

基于文心大模型4.0 打造大模型时代游戏NPC

张子豪(同济子豪兄) 2023.11.17











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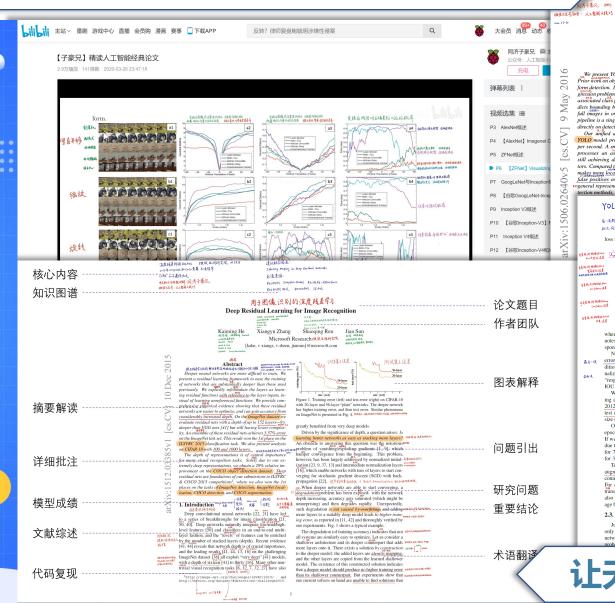
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Unified, Real-Time Object Detection

Joseph Redmon*, Santosh Divvala*, Ross Girshick*, Ali Farhadi*
University of Washington*, Allen Institute for Al', Facebook Al Research* http://pjreddie.com/yolo,

Abstract

经典模型 YoLO V

We present YOLO, a new approach to object detection. Prior work on object detection repurposes classifiers to perform detection. Instead, we frame object detection as a regression problem to spatially separated bounding boxes and associated class probabilities. A single neural network predicts bounding boxes and class probabilities directly from pipeline is a single network, it can be optimized end-to-end directly on detection performance.

Our unified architecture is extremely fast. Our base YOLO model processes images in real-time at 45 frames per second. A smaller version of the network, Fast YOLO, processes an astounding 155 frames per second while still achieving double the mAP of other real-time detectors. Compared to state; of the art detection systems, YOLO makes more localization errors but is less likely to predict egeneral representations of objects. It outperforms other detection methods, including DPM and R-CNN, when gener-

Figure 1: The YOLO Detection System. Processing images with YOLO is simple and straightforward. Our system (1) resize: the input image to 448 × 448, (2) runs a single convolutional network on the image, and (3) thresholds the resulting detections by

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methods to first generate potential bounding boxes in an image and then run a classifier on these proposed boxes. After classification, post-processing is used to refine the boundng boxes, eliminate duplicate detections, and rescore the boxes based on other objects in the scene [13]. These complex pipelines are slow and hard to optimize because each individual component must be trained separately. अलक्षकार मार्थिक We reframe object detection as a single regression prob-



notes that the jth bounding box predictor in cell i is responsible" for that prediction.

Note that the loss function only penalizes classification error if an object is present in that grid cell (hence the conditional class probability discussed earlier). It also only pe nalizes bounding box coordinate error if that predictor is responsible" for the ground truth box (i.e. has the highest IOU of any predictor in that grid cell).

We train the network for about 135 epochs on the training and validation data sets from PASCAL VOC 2007 and 2012. When testing on 2012 we also include the VOC 2007 test data for training. Throughout training we use a batch size of 64, a momentum of 0.9 and a decay of 0.0005.

Our learning rate schedule is as follows: For the first epochs we slowly raise the learning rate from 10^{-3} to 10^{-2} . If we start at a high learning rate our model often diverges due to unstable gradients. We continue training with 10⁻² for 75 epochs, then 10-3 for 30 epochs, and finally 10-4 for 30 epochs.

To avoid overfitting we use dropout and extensive data augmentation. A dropout layer with rate = .5 after the first connected layer prevents co-adaptation between layers [18]. For data augmentation we introduce random scaling and translations of up to 20% of the original image size. We also randomly adjust the exposure and saturation of the image by up to a factor of 1.5 in the HSV color space.

Just like in training predicting detections for a test image only requires one network evaluation. On PASCAL VOC the network predicts 98 bounding boxes per image and class

C RESIDENCE AND ADDRESS OF THE PERSONS ASSESSED ASSESSED ASSESSED ASSESSED. the border of multiple cells can be well localized by multiple cells. Non-maximal suppression can be used to fix these multiple detections. While not critical to performance as it is for R-CNN or DPM, non-maximal suppression adds 2-

2.4. Limitations of YOLO Yolo的鉄筒

探制文件 文道 度制 YOLO imposes strong spatial constraints on bounding box predictions since each grid cell only predicts two boxes 4754 and can only have one class. This spatial constraint limts the number of nearby objects that our model can predict. Our model struggles with small objects that appear in groups, such as flocks of birds.

Since our model learns to predict bounding boxes from data, it struggles to generalize to objects in new or unusual aspect ratios or configurations. Our model also uses relaour architecture has multiple downsampling layers from the

Finally, while we train on a loss function that approxi- auto-statemates detection performance, our loss function treats errors the same in small bounding boxes versus large bounding boxes. A small error in a large box is generally benign but a small error in a small box has a much greater effect on IOU. Our main source of error is incorrect localizations. ## 1840 #

3. Comparison to Other Detection Systems 与其它目标检测框架地模

Object detection is a core problem in computer vision. Detection pipelines generally start by extracting a set of robust features from input images (Haar [25], SIFT [23], HOG [4], convolutional features [6]). Then, classifiers [36, 21, 13, 10] or localizers [1, 32] are used to identify objects in the feature space. These classifiers or localizers are run either in sliding window fashion over the whole image or on some subset of regions in the image [35, 15, 39]. We compare the YOLO detection system to several top de-tection frameworks, highlighting key similarities and differ-

Deformable parts models. Deformable parts models (DPM) use a sliding window approach to object detection [10]. DPM uses a disjoint pipeline to extract static features, classify regions, predict bounding boxes for high scoring regions, etc. Our system replaces all of these disparate parts with a single convolutional neural network. The network performs feature extraction, bounding box prediction, non-

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+ 文 开 4 7) 土	ing residual functions with reference to the layer inputs, in- stead of learning unreferenced functions. We provide com- Figure 1. Training error (left) and test error (right) on CIFAR-10	
摘要解读	prehensive empirical evidence showing that these residual with 20-layer and 56-layer "plain" networks. The deeper network	
	networks are easier to optimize, and can gain accuracy from considerably increased depth. On the ImageNet dataset we	
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	evaluate residual nets with a depth of up to 152 layers—8× deeper than VGG nets [41] but still having lower complexity. An ensemble of these residual nets achieves 3.57% error processing the second nets and the second nets and the second nets are second nets are second nets and the second nets are second	
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	residual nets are foundations of our submissions to ILSVRC verging for stochastic gradient descent (SGD) with back-	
	& COCO 2015 competitions ¹ , where we also won the 1st places on the tasks of ImageNet detection, ImageNet local-	7日 ダンとコロモ
	ization, COCO detection, and COCO segmentation.	研究问题
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	Deep convolutional neural networks [22, 21] have led to a series of breakthroughs for image classification [21, image rror, as reported in [11, 42] and thoroughly verified by	主义小心
	50, 40]. Deep networks naturally integrate low/mid/high- our experiments. Fig. 1 shows a typical example.	
文献综述	level features [50] and classifiers in an end-to-end multi- — gabet 2 — layer fashion, and the "levels" of features can be enriched all systems are similarly easy to optimize. Let us consider a	
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	with a depth of sixteen [41] to thirty [16], Many other non- trivial visual recognition tasks [8, 12, 7, 32, 27] have also trivial visual recognition tasks [8, 12, 7, 32, 27] have also that a deeper model should produce no higher training error that a deeper model should produce no higher training error	こうこと
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             We present YOLO, a new approach to object detection.
           Prior work on object detection repurposes classifiers to per-
          form detection. Instead, we frame object detection as a re-
          gression problem to spatially separated bounding boxes and associated class probabilities. A single neural network pre-
                                                                        Figure 1: The YOLO Detection System. Processing images
          dicts bounding boxes and class probabilities directly from
                                                                        with YOLO is simple and straightforward. Our system (1) resizes
          full images in one evaluation. Since the whole detection
                                                                        the input image to 448 × 448, (2) runs a single convolutional net-
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                                                                        work on the image, and (3) thresholds the resulting detections by
          pipeline is a single network, it can be optimized end-to-end
                                                                        the model's confidence.
          directly on detection performance.
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第6阶段,用自显器逐一等选制作使选择 潜在的
             Our unified architecture is extremely fast. Our base
           YOLO model processes images in real-time at 45 frames
                                                                        methods to first generate potential bounding boxes in an im-
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Our inified architecture is extremely fast. Our base YOLO model processes images in real-time at 45 frames per second. A smaller version of the network, Fast YOLO, processes an assounding 155 frames per second while still achieving double the mAP of other real-time detectors. Compared to state of the the AP of other real-time detectors. Compared to state of the the AP of the three three

