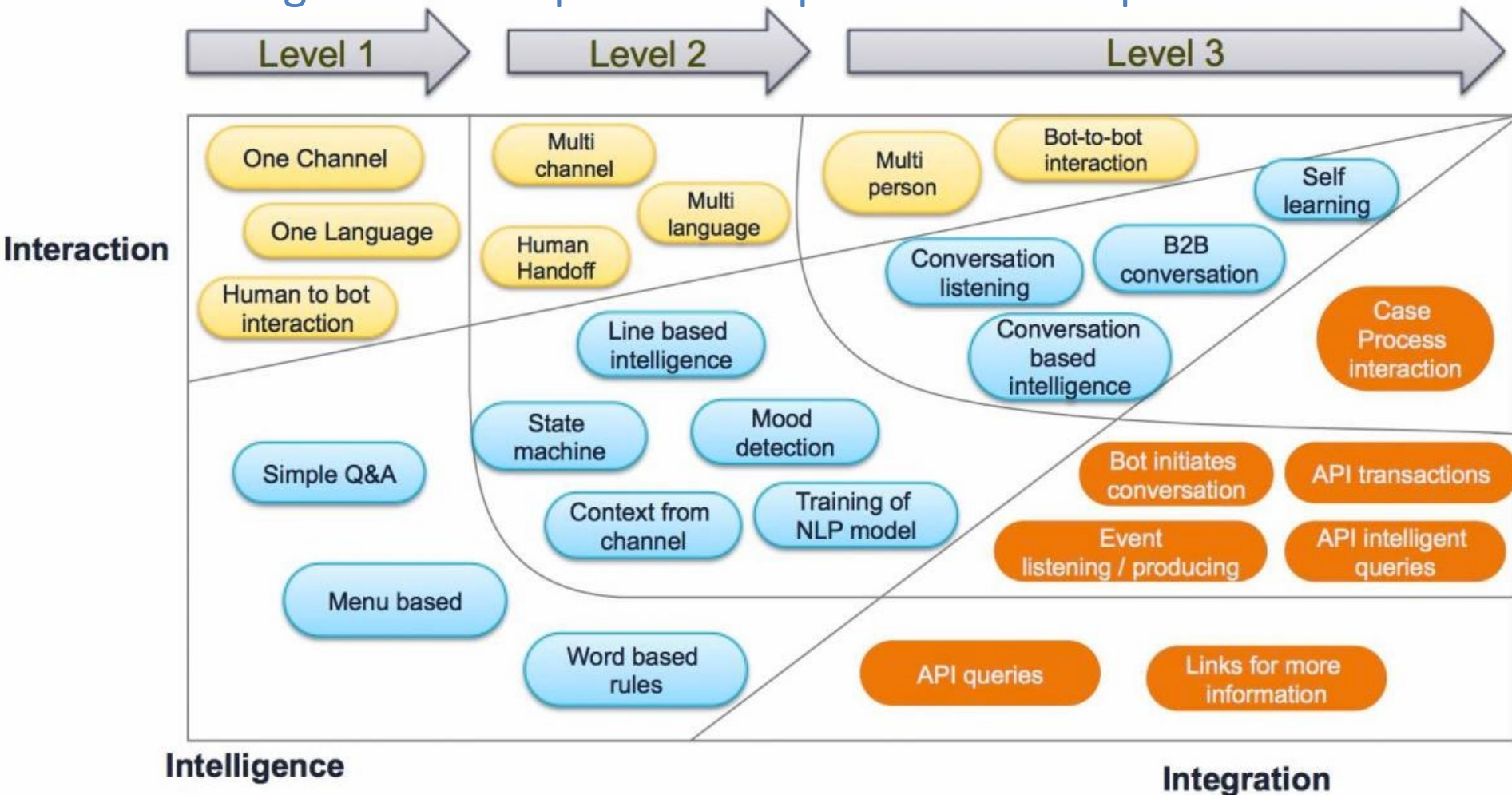
# Chatbot Conversation Framework



# Chatbots

## Bot Maturity Model

Customers want to have simpler means to interact with businesses and get faster response to a question or complaint.