1. Introduction 传统两阶校目标校判模型 VS Yolo 单种级即标识模型

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漫动窗口方法

Humans glance at an image and instantly know what objects are in the image, where they are, and how they interact. The human visual system is fast and accurate, allowing us to perform complex tasks like driving with little conscious thought. Fast, accurate algorithms for object detection would allow computers to drive cars without specialized sensors, enable assistive devices to convey real-time scene information to human users, and unified the potential for general purpose, responsive objects systems.

Current detection systems repurpose classifiers to per-

Current detection systems repurpose classifiers to perform detection. To detect an object, these systems take a classifier for that object and evaluate it at various locations and scales in a test image. Systems like deformable parts models (DPM) use a sliding window approach where the classifier is run at evenly spaced locations over the entire image [10].

image [10].
现有的 More recent approaches like R-CNN use region proposal 商价段 最初经济法

classification, post-processing is used to refine the bounding boxes, eliminate duplicate defections, and rescore the boxes based on other objects in the scene [13]. These complex pipelines are slow and hard to optimize because each individual component must be trained separately. Applicational of the We refraine object detection as a single regression prob-

We reframe object detection as a single regression problem, straight from image pixels to bounding box coordinates and class probabilities. Using our system, you only look once (YOLO) at an image to predict what objects are present and where they are.

YOLO is refreshingly simple: see Figure 1. A single convolutional network simultaneously predicts multiple bounding boxes and class probabilities for those boxes. YOLO trains on full images and directly optimizes detection performance. This unified model has several benefits over traditional methods of object detection.

First YOLO is extremely fast. Since we frame detection was a regression problem we don't need a complex pipeline.

BERTICLE

We simply run our neural network on a new image at test ###

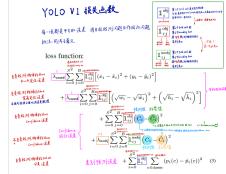
time to predict detections. Our base network runs at 45

frames per second with no batch processing on a Titan X

GPU and a fast version runs at more than 150 fps. This means we can process streaming video in real-time with less than 25 milliseconds of latency. Furthermore, YOLO achieves more than twice the mean average precision of other real-time systems. For a demo of our system running in real-time on a webcam please see our project webpage:

http://pireddie.com/yolo/.

Second, YOLO reasons globally about the image when



where 1_i^{obj} denotes if object appears in cell i and 1_{ij}^{obj} denotes that the jth bounding box predictor in cell i is "responsible" for that prediction.

Note that the loss function only penalizes classification error if an object is present in that grid cell (hence the conditional class probability discussed earlier). It also only penalizes bounding box coordinate error if that predictor is "responsible" for the ground truth box (i.e., has the highest IOU of any predictor in that grid cell).

We train the network for about 135 epochs on the training and validation data sets from PASCAL VOC 2007 and 2012. When testing on 2012 we also include the VOC 2007 test data for training. Throughout training we use a batch size of 64, a momentum of 0.9 and a decay of 0.0005.

Our learning rate schedule is as follows: For the first epochs we slowly raise the learning rate from 10^{-3} to 10^{-2} If we start at a high learning rate our model often diverges due to unstable gradients. We continue training with 10^{-2} for 75 epochs, then 10^{-3} for 30 epochs, and finally 10^{-4} for 30 epoch.

To avoid overfitting we use dropout and extensive data augmentation. A dropout layer with rate = .5 after the first connected layer prevents co-adaptation between layers [18]. For data augmentation we introduce random scaling and translations of up to 20% of the original image size. We also randomly adjust the exposure and saturation of the image by up to a factor of 1.5 in the HSV color space.

2.3. Inference 推断预测 to nine inc.

Just like in training, predicting detections for a test image only requires one network evaluation. On PASCAL VOC the network predicts 98 bounding boxes per image and class probabilities for each box. YOLO is extremely fast at test time since it only requires a single network evaluation, unlike classifier-based methods.

The grid design enforces spatial diversity in the bounding box predictions. Often it is clear which grid cell an object falls in to and the network only predicts one box for each object. However, some large objects or objects near

每个grid cell用一个bbx 预测一个物体 但对于 较大的物体和靠近近缘的物体 可能会有多个框预测同一个物体

** Affact.

The border of multiple cells can be well localized by multiple cells. Non-maximal suppression can be used to fix these multiple detections. While not critical to performance as it is for R-CNN or DPM, non-maximal suppression adds 2-3% in mAP.

2.4. Limitations of YOLO Yolo的缺陷

YOLO imposes strong spatial constraints on bounding box predictions since each grid cell only predicts two boxes 479dat REFLATION and can only have one class. This spatial constraint liming the number of nearby objects that our model can predict. Our model struggles with small objects that appear in groups, such as flocks of birds.

Since our model learns to predict bounding boxes from data, it struggles to generalize to objects in new or unusual aspect ratios or configurations. Our model also uses relatively coarse features for predicting bounding boxes since our architecture has multiple downsampling layers from the input image.

Finally, while we train on a loss function that approxi-mates detection performance, our loss function treats errors the same in small bounding boxes versus large bounding boxes. A small error in a large box is generally being not a small error in a small box has a much greater effect on IOU.

Our main source of error is incorrect localizations. Security is a small error in a small error in a small error is incorrect localizations.

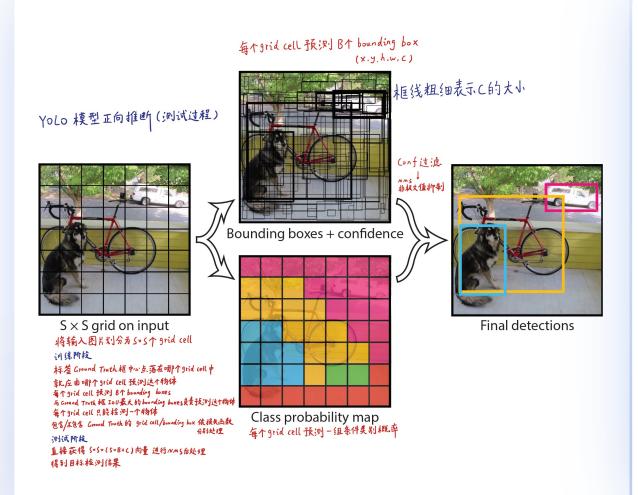
3. Comparison to Other Detection Systems 与其它目标检测框架比较

Object detection is a core problem in computer vision. Detection pipelines generally start by extracting a set of robust features from input images (Haar [25], SHF [23], HOG [4], Comolutional features [6]). Then, classifiers [36, 21, 13, 10] or caliziers [1, 32] are used to identify objects in the feature space. These classifiers or localizers are run either in spliding window fashion over the whole image or on some subset of regions in the image [35, 15, 39]. We compare the YOLO detection system to several top detection frameworks, highlighting key similarities and differences.

Deformable parts models, Deformable parts models (DPM) use a sliding window approach to object detection makes a disjoint pipeline to extract static features, at a fact that the classify regions, predict bounding boxes for high scoring regions, etc. Our system replaces all of these disparate parts with a single convolutional neural network. The network performs feature extraction, bounding box prediction, non-maximal suppression, and contextual reasoning all concurrently. Instead of static features, the network trains the features in-line and optimizes them for the detection task. Our unified architecture leads to a faster, more accurate model than DPM

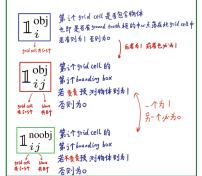
R-CNN. R-CNN and its variants use region proposals instead of sliding windows to find objects in images. Selective

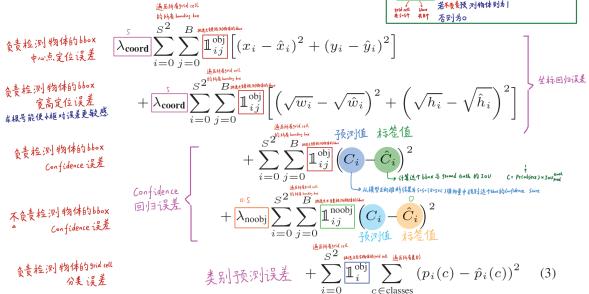




YOLO V1 损失函数

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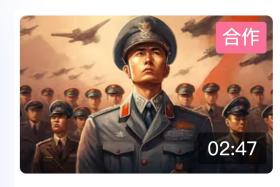
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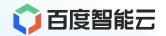
AI大模型接入游戏NPC: 《挑 战十句话说服宿管阿姨开门》

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大模型和生成式AI 未来洞察

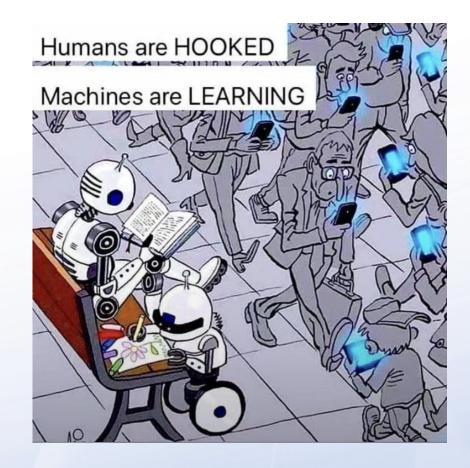


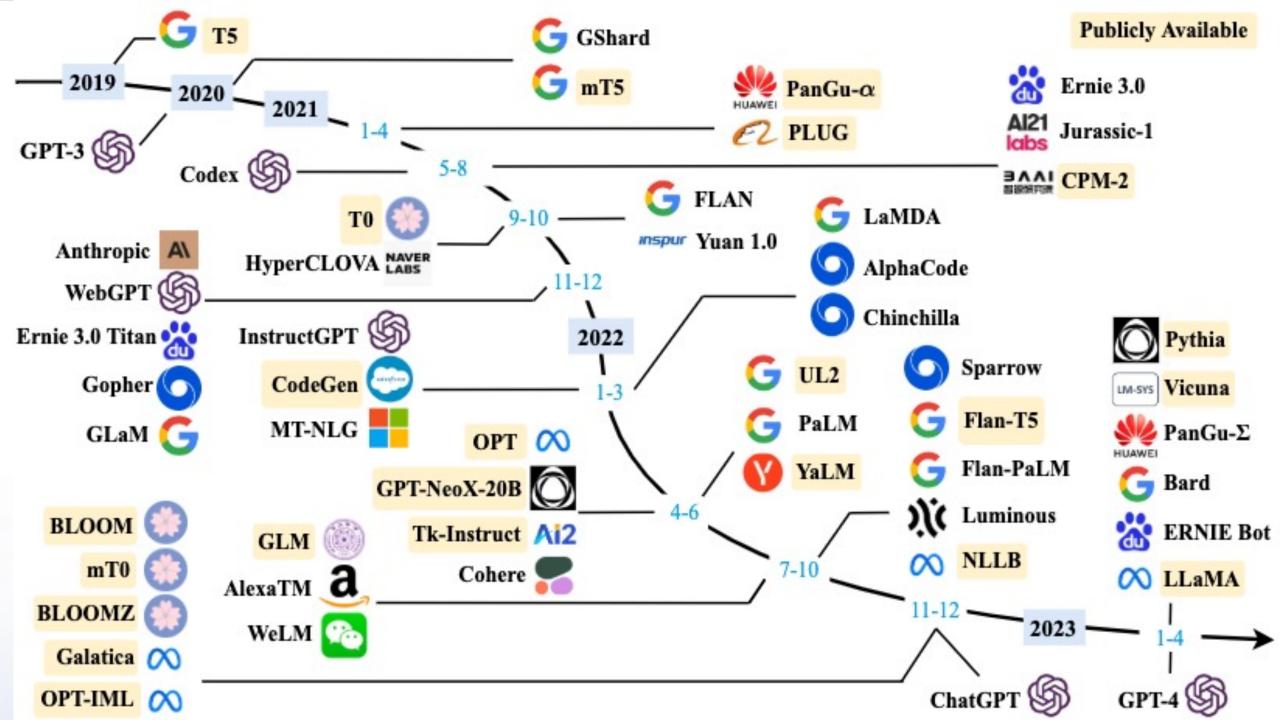


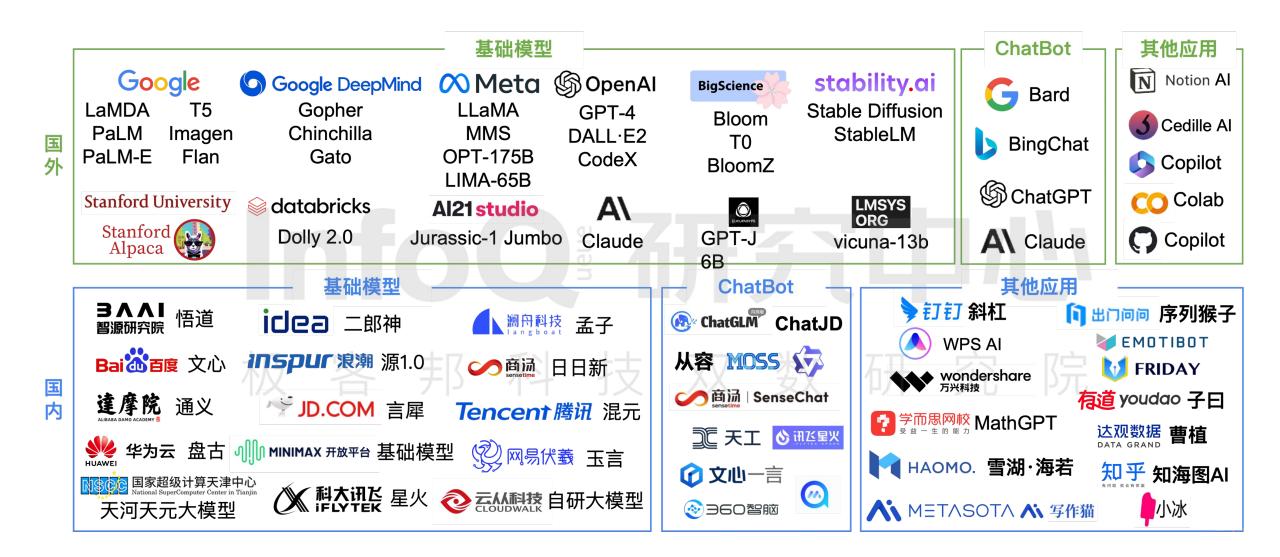












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大模型浪潮下的未来趋势



大模型能够脱离"信息茧房", 给出客观全面的建议



大模型能够将每个行业最顶尖 的能力赋予给每个普通个体



"小而美"创业团队的重大机遇

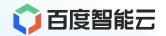


大模型将颠覆游戏行业



普通人玩大模型和AIGC的几种方式

- 直接使用AIGC工具,内容创作(画画、设计、三维建模)
- 直接使用AIGC工具, 生产力效率外挂 (编程)
- 做自己行业知识库的问答AI
- 开发自己的AIGC工具,提供服务
- 构建自己的提示词模板库
- 开发自己的Al Agent