# **Task-Oriented Dialogue System**

# Dialogue Subtasks

Browse SoTA > Natural Language Processing > Dialogue

## Dialogue subtasks

### Dialogue Generation

Dialogue Generation

📄 9 benchmarks

78 papers with code



Dialogue State Tracking

📄 2 benchmarks

51 papers with code

### Task-Oriented Dialogue Systems

Task-Oriented Dialogue Systems

📄 2 benchmarks

48 papers with code



Visual Dialog

📄 8 benchmarks

37 papers with code



Goal-Oriented Dialogue

📄 1 benchmark

20 papers with code



Dialogue Management

12 papers with code



Dialogue Understanding

📄 11 benchmarks

8 papers with code



Dialogue Act Classification

📄 2 benchmarks

8 papers with code

### Short-Text Conversation

Short-Text Conversation

7 papers with code

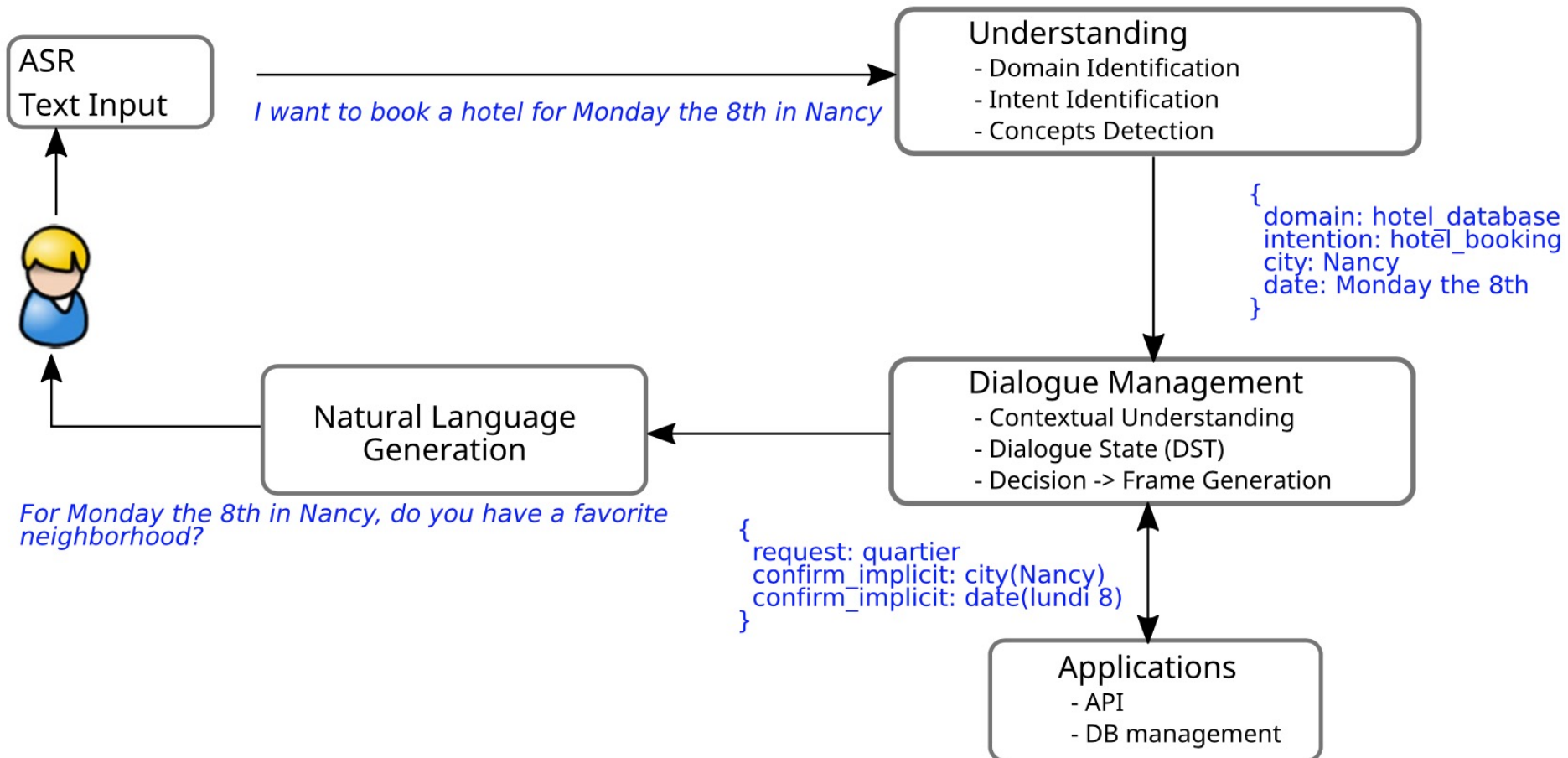


Goal-Oriented Dialogue Systems

7 papers with code

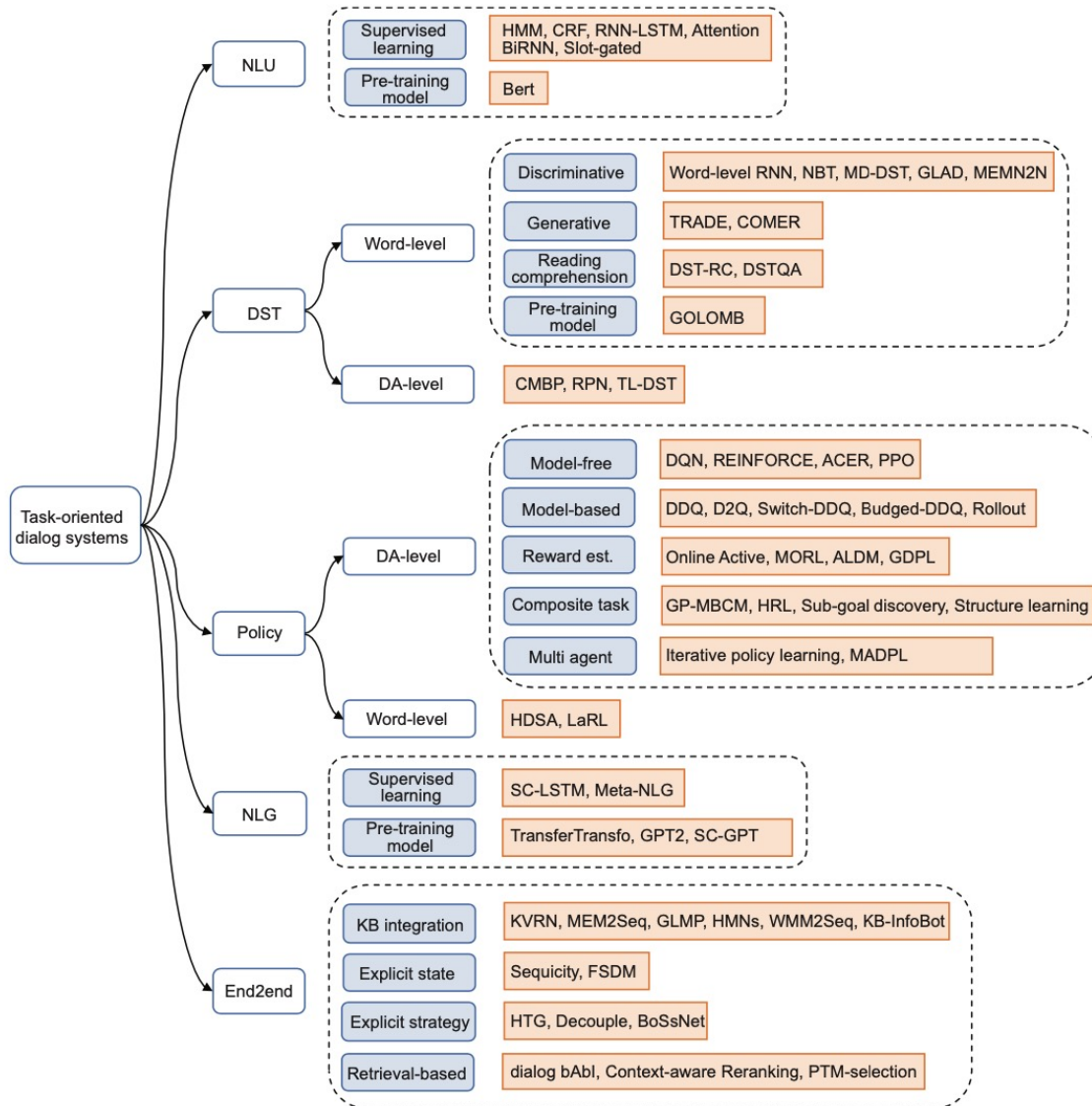
# Task-Oriented Dialogue System

(Deriu et al., 2021)



# Task-Oriented Dialogue Systems

(Zhang et al., 2020)



Source: Zhang, Zheng, Ryuichi Takanobu, Qi Zhu, Minlie Huang, and Xiaoyan Zhu (2020).

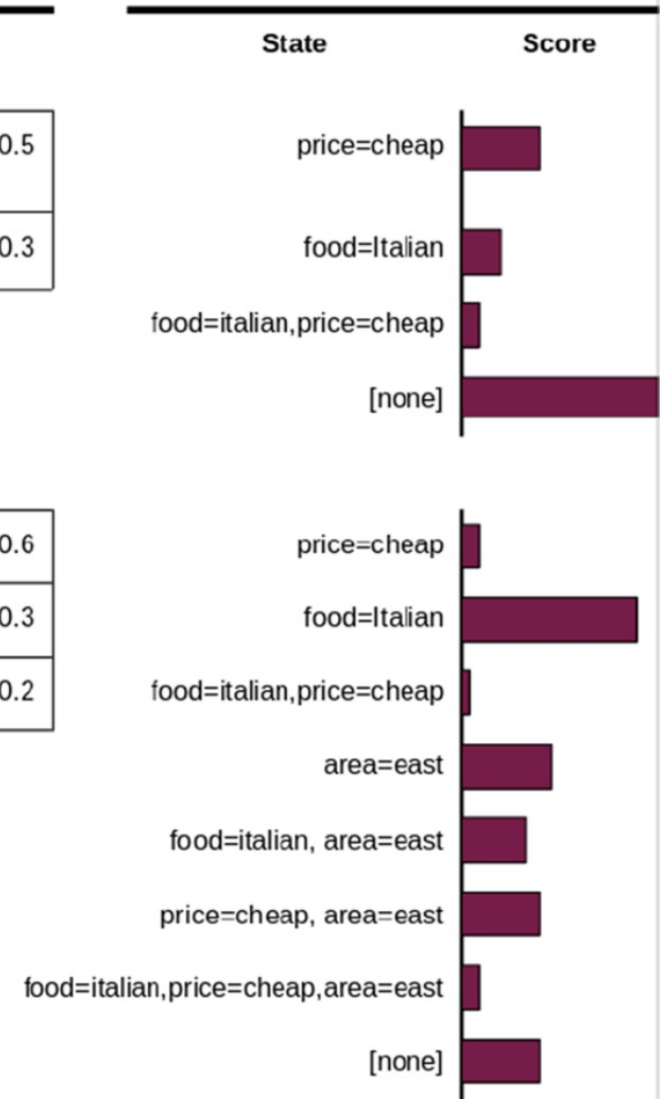
"Recent advances and challenges in task-oriented dialog systems." Science China Technological Sciences (2020): 1-17.

# Dialog State Tracker (DST)

Dialog state tracker inputs

System action/ user response	ASR output	SLU output
How can I help you? <i>welcome()</i>	Cheap restaurant 0.6	inform(price=cheap) 0.5
	Restaurant 0.2	inform(food=italian) 0.3
	Italian 0.1	
An Italian restaurant		
What price did you want? <i>request(price)</i>	East Area 0.5	inform(area=east) 0.6
	Italian 0.3	inform(food=italian) 0.3
	Yeah 0.1	affirm() 0.2
Uh, Italian		

Dialog state tracker outputs



Source: Deriu, Jan, Alvaro Rodrigo, Arantxa Otegi, Guillermo Echevoyen, Sophie Rosset, Eneko Agirre, and Mark Cieliebak (2021). "Survey on evaluation methods for dialogue systems." Artificial Intelligence Review 54, no. 1 (2021): 755-810.



# Dialogue Acts

(Young et al., 2010)

---

Dialogue act	Description
hello( $a = x, b = y, \dots$ )	Open a dialogue and give info $a = x, b = y, \dots$
inform( $a = x, b = y, \dots$ )	Give information $a = x, b = y, \dots$
request( $a, b = x, \dots$ )	Request value for $a$ given $b = x, \dots$
reqalts( $a = x, \dots$ )	Request alternative with $a = x, \dots$
confirm( $a = x, b = y, \dots$ )	Explicitly confirm $a = x, b = y, \dots$
confreq( $a = x, \dots, d$ )	Implicitly confirm $a = x, \dots$ and request value of $d$
select( $a = x, a = y$ )	Select either $a = x$ or $a = y$
affirm( $a = x, b = y$ )	Affirm and give further info $a = x, b = y, \dots$
negate( $a = x$ )	Negate and give corrected value $a = x$
deny( $a = x$ )	Deny that $a = x$
bye()	Close a dialogue

---

# Sample Dialogue Acts

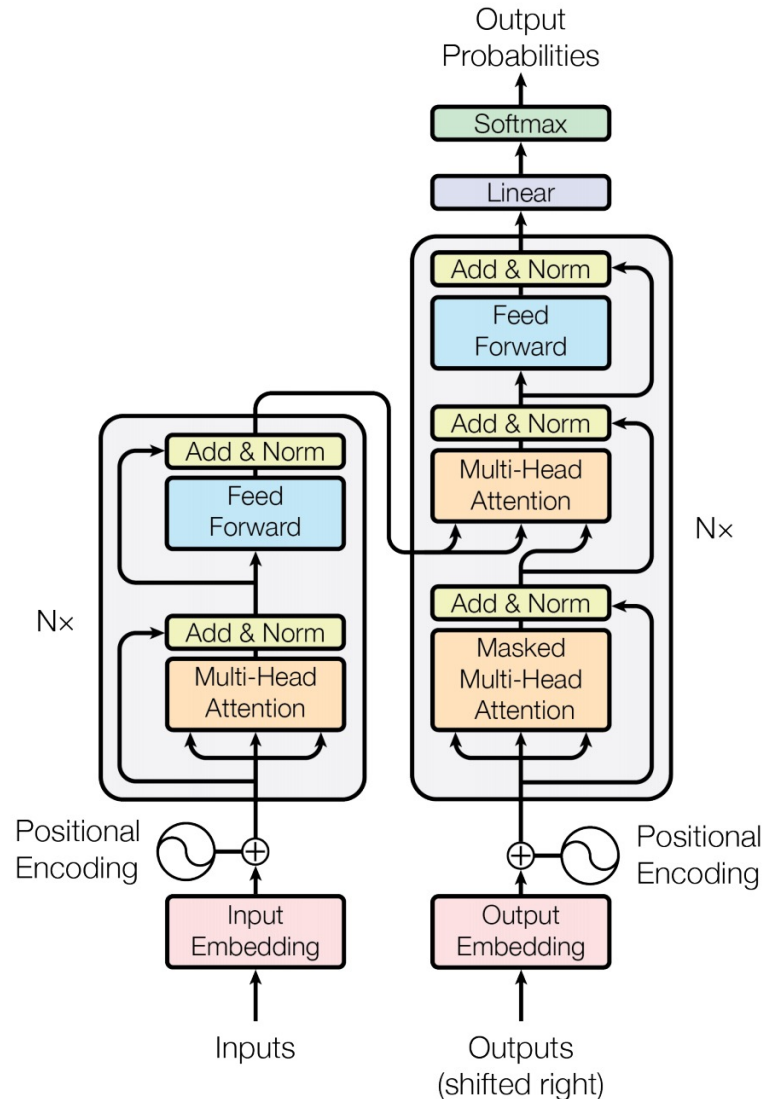
---

Utterance	Dialogue Act
U: Hi, I am looking for somewhere to eat	hello(task = find,type=restaurant)
S: You are looking for a restaurant. What type of food?	confreq(type = restaurant,food)
U: I'd like an Italian somewhere near the museum.	inform(food = Italian,near=museum)
S: Roma is a nice Italian restaurant near the museum.	inform(name = "Roma", type = restaurant, food = Italian, near = museum)
U: Is it reasonably priced?	confirm(pricerange = moderate)
S: Yes, Roma is in the moderate price range.	affirm(name = "Roma", pricerange = moderate)
U: What is the phone number?	request(phone)
S: The number of Roma is 385456.	inform(name = "Roma", phone = "385456")
U: Ok, thank you goodbye.	bye()