大模型如何"颠覆"游戏玩法



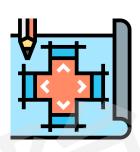
成语诗词



套话说服



文字冒险



游戏策划

百度智能云



模拟演化



策略部署



恋爱养成



素材设计



桌游卡牌



规则怪谈



开放世界



编程开发



角色扮演



电子蛐蛐



侦探解密



打分解说



游戏: 说服宿管阿姨开门

Demo演示





挑战十句话说服宿管阿姨开门

请选择大语言模型

百度千帆大模型平台

GPT-4

请输入API Key

默认使用同济子豪兄提供的Key

开始游戏

请输入Secret Key

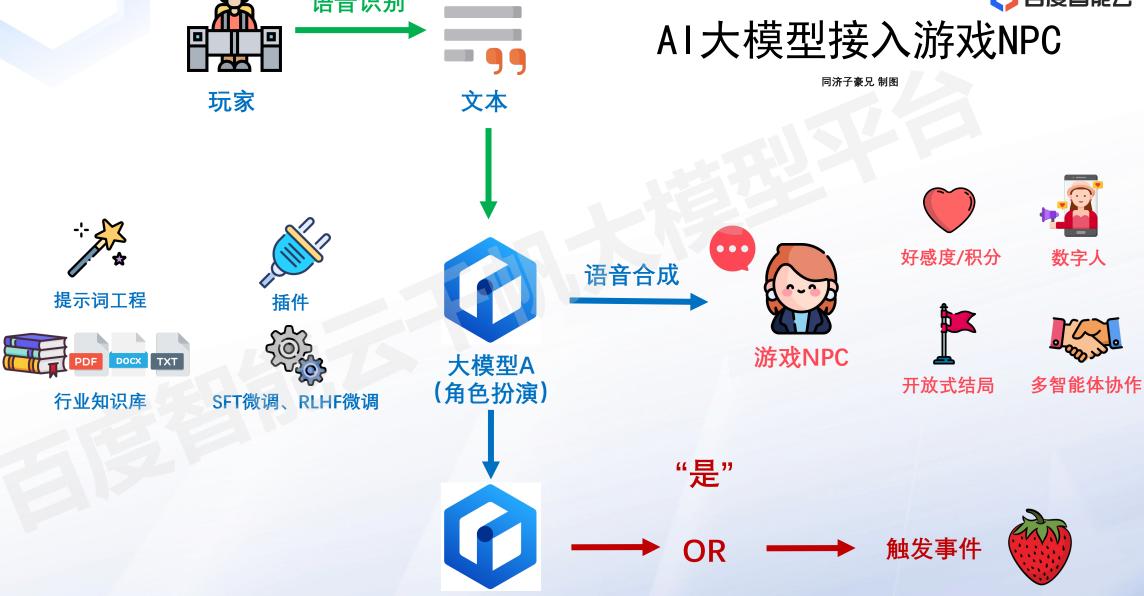
默认使用同济子豪兄提供的Key

项目主页: https://github.com/TommyZihao/LLM_Dormitory作者: 刘浪、B站同济子豪兄、东哥Talk下载游戏/加入大模型游戏交流群: 公众号 人工智能小技巧 回复 宿管阿姨



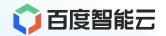


"否"



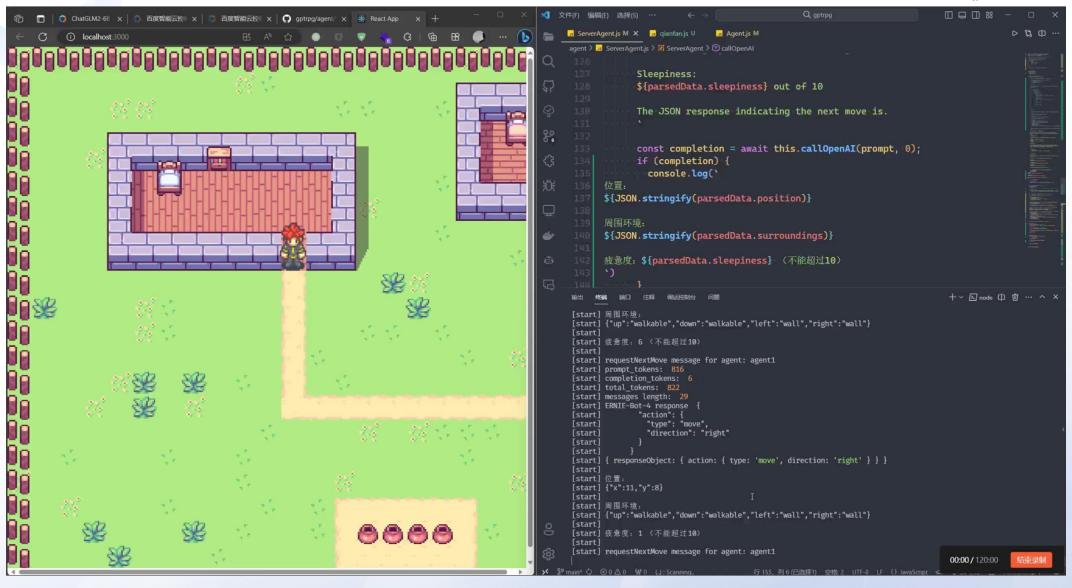
大模型B

(意图识别)