---

# Transformer (Attention is All You Need)

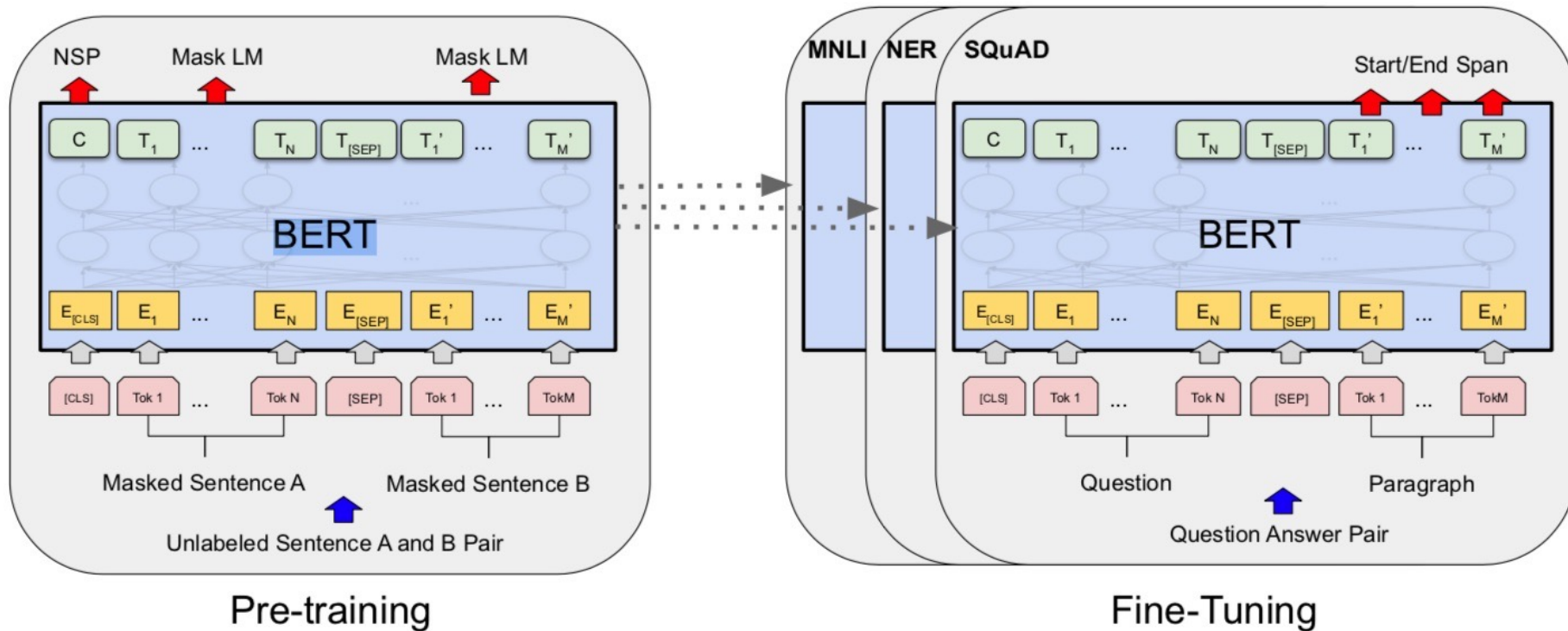
(Vaswani et al., 2017)



# BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding

BERT (Bidirectional Encoder Representations from Transformers)

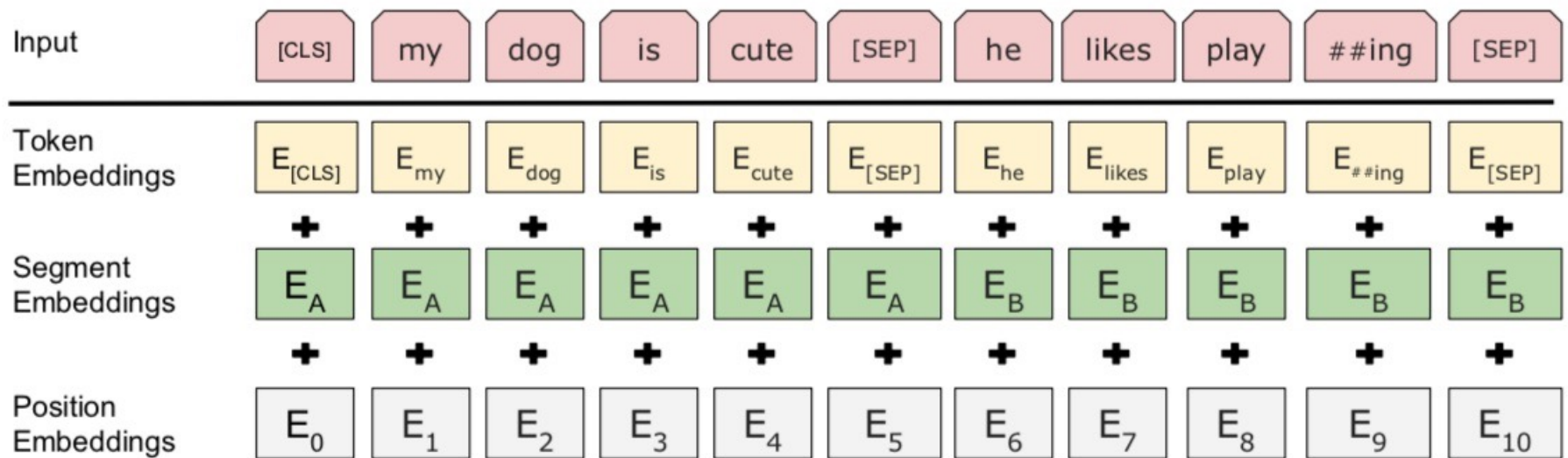
Overall pre-training and fine-tuning procedures for BERT



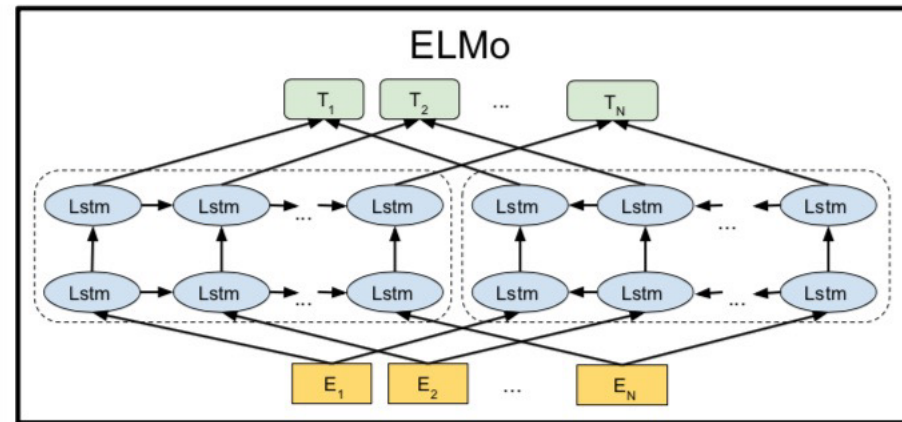
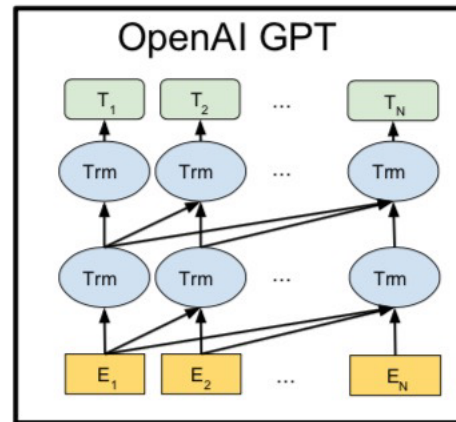
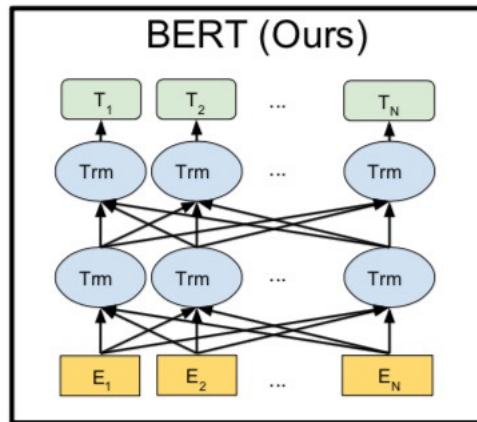
# BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding

BERT (Bidirectional Encoder Representations from Transformers)

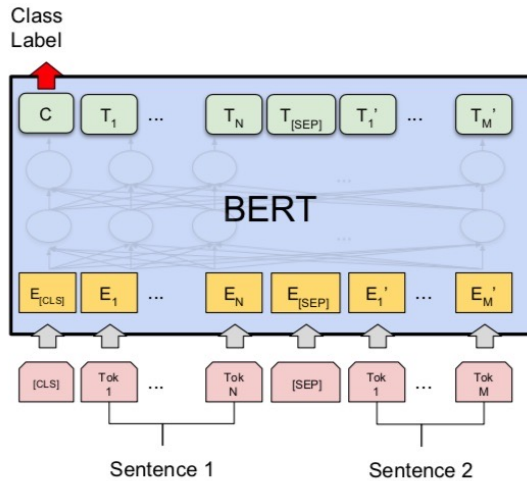
## BERT input representation



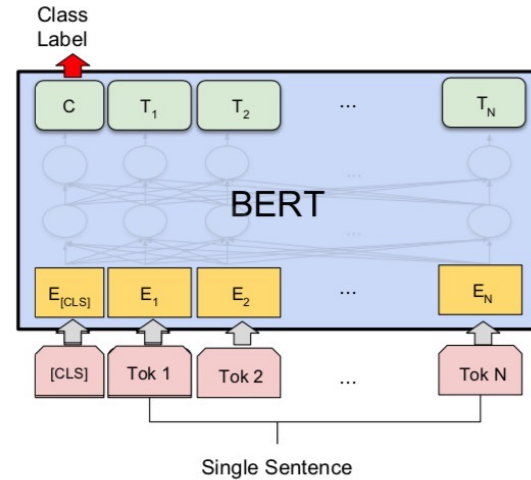
# BERT, OpenAI GPT, ELMo



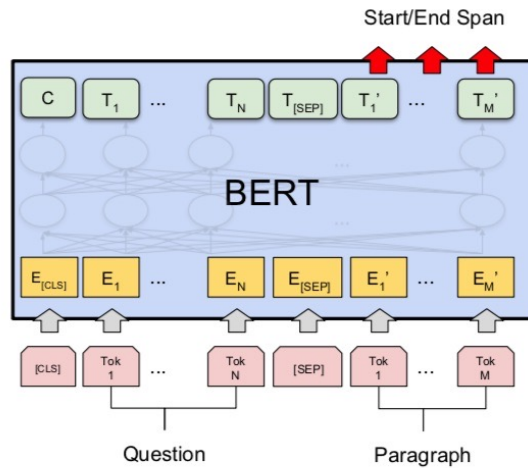
# Fine-tuning BERT on Different Tasks



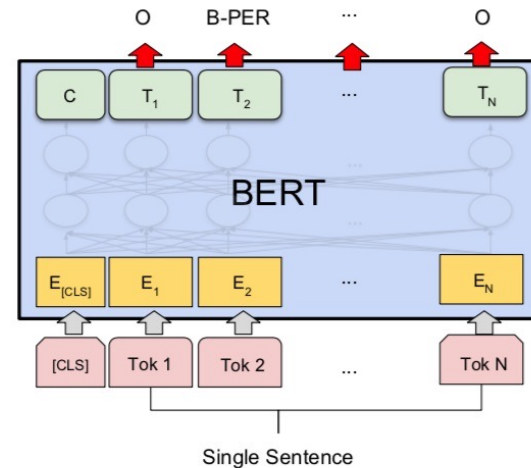
(a) Sentence Pair Classification Tasks:  
MNLI, QQP, QNLI, STS-B, MRPC,  
RTE, SWAG



(b) Single Sentence Classification Tasks:  
SST-2, CoLA



(c) Question Answering Tasks:  
SQuAD v1.1

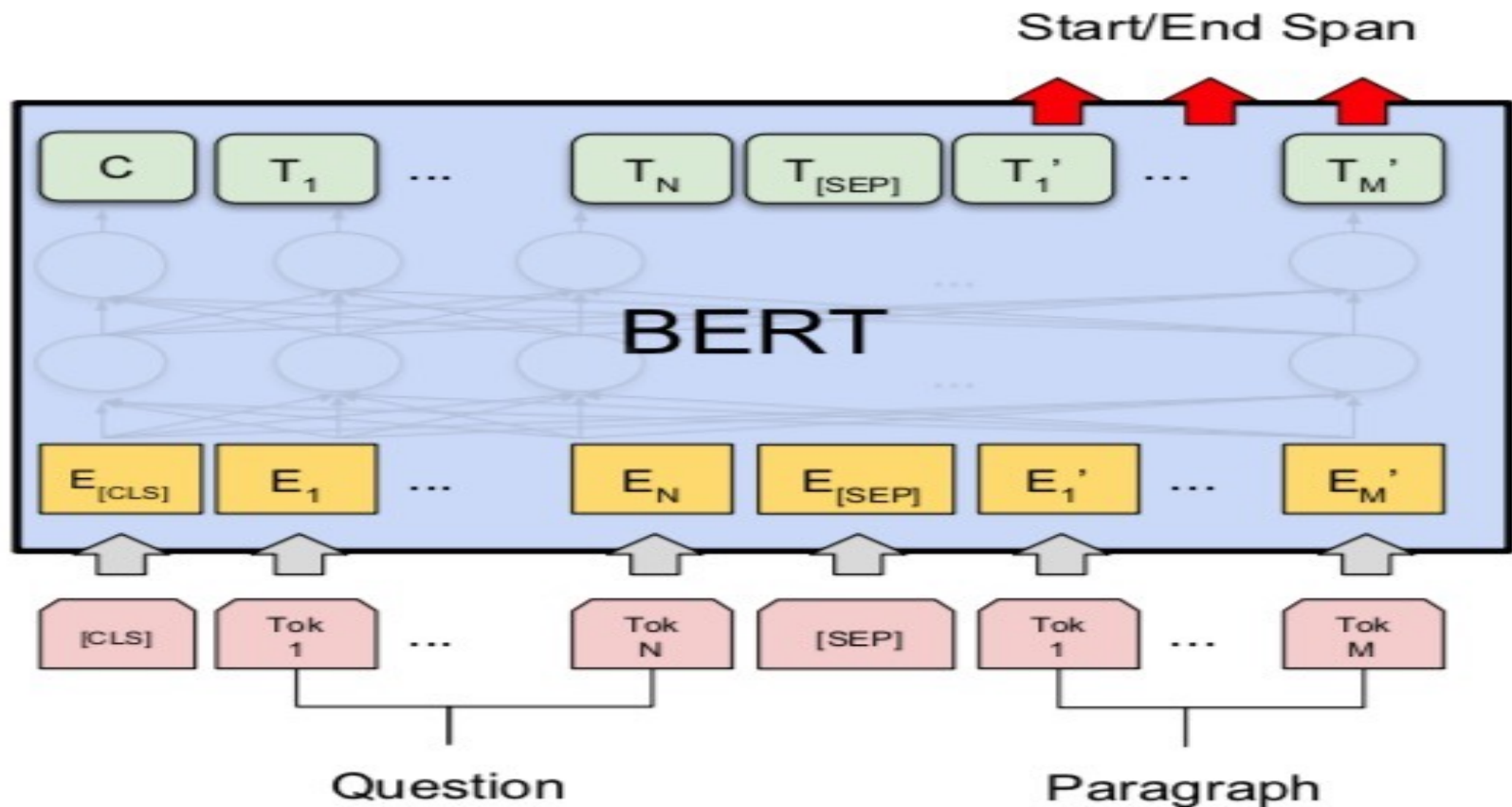


(d) Single Sentence Tagging Tasks:  
CoNLL-2003 NER

Source: Devlin, Jacob, Ming-Wei Chang, Kenton Lee, and Kristina Toutanova (2018).

"Bert: Pre-training of deep bidirectional transformers for language understanding." arXiv preprint arXiv:1810.04805.

# Fine-tuning BERT on Question Answering (QA)

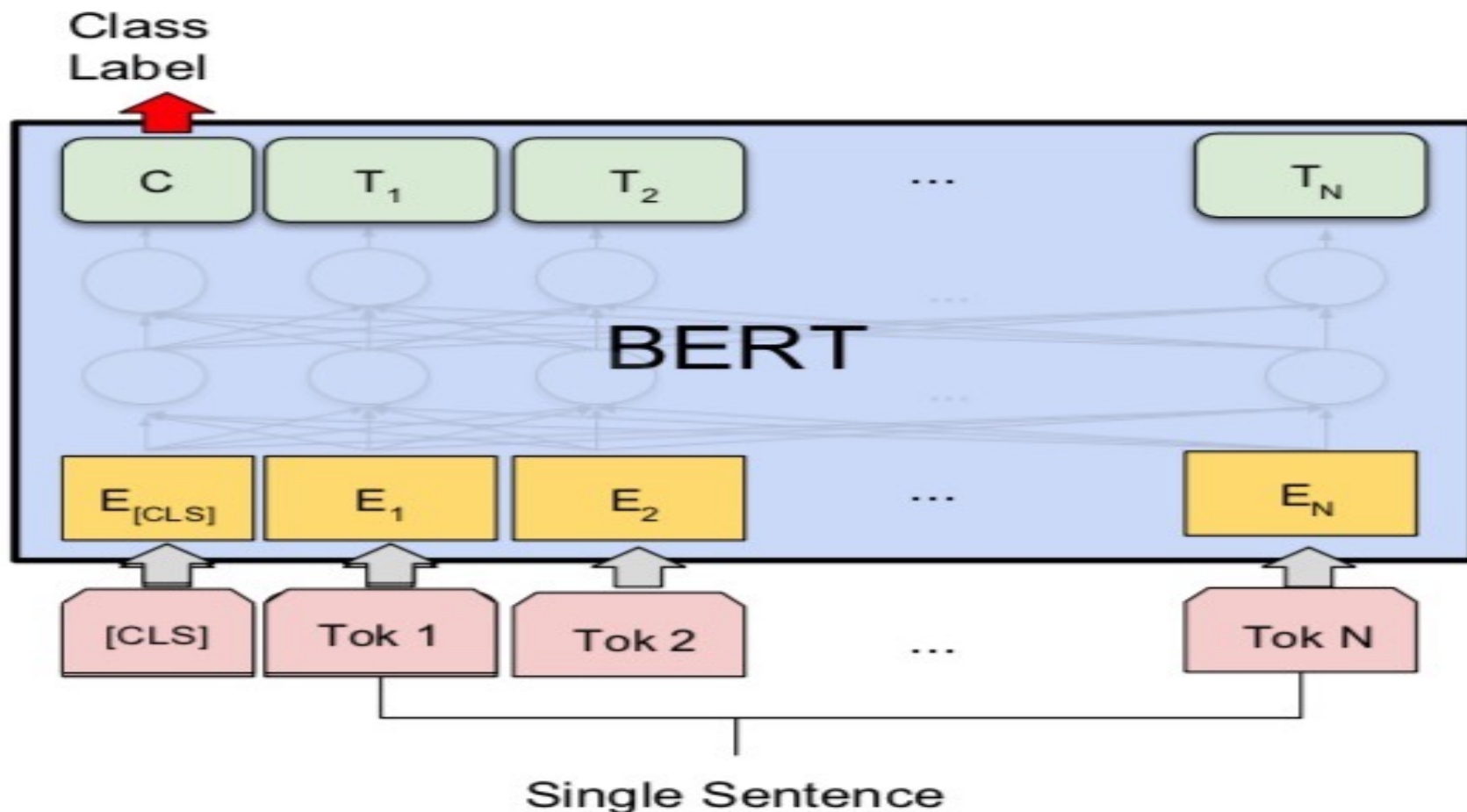


(c) Question Answering Tasks:  
SQuAD v1.1



# Fine-tuning BERT on Dialogue

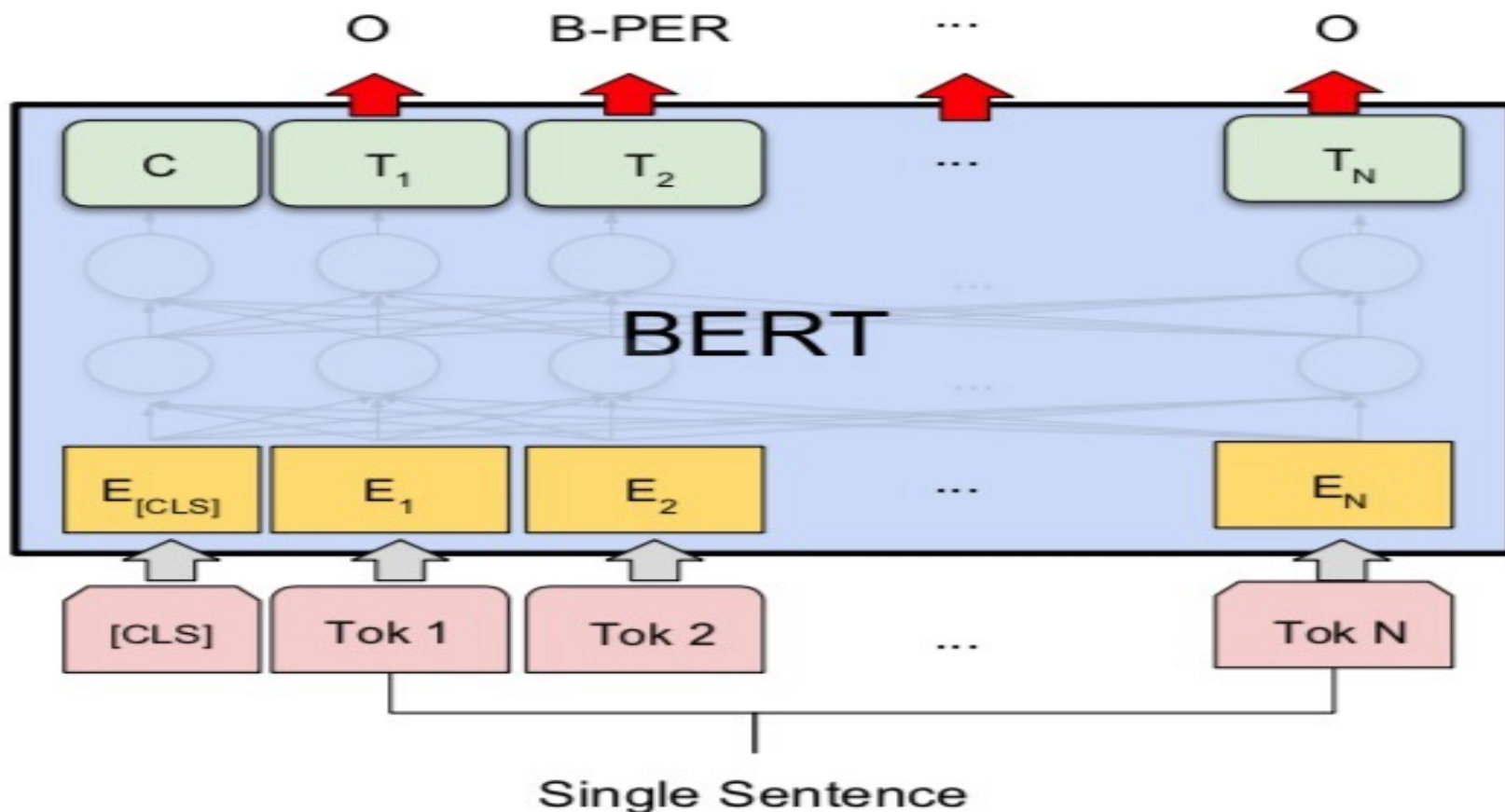
## Intent Detection (ID; Classification)



(b) Single Sentence Classification Tasks: SST-2, CoLA

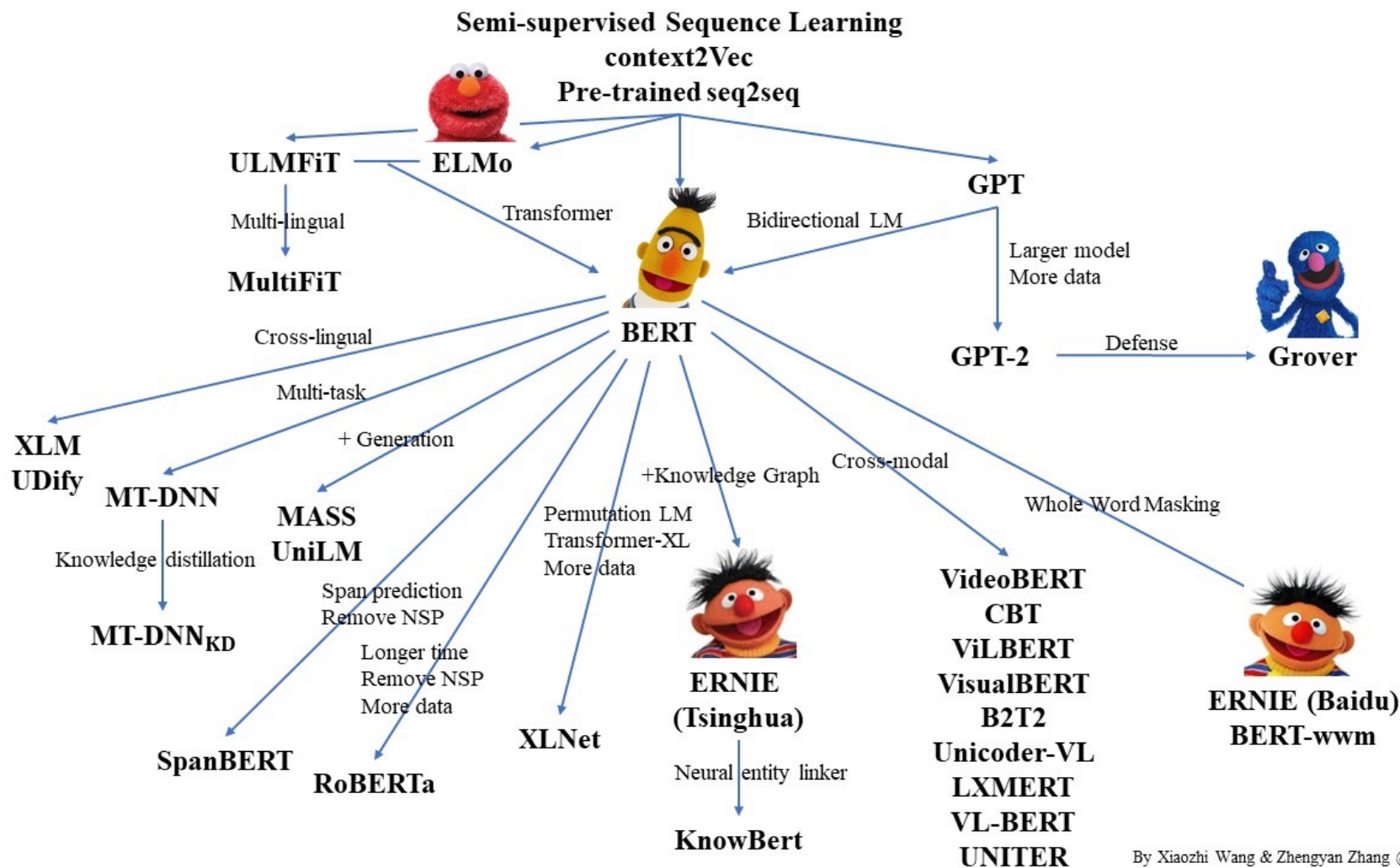
# Fine-tuning BERT on Dialogue

## Slot Filling (SF)



(d) Single Sentence Tagging Tasks:  
CoNLL-2003 NER

# Pre-trained Language Model (PLM)



By Xiaozhi Wang & Zhengyan Zhang @THUNLP

# Transformers Transformers

## State-of-the-art Natural Language Processing for TensorFlow 2.0 and PyTorch

- Transformers
  - pytorch-transformers
  - pytorch-pretrained-bert
- provides state-of-the-art general-purpose architectures
  - (BERT, GPT-2, RoBERTa, XLM, DistilBert, XLNet, CTRL...)
  - for Natural Language Understanding (NLU) and Natural Language Generation (NLG) with over 32+ pretrained models in 100+ languages and deep interoperability between TensorFlow 2.0 and PyTorch.