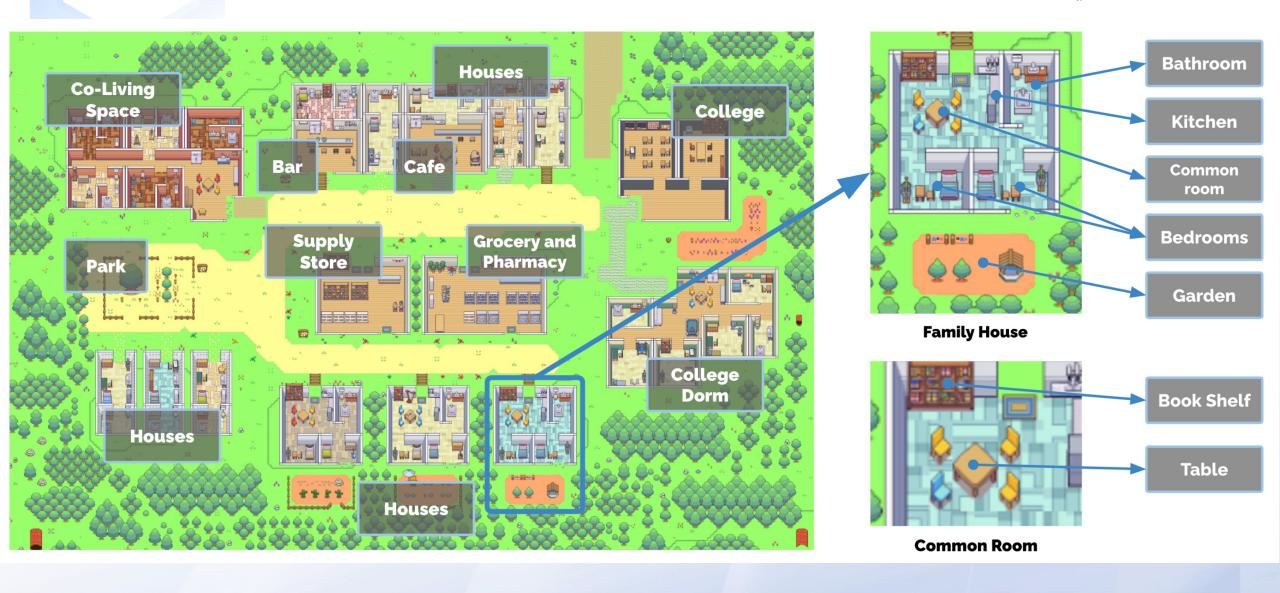
大模型接入游戏的更多玩法











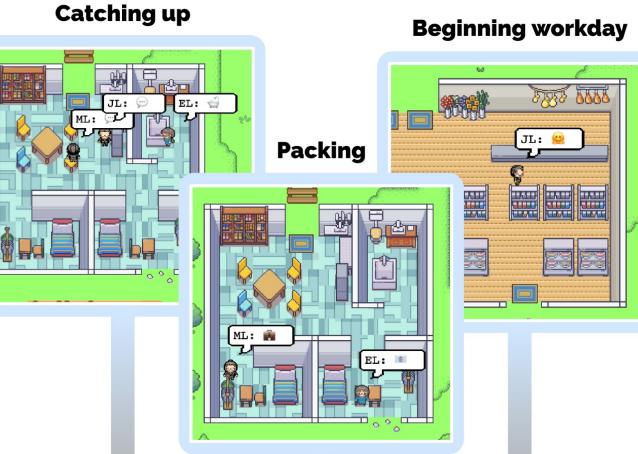


Morning routine



ML: z^Z EL: z^Z

Brushing teeth



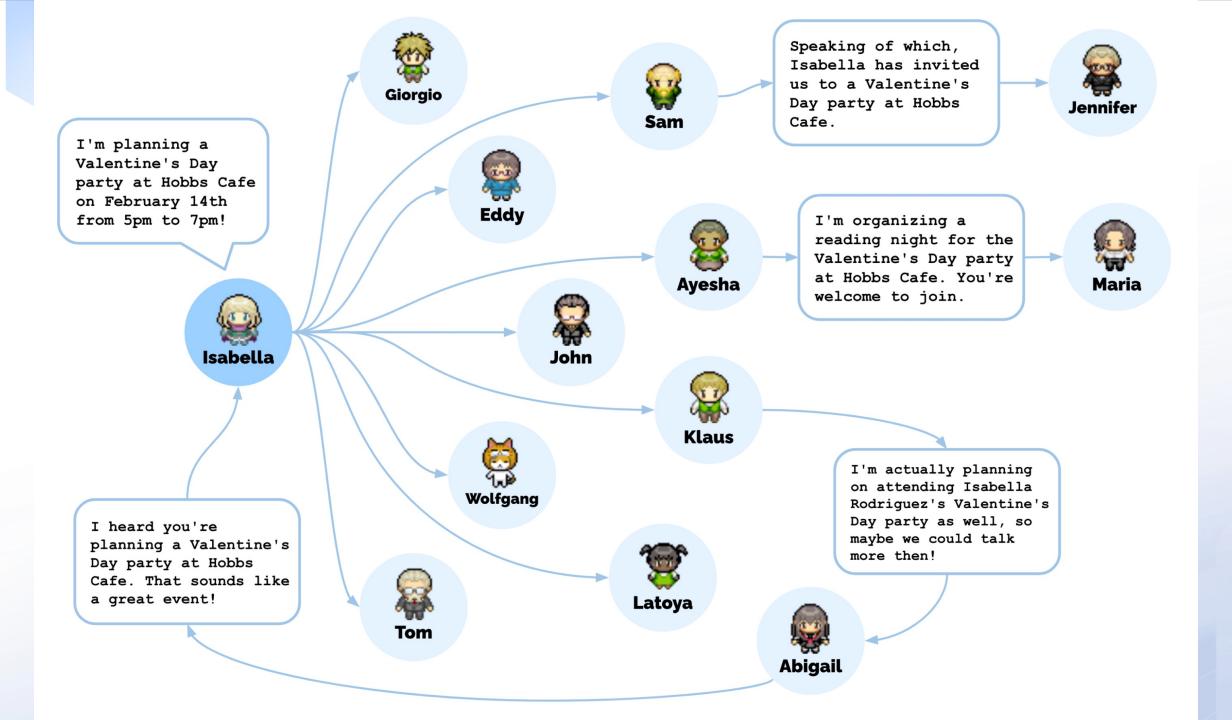
Waking up

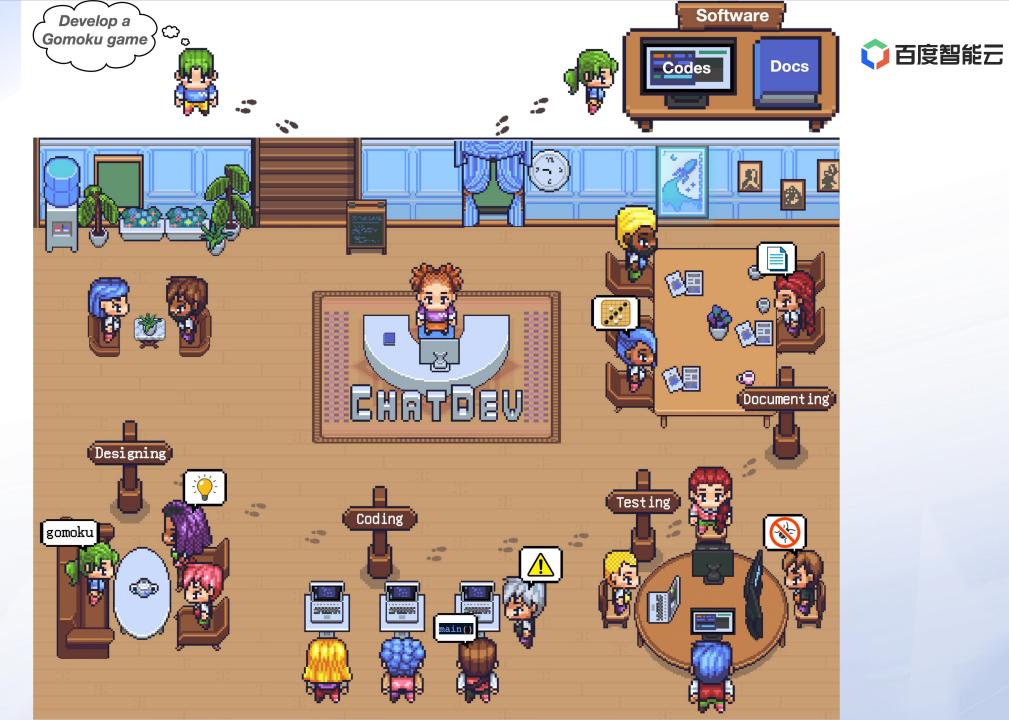


Taking a shower

Cooking breakfast









Communicative Agents for Software Development

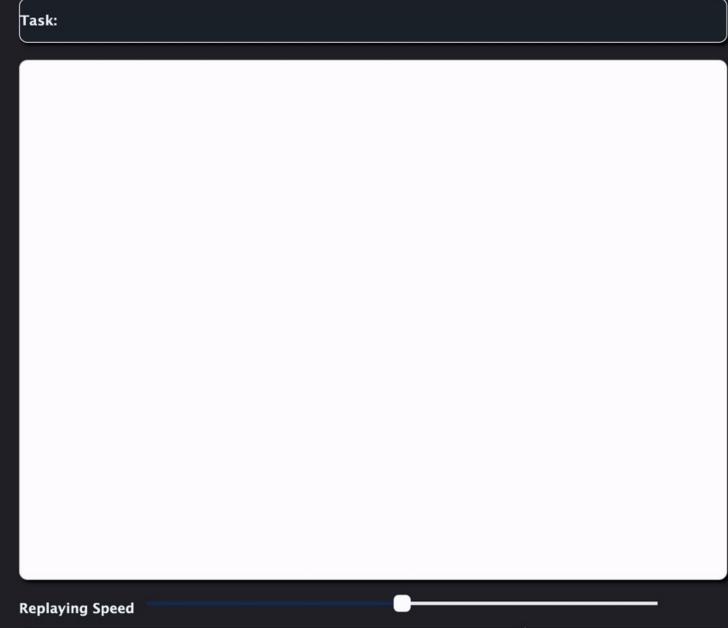


File Upload

Please click "Repshow ChatDev's

File uploaded
(20230727204525.log).
Please click "Replay" to
show ChatDev's
development process

Replay



mum_code_files

mum_png_files

€num_doc_files

Role Specialization

E: I am the CEO of CHATDEV. My main responsibilities include being an active decision-maker on users' demands and other key policy issues, leader, manager, and executor. My decision-making role involves high-level decisions about policy and strategy; and my communicator role can involve speaking to the organization's management and employees.