**Dialogue  
on  
Airline Travel  
Information System  
(ATIS)**

# The ATIS (Airline Travel Information System) Dataset

<https://www.kaggle.com/siddhadev/atis-dataset-from-ms-cntk>

<b>Sentence</b>	what	flights	leave	from	phoenix
<b>Slots</b>	O	O	O	O	B-fromloc
<b>Intent</b>	atis_flight				

Training samples: 4978

Testing samples: 893

Vocab size: 943

Slot count: 129

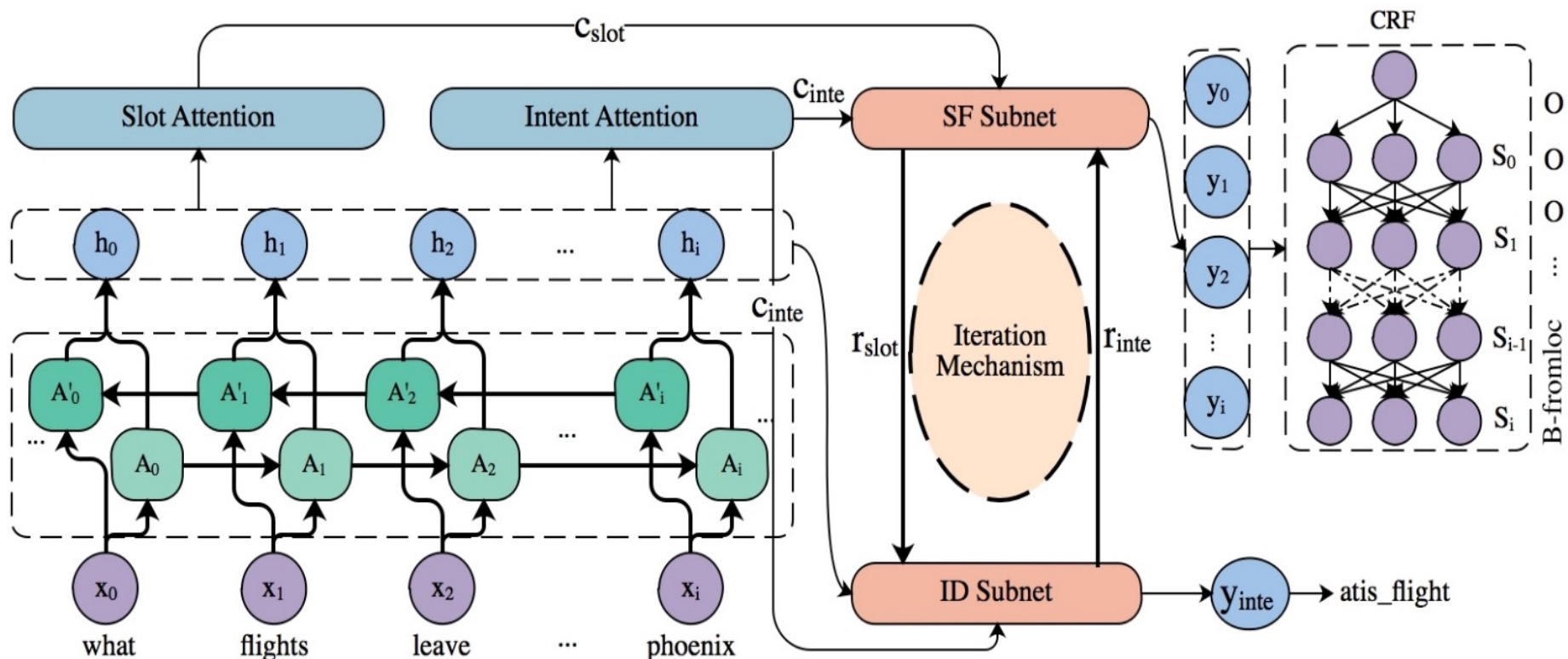
Intent count: 26

# SF-ID Network (E et al., 2019)

## Slot Filling (SF)

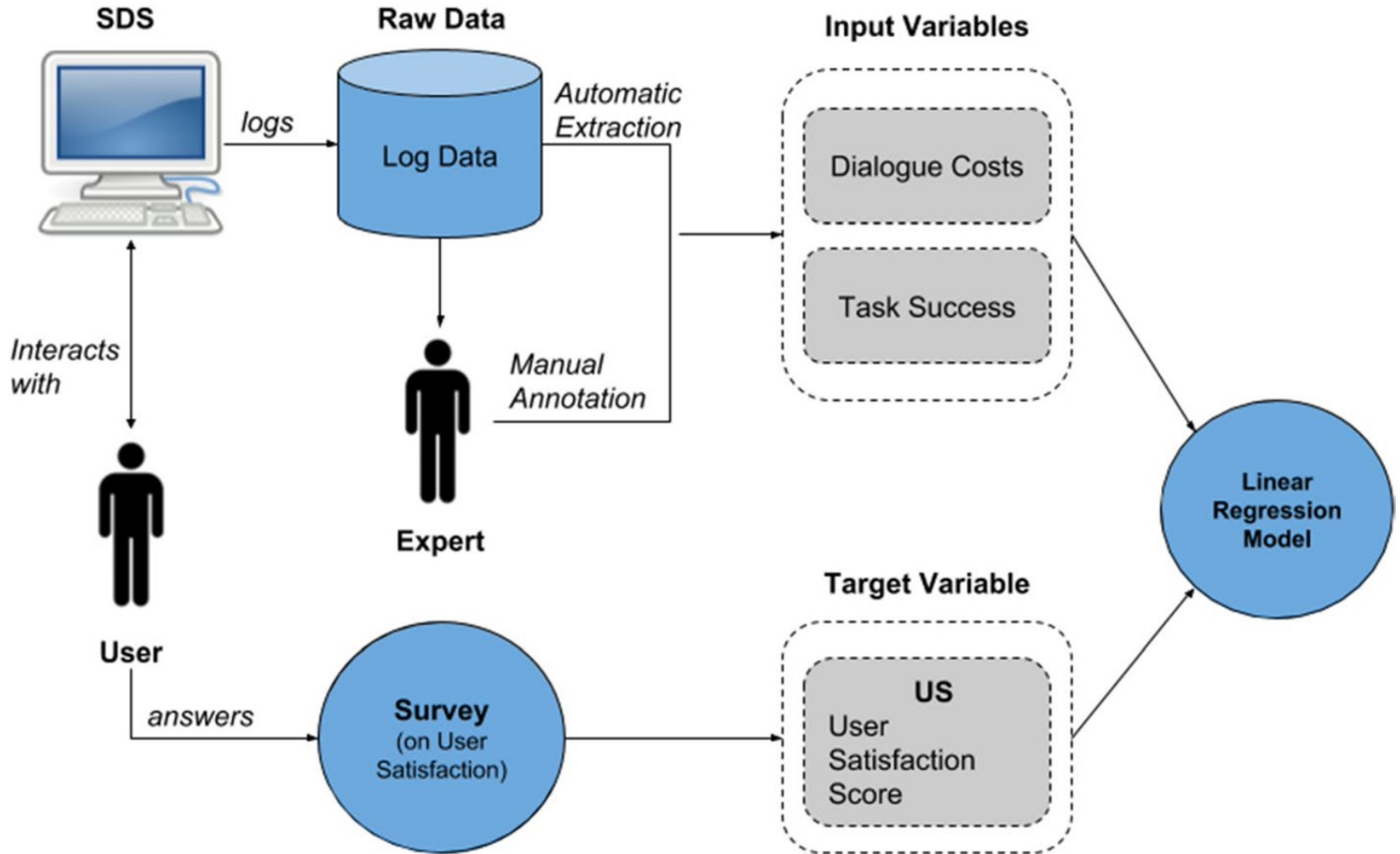
## Intent Detection (ID)

A Novel Bi-directional Interrelated Model for Joint Intent Detection and Slot Filling