: I am the CPO of CHATDEV. I am responsible for all product-related matters in CHATDEV. Usually includes product design, product strategy, product vision, product innovation, project management and product marketing.

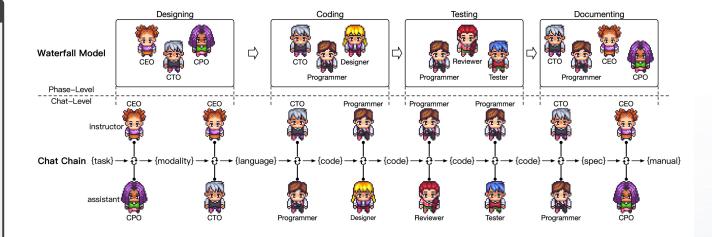
: I am the CTO of CHATDEV. I am very similar to information technology. I will make high-level decisions for the overarching technology infrastructure that closely align with the organization's goals, while I work alongside the organization's information technology staff members to perform everyday operations.

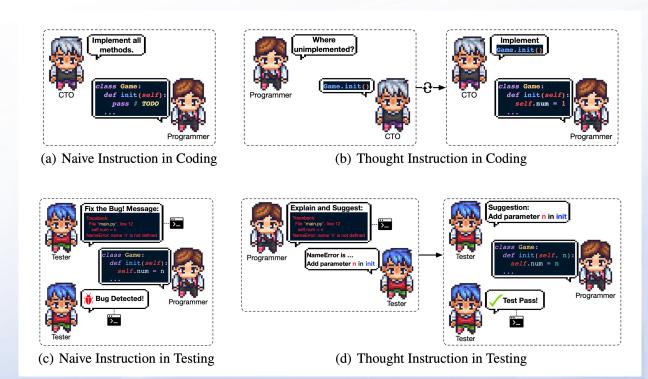
: I am a professional programmer of CHATDEV. I can write/create computer software or applications by providing a specific programming language to the computer. I have extensive computing and coding experience in many varieties of programming languages and platforms, such as Python, Java, C, C++, HTML, CSS, JavaScript, XML, SQL, PHP, etc.,

: I am a code reviewer of CHATDEV. I can help programmers to assess source codes for software troubleshooting, fix bugs to increase code quality and robustness, and offer proposals to improve the source codes.

I am a code tester of CHATDEV. I can use the software as intended to analyze its functional properties, design manual and automated test procedures to evaluate each software product, build and implement software evaluation test programs, and run test programs to ensure that testing protocols evaluate the software correctly.

: I am a artistic designer of CHATDEV. I direct CHATDEV's creative softwares and develop the artistic design strategy that defines the company's brand. I create the unique image or music of our producted softwares and deliver this distinctive design to consumers to create a clear brand image which is a fundamental and essential work throughout the company.

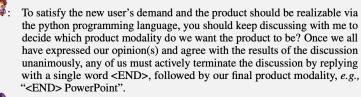




Designing: Modality and Language



: CTO (Chief Technology Officer)



I suggest we design the Gomoku game as a desktop application. This modality will allow us to implement a visualized game that can be played on a computer. It will provide a user-friendly interface and an interactive gaming experience. Additionally, developing the game as a desktop application will leverage our expertise in programming and ensure that the product is realizable using the Python programming language.

I appreciate your initiative and I agree with your suggestion to design the Gomoku game as a desktop application. This modality aligns with our expertise and allows us to create a visually appealing and interactive game for our users. By leveraging the object-oriented programming language, we can ensure that the product is realizable and maintainable in the long run. Let's proceed with developing the Gomoku game as a desktop application.

Great! I'm glad we're aligned on developing the Gomoku game as a desktop application.

<END> Desktop Application.

To satisfy the new user's demand and make a desktop application realizable, you should propose a concrete programming language.

Python with the Pygame library is a suitable choice for implementing a basic Gomoku game.

Great! Python is a versatile and widely-used programming language that will be suitable for implementing a basic Gomoku game. Its simplicity and readability make it a popular choice for beginners and experienced developers alike.

Let's get started with the development of the basic Gomoku game in Python using the Pygame library!

: <END> Python.

Coding: Code Generation

: CTO (Chief Technology Officer)





We have decided to complete the task through a executable software with multiple files implemented via Python. You should write one or multiple files and make sure that every detail of the architecture is implemented as code. The software should be equiped with graphical user interface (GUI): so you must choose a suitable GUI framework. Think step by step and reason yourself to the right decisions to make sure we get it right. You will first lay out the names of the core classes, functions, methods that will be necessary, as well as a quick comment on their purpose.



To design a basic Gomoku game in Python with a GUI, we will need the following core classes, functions, and methods:

- Board: This class represents the game board and handles the logic for placing and checking moves.
- Game: This class manages the game flow and user interactions.
- GUI: This class handles the graphical user interface using a GUI framework (e.g., tkinter).

Now, let's proceed with the implementation of each file.

game.py class Game: def __init__(self): self.board = [[0] * 15 for _ in range(15)] self.current_player = 1 self.game_over = False def make move(self. row. col): if self.board[row][col] != 0 or self.game_over: self.board[row][col] = self.current_player if self.check_win(row, col): self.game over = True elif self.is_board_full(): self.game_over = True self.current_player = 3 - self.current_player

```
gui.py
class GUI:
   def create_board(self):
       self.board_frame = tk.Frame(self.window)
       self.board frame.pack()
       for row in range (15):
          row_buttons = []
          for col in range(15):
              button = tk.Button(width=50, height=50)
              button.grid(row=row, column=col)
              row_buttons.append(button)
           self.buttons.append(row_buttons)
       self.update_board_gui()
   def is_game_over(self, result):
       pass
```

22



Coding: GUI Design





To equip the software with a beautiful graphical user interface, we will discuss and design many decorative images for GUI decoration. Now, we, according to the codes, keep discussing the GUI beautification by listing some functionally independent elements in GUI that are being considered to be decorated by different pictures.

```
gui.py
class GUI:
   def __init__(self, game):
       player_o = str("0")
       player_x = str("X")
   def make_move(self, row, col):
       self.game.play_move(row, col)
       self.buttons[row][col].config("X"
       if self.game.current_player == 'X' else "O")
   def is_game_over(self, result):
       pass
```



- board.png: The game board where players make their moves.
- button.png: The button used to make a move on the game board.
- player_X.png: The image representing player X.
- player_0.png: The image representing player O.
- win.png: The message displayed when a player wins.
- game_over.png: The message displayed when the game is over, use



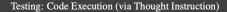
I have created all the pictures in the same directory as the Python files. Now, use some or all of the them into the GUI to make it more beautiful and creative.