# PARAdigm for Dialog System Evaluation

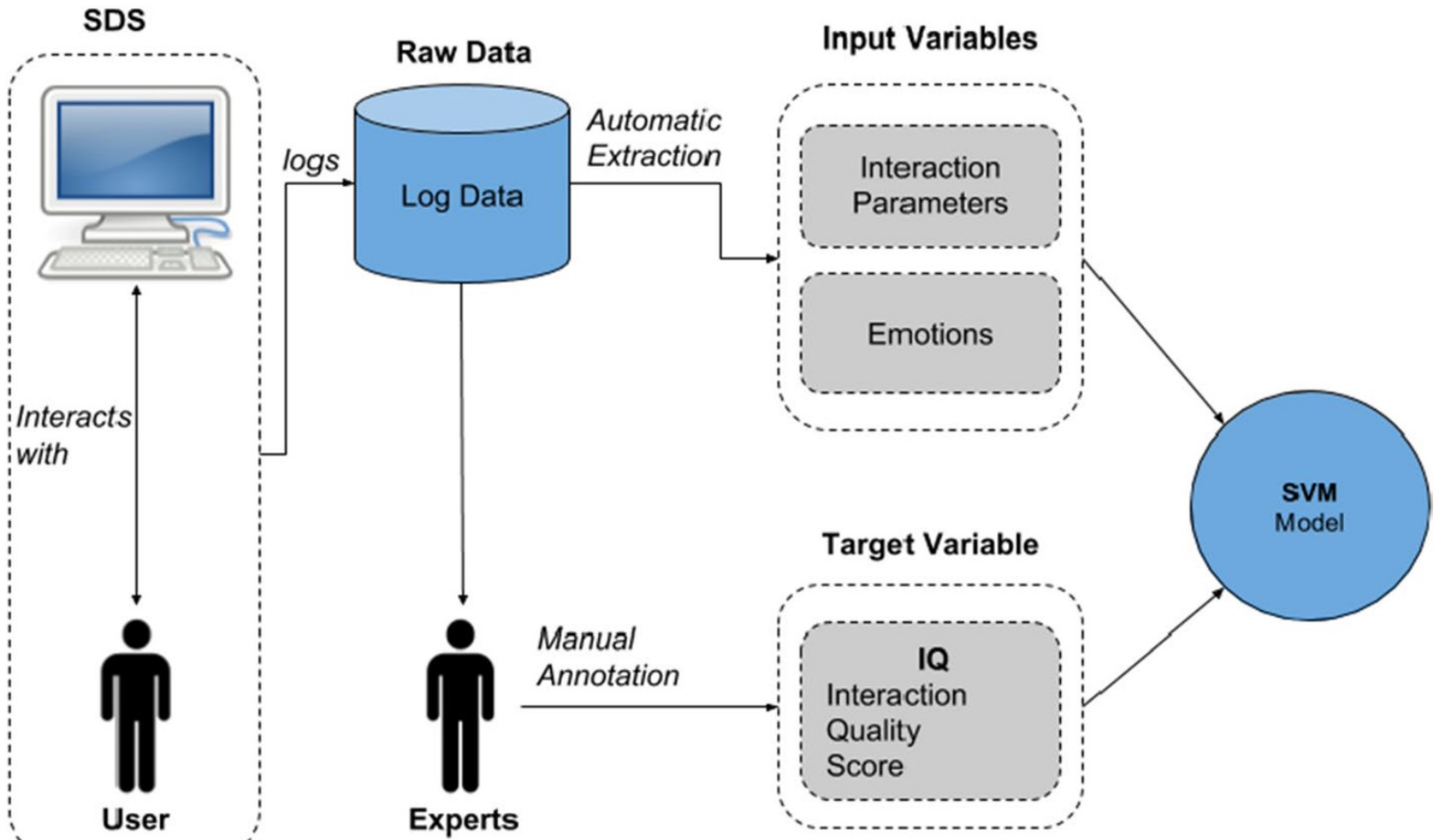
## PARADISE Framework (Walker et al. 1997)





# Interaction Quality procedure

(Schmitt and Ultes, 2015)



# Datasets for task-oriented dialogue systems

Name	Topics	# dialogues	Reference
DSTC1	Bus schedules	15,000	(Williams et al. <a href="#">2013</a> )
DSTC2	Restaurants	3000	(Henderson et al. <a href="#">2014</a> )
DSTC3	Tourist information	2265	(Henderson et al. <a href="#">2013a</a> )
DSTC4 & DSTC5	Tourist information	35	(Kim et al. <a href="#">2016</a> )
DSTC6	Restaurant reservation	–	(Perez et al. <a href="#">2017</a> )
DSTC7 (Flex Data)	Student guiding	500	(Gunasekara et al. <a href="#">2019</a> )
DSTC8 (MetaLWOz)	47 domains	37,884	(Lee et al. <a href="#">2019</a> )
DSTC8 (Schema-Guided)	20 domains	22,825	(Rastogi et al. <a href="#">2019</a> )
MultiWOZ	Tourist information	10,438	(Budzianowski et al. <a href="#">2018</a> )
Taskmaster-1	6 domains	13,215	(Byrne et al. <a href="#">2019</a> )
MultiDoGo	6 domains	86,698	(Peskov et al. <a href="#">2019</a> )

# Restaurants Dialogue Datasets

- MIT Restaurant Corpus
  - <https://groups.csail.mit.edu/sls/downloads/restaurant/>
- CamRest676  
(Cambridge restaurant dialogue domain dataset)
  - <https://www.repository.cam.ac.uk/handle/1810/260970>
- DSTC2 (Dialog State Tracking Challenge 2 & 3)
  - <http://camdial.org/~mh521/dstc/>

# CrossWOZ:

## A Large-Scale Chinese Cross-Domain Task-Oriented Dialogue Dataset

MultiWOZ	CrossWOZ
usr: I'm looking for a college type attraction. ... usr: I would like to visit <u>in town centre</u> please. ... usr: Can you find an Indian restaurant for me that is <b>also in the town centre</b> ?	usr: 你好，可以帮我推荐一个评分是4.5分以上的景点吗？ Hello, could you recommend an attraction with a rating of 4.5 or higher? sys: 天安门城楼，簋街小吃和 <u>北京欢乐谷</u> 都是很不错的地方呢。 Tiananmen, Gui Street, and <u>Beijing Happy Valley</u> are very nice places. usr: 我喜欢 <u>北京欢乐谷</u> ，你知道这个景点周边的酒店都是什么吗？ I like <u>Beijing Happy Valley</u> . What hotels are around this attraction? sys: 那可多了，有A酒店，B酒店，C酒店。 There are many, such as hotel A, hotel B, and hotel C. usr: 太好了，我正打算在 <u>景点附近</u> 找个酒店住宿呢，知道哪家评分是4分以上，提供叫醒服务的不？ Great! I am planning to find a hotel to stay <b>near the attraction</b> . Which one has a rating of 4 or higher and offers wake-up call service?
Schema	
usr: I want a hotel in <u>San Diego</u> and I want to check out on Thursday next week. ... usr: I need a one way flight to <b>go there</b> .	

# CrossWOZ:

## A Large-Scale Chinese Cross-Domain Task-Oriented Dialogue Dataset

Type	Single-domain goal					Multi-domain goal		
Dataset	DSTC2	WOZ 2.0	Frames	KVRET	M2M	MultiWOZ	Schema	<b>CrossWOZ</b>
Language	EN	EN	EN	EN	EN	EN	EN	<b>CN</b>
Speakers	H2M	H2H	H2H	H2H	M2M	H2H	M2M	<b>H2H</b>
# Domains	1	1	1	3	2	7	16	<b>5</b>
# Dialogues	1,612	600	1,369	2,425	1,500	8,438	16,142	<b>5,012</b>
# Turns	23,354	4,472	19,986	12,732	14,796	115,424	329,964	<b>84,692</b>
Avg. domains	1	1	1	1	1	1.80	1.84	<b>3.24</b>
Avg. turns	14.5	7.5	14.6	5.3	9.9	13.7	20.4	<b>16.9</b>
# Slots	8	4	61	13	14	25	214	<b>72</b>
# Values	212	99	3,871	1363	138	4,510	14,139	<b>7,871</b>

# Task-Oriented Dialogue

## Initial user state (=user goal)

id=1(Attraction): fee=free,  
name=?, nearby hotels=?

id=2(Hotel): **name=near (id=1)**,  
wake-up call=yes, rating=?

id=3(Taxi): **from=(id=1), to=(id=2)**,  
car type=? plate number=?

...

## Final user state

id=1 (Attraction): name=Tiananmen Square,  
fee=free, nearby hotels=[Beijing Capital  
Hotel, Guidu Hotel Beijing]

id=2 (Hotel): **name=Beijing Capital Hotel**,  
wake-up call=yes, rating=4.6

id=3 (Taxi): **from=Tiananmen Square**,  
**to=Beijing Capital Hotel**,  
car type=#CX, plate number=#CP



Source: Zhu, Qi, Kaili Huang, Zheng Zhang, Xiaoyan Zhu, and Minlie Huang. "Crosswoz: A large-scale chinese cross-domain task-oriented dialogue dataset." arXiv preprint arXiv:2002.11893 (2020).

# 任務型對話系統

## The Evaluation of Chinese Human-Computer Dialogue Technology, SMP2019-ECDT

- 自然語言理解  
Natural Language Understanding (NLU)
- 對話管理  
Dialog Management (DM)
- 自然語言生成  
Natural Language Generation (NLG)

# Python in Google Colab (Python101)

<https://colab.research.google.com/drive/1FEG6DnGvwfUbeo4zJ1zTunjMqf2RkCrT>

The screenshot shows a Google Colab notebook titled "python101.ipynb". The interface includes a top menu bar with "File", "Edit", "View", "Insert", "Runtime", "Tools", and "Help". A "Table of contents" sidebar on the left lists various topics, with "Question Answering and Dialogue Systems" highlighted. The main content area shows a code cell with the following text:

```
+ Code + Text
```

**Question Answering and Dialogue Systems**

- Question Answering (QA)
- BERT for Question Answering**

Source: Apoorv Nandan (2020), BERT (from HuggingFace Transformers) for Text Extraction, [https://keras.io/examples/nlp/text\\_extraction\\_with\\_bert/](https://keras.io/examples/nlp/text_extraction_with_bert/)

**Description:** Fine tune pretrained BERT from HuggingFace Transformers on SQuAD.

**Introduction**

This demonstration uses SQuAD (Stanford Question-Answering Dataset). In SQuAD, an input consists of a question, and a paragraph for context. The goal is to find the span of text in the paragraph that answers the question. We evaluate our performance on this data with the "Exact Match" metric, which measures the percentage of predictions that exactly match any one of the ground-truth answers.