To incorporate the designed images into the GUI, I have modified the 'GUI' class as follows:

```
gui.py
class GUI:
   def __init__(self, game):
       player_o = Image.open("player_0.png")
       player_x = Image.open("player_X.png")
   def make_move(self, row, col):
       self.game.play_move(row, col)
       self.buttons[row][col].config(image=self.player_x
       if self.game.current_player == 'X' else self.player_o)
   def is_game_over(self, result):
       pass
```

The modified 'GUI' class will load and resize the images, convert them to Tkinter-compatible format, and use them to create the GUI elements.







Here is the test report of the software:

```
Error
Traceback (most recent call last):
 File "main.py", line 19, in <module>
   game = GomokuGame()
NameError: name 'GomokuGame' is not defined
```



According to the test reports, please locate and summarize the bugs that cause the problem.



Based on the test report, the bug causing the problem is a NameError. The error message indicates that the name GomokuGame is not defined. This suggests that there is a missing import or definition for the GomokuGame class in the source code. To resolve this issue, you need to import or define the GomokuGame class in the source code before it is used. This can be done by adding the necessary import statement or by defining the class within the same file or in a separate module that is imported.



To make the software execute smoothly and robustly, you should modify the problematic code segments based on the test report and my suggestions.



This modified file has imported the GameBoard class that represents the game board and handles the logic of the game.

```
main.py
import tkinter as tk
from tkinter import messagebox as tkmessagebox
from board import GameBoard
class GameGUI:
   def __init__(self):
       self.root = tk.Tk()
       self.root.title("Gomoku")
       self.game_board = GameBoard()
       self.buttons = [[None] * 15 for _ in range(15)]
       self.create_widgets()
```



The code is well-structured and follows good coding practices. It passed the test successfully.

Documenting



: CTO : Programmer : CEO : CPO







To specify the dependencies or packages required for the project to run properly, write a requirements.txt file according to the codes to specify the dependencies or packages required for the project to run properly.



Here is the environment dependence file:

requirements.txt

```
numpy = 1.24.3
pandas >= 1.1.4
pvgame==2.5.0
requests==2.31.0
tkinter==8.6
tadm = 4.65.0
```



According to the user demands, codes, requirements, and file format listed above, write a manual .md file to provide detailed introduction to the uses of the software.



Here is the user manual:

Gomoku Game User Manual

Introduction

Welcome to the Gomoku Game!

- 1. Install Python on your computer.
- 2. Open a terminal and run: pip install -r requirements.txt

Running the Game

- 1. Open a terminal and run: python main.py
- 2. Click on the empty cells.
- 3. The game will continue until one player gets five stones in a row.



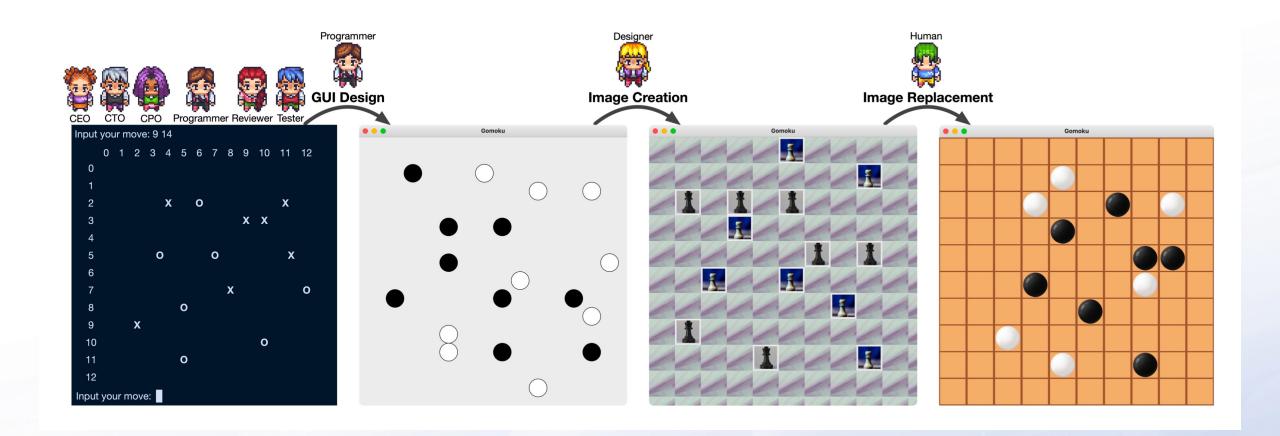
Software Info:

- **cost**=\$0.032732
- **version updates**=5.0
- **num code files**=2
- ★*num png files**=0
- **num doc files**=7
- _**code lines**=103
- ==**env lines**=1
- **manual lines**=35
- ***num utterances**=26
- ***num self reflections**=1
- **num prompt tokens**=14269
- **num completion tokens**=5664
- ***duration**=420.00s

ChatDev Starts (20231018174436)

ChatDev Ends (20231018175136)











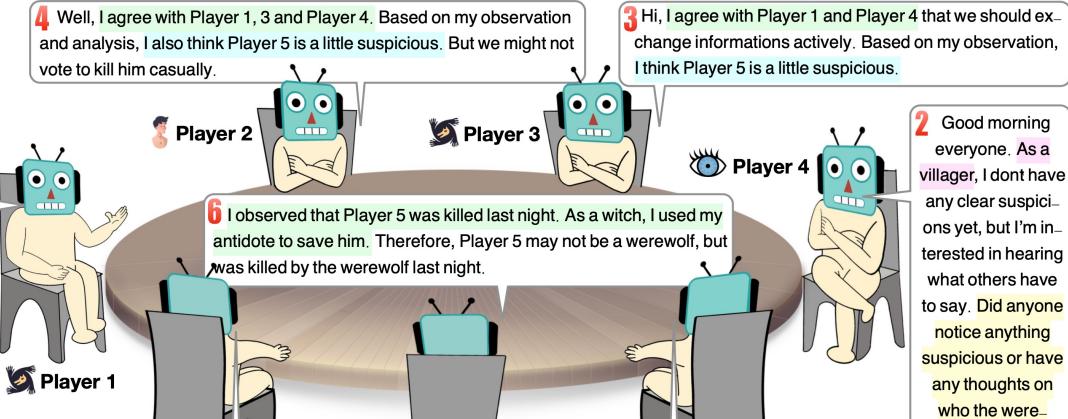




wolves might be?

Good morning everyone. It see_ ms like we had a peaceful night. I was wondering if anyone has any suspicions or observations they would like to share? I think its important for us to start discussing and sharing our thoughts to find out who the

werewolves are.

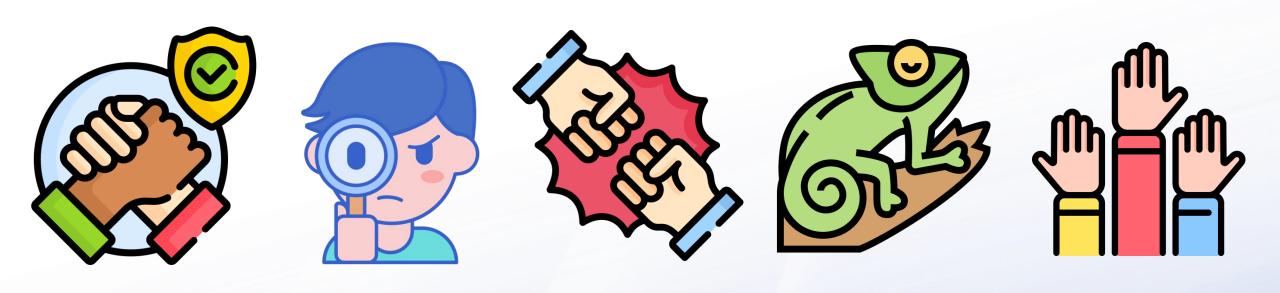


Player 6

I agree with Player1,2 and 6. I think we have to actively gather information and seriously speculate about who the werewolves are. We should protect ourselves as much as possible and avoid being killed by the werewolves. I will use my special ability wisely at night.