We fine-tune a BERT model to perform this task as follows:

1. Feed the context and the question as inputs to BERT.
2. Take two vectors  $S$  and  $T$  with dimensions equal to that of hidden states in BERT.
3. Compute the probability of each token being the start and end of the answer span. The probability of a token being the start of the answer is given by a dot product between  $S$  and the representation of the token in the last layer of BERT, followed by a softmax over all tokens. The probability of a token being the end of the answer is computed similarly with the vector  $T$ .
4. Fine-tune BERT and learn  $S$  and  $T$  along the way.

**References:**

- [BERT](#)
- [SQuAD](#)

<https://tinyurl.com/aintpupython101>



# Python in Google Colab (Python101)

<https://colab.research.google.com/drive/1FEG6DnGvwfUbeo4zJ1zTunjMqf2RkCrT>

python101.ipynb ☆

File Edit View Insert Runtime Tools Help All changes saved

Comment Share Settings A

RAM [Progress Bar] Disk [Progress Bar] Editing

### Table of contents

- RandomizedSearchCV
- Sentiment Analysis
  - Sentiment Analysis - Unsupervised Lexical
  - Sentiment Analysis - Supervised Machine Learning
  - Sentiment Analysis - Supervised Deep Learning Models
  - Sentiment Analysis - Advanced Deep Learning
- Deep Learning and Universal Sentence-Embedding Models
  - Universal Sentence Encoder (USE)
  - Universal Sentence Encoder Multilingual (USEM)
- Question Answering and Dialogue Systems
  - Question Answering (QA)
    - BERT for Question Answering**
  - Dialogue Systems
    - Joint Intent Classification and Slot Filling with Transformers
- Data Visualization
- Section

```
+ Code + Text
```

Downloading: 100% [Progress Bar] 433/433 [00:29<00:00, 14.5B/s]

Downloading: 100% [Progress Bar] 536M/536M [00:29<00:00, 18.3MB/s]

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 384)]	0	
input_3 (InputLayer)	[(None, 384)]	0	
input_2 (InputLayer)	[(None, 384)]	0	
tf_bert_model (TFBertModel)	((None, 384, 768), ( 109482240		input_1[0][0]
start_logits (Dense)	(None, 384, 1)	768	tf_bert_model[0][0]
end_logits (Dense)	(None, 384, 1)	768	tf_bert_model[0][0]
flatten (Flatten)	(None, 384)	0	start_logits[0][0]
flatten_1 (Flatten)	(None, 384)	0	end_logits[0][0]
activation_7 (Activation)	(None, 384)	0	flatten[0][0]
activation_8 (Activation)	(None, 384)	0	flatten_1[0][0]

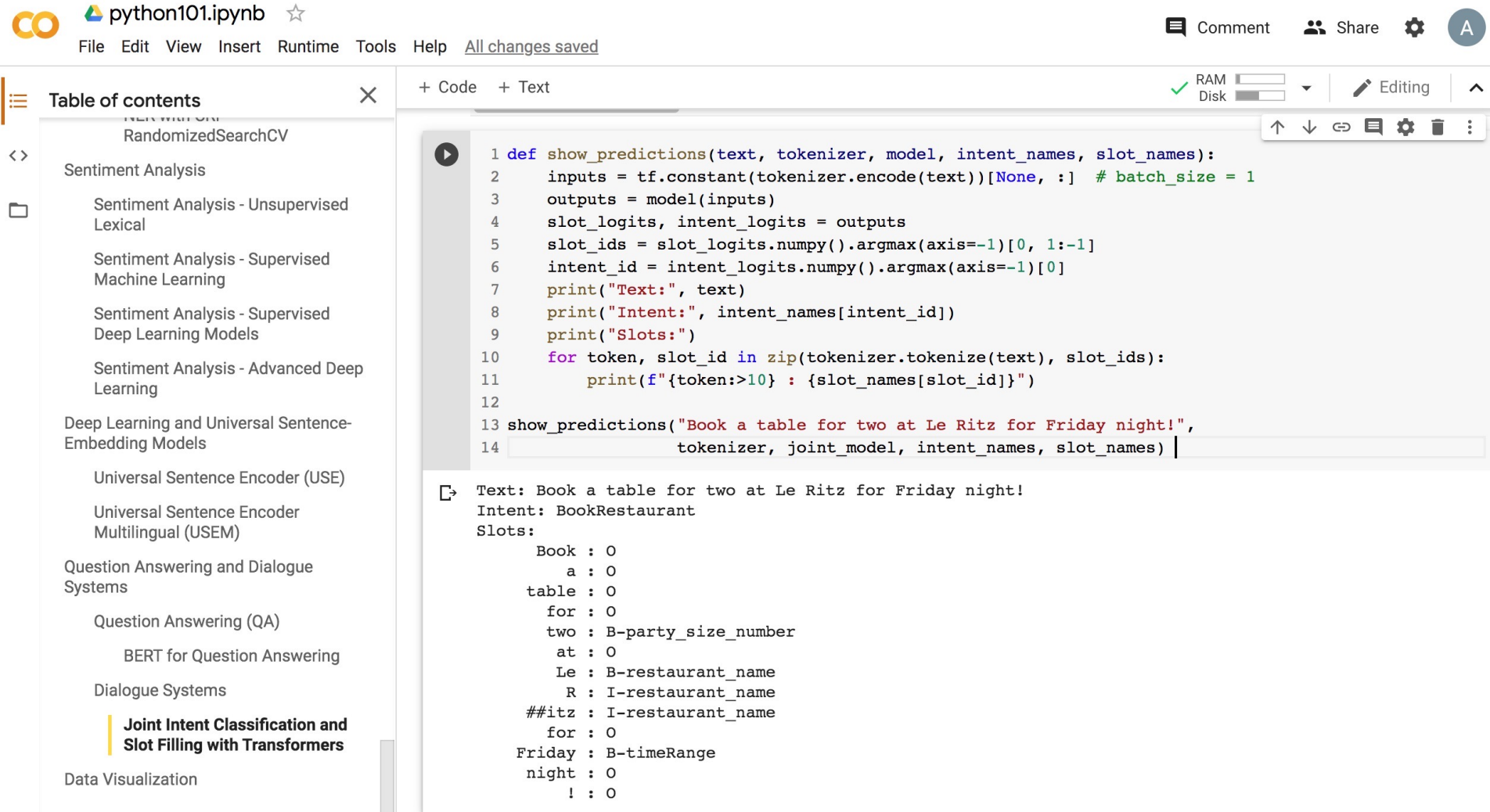
Total params: 109,483,776  
Trainable params: 109,483,776  
Non-trainable params: 0

CPU times: user 20.8 s, sys: 7.75 s, total: 28.5 s  
Wall time: 1min 42s

<https://tinyurl.com/aintpuppython101>

# Python in Google Colab (Python101)

<https://colab.research.google.com/drive/1FEG6DnGvwfUbeo4zJ1zTunjMqf2RkCrT>



python101.ipynb ☆

File Edit View Insert Runtime Tools Help [All changes saved](#)

Comment Share

RAM Disk

Editing

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- Question Answering and Dialogue Systems
  - Question Answering (QA)
    - BERT for Question Answering
  - Dialogue Systems
    - Joint Intent Classification and Slot Filling with Transformers**
- Data Visualization

```
1 def show_predictions(text, tokenizer, model, intent_names, slot_names):
2     inputs = tf.constant(tokenizer.encode(text))[None, :] # batch_size = 1
3     outputs = model(inputs)
4     slot_logits, intent_logits = outputs
5     slot_ids = slot_logits.numpy().argmax(axis=-1)[0, 1:-1]
6     intent_id = intent_logits.numpy().argmax(axis=-1)[0]
7     print("Text:", text)
8     print("Intent:", intent_names[intent_id])
9     print("Slots:")
10    for token, slot_id in zip(tokenizer.tokenize(text), slot_ids):
11        print(f"{token:>10} : {slot_names[slot_id]}")
12
13 show_predictions("Book a table for two at Le Ritz for Friday night!",
14                 tokenizer, joint_model, intent_names, slot_names)
```

Text: Book a table for two at Le Ritz for Friday night!  
Intent: BookRestaurant  
Slots:  
Book : 0  
a : 0  
table : 0  
for : 0  
two : B-party\_size\_number  
at : 0  
Le : B-restaurant\_name  
R : I-restaurant\_name  
##itz : I-restaurant\_name  
for : 0  
Friday : B-timeRange  
night : 0  
! : 0

<https://tinyurl.com/aintpuppython101>

# Summary

- Artificial Intelligence
- FinTech
- Conversational Commerce
- Task Oriented Dialogue System

# References

- Stuart Russell and Peter Norvig (2020), Artificial Intelligence: A Modern Approach, 4th Edition, Pearson
- Deriu, Jan, Alvaro Rodrigo, Arantxa Otegi, Guillermo Echegoyen, Sophie Rosset, Eneko Agirre, and Mark Cieliebak (2021). "Survey on evaluation methods for dialogue systems." Artificial Intelligence Review 54, no. 1 (2021): 755-810.
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# Q & A

# AI Task-Oriented Dialogue System for Conversational Commerce in FinTech

Host: Prof. Kuan-Yu Memphis Chen

Computer Science and Information Engineering, National Taiwan University of Science and Technology

Time: 14:00-15:00, May 10, 2021 (Monday)

Place: CSIE, NTUST

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