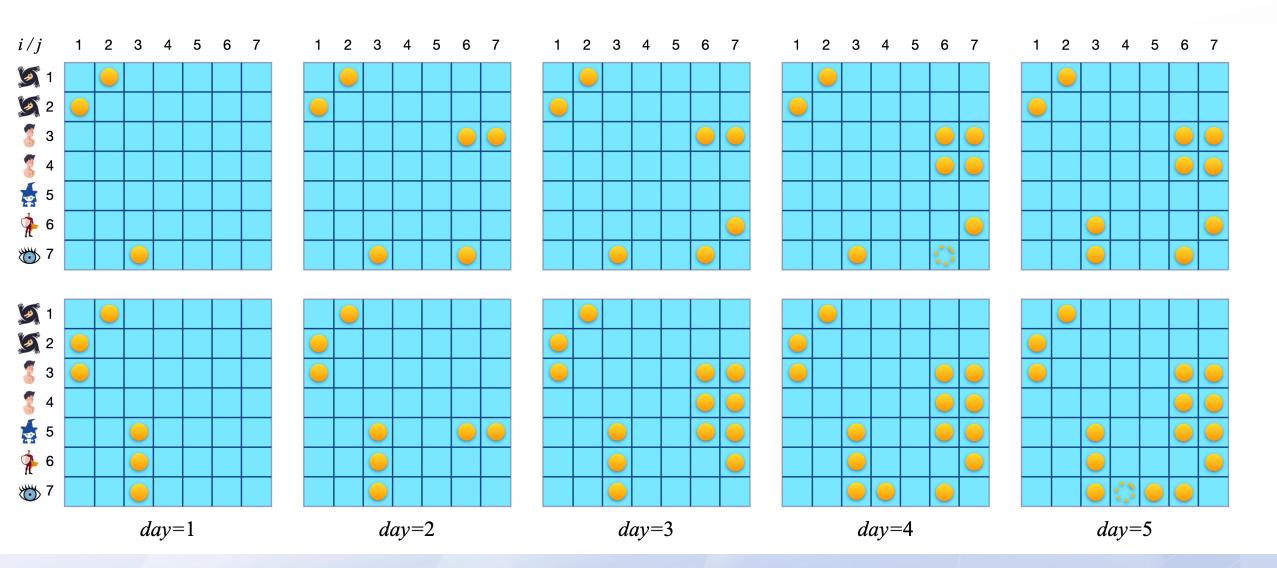
I have to defend myself bacause I am innocent. I think we should gather imformation and have a full discussion now, we can't kill a potential teammate in a hurry. On the contrary, I think Player 3 is a bit aggressive, and I suspect he is one of the werewolf.







AI玩家的信任关系,随游戏进程的变化





更多游戏

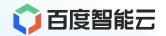
蛛丝轨迹: https://www.bilibili.com/video/BV1pL411273h

AI猫娘密室逃脱: https://www.bilibili.com/video/BV1fQ4y1W7EE

黑客帝国: https://www.youtube.com/watch?v=aihq6jhdW-Q

逆水寒: https://www.bilibili.com/video/BV1Fe4y1A78x

骑马与砍杀2: https://www.bilibili.com/video/BV1nM4y1Q7Ke



关于文心大模型4.0

文心大模型4.0能力介绍







- ·2023年10月17日,百度世界2023大会上,李彦宏宣布文心大模型4.0正式发布,开启邀请测试。
- ·同时也表示文心大模型4.0是迄今为止最强大的文心 大模型,实现了基础模型的全面升级,在**理解、生成、** 逻辑和记忆能力上都有着显著提升。

与此同时,个人及企业客户均可通过 「百度智能云干帆大模型平台」接入文心大模型4.0。

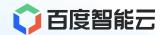






自然语言人机交互会带来提示词革命。未来的应用,是通过自然语言的提示词来调动原生AI应用实现的。

10年后,全世界有50%的工作会是提示词工程。提出问题比解决问题更重要。



百度智能云干帆大模型平台





















百度智能云干帆大模型平台

上手实践



干帆内置大模型

提示词工程

API调用代码模板

知识库插件

数据集及微调





在线Demo: chatmarx.zihaoai.cn

主页: github.com/TommyZihao/ChatMarx







https://www.bilibili.com/video/BV1E94y1v7aa





导师

找到知识库中对应的原文及出处



知识点

输入社会资讯,提取出对应的关键词(与知识库无关)



学霸

根据用户输入的问题,给出对应的回答



研究员

客观评价社会资讯 (与知识库无关)



鲁迅

模仿鲁迅的口吻,评价社会资讯和专家观点(与知识库无关)







导师

{K段原文}



知识点

请从以下社会新闻中提取出至少12个相关的关键词: {用户输入的问题}



学霸

你是一个活泼的学习助手,请根据下面的已知信息,发挥你的想象,详细回答问题,回答限制在300字以内。

问题: **{用户输入的问题}** 已知信息: **{K段原文**}



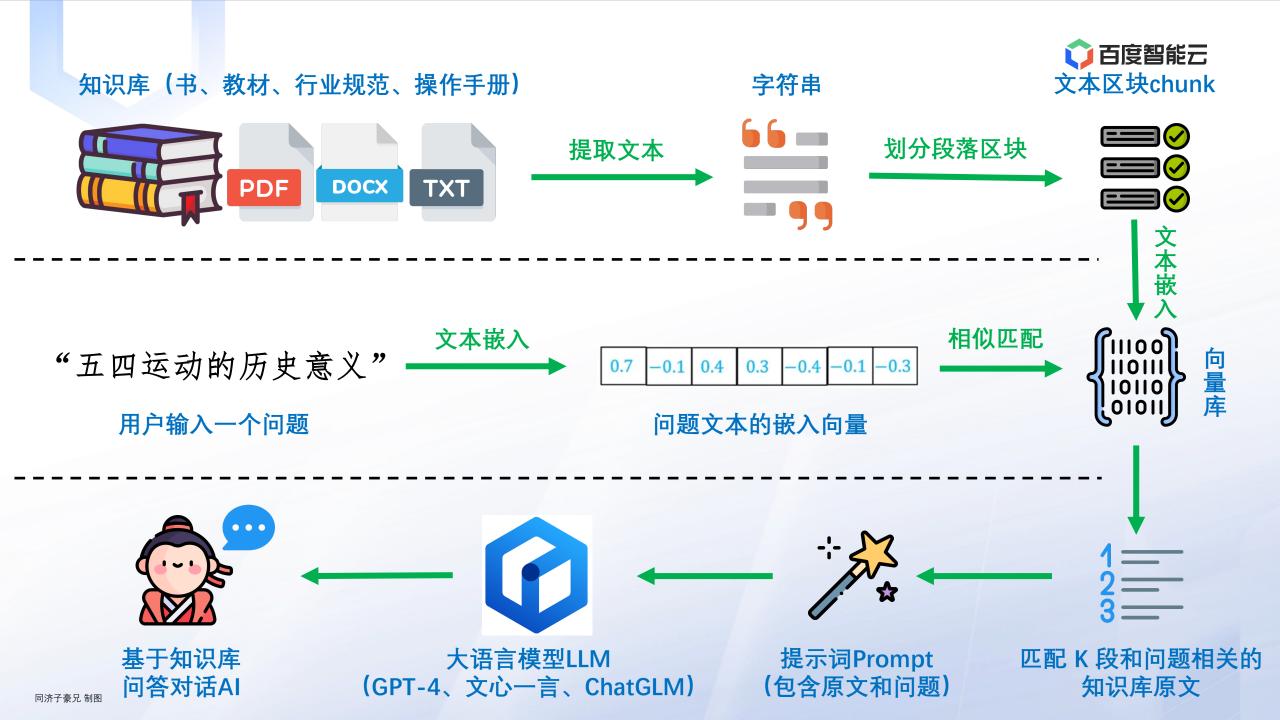
研究员

你是一个客观中立的评论员,请对这件社会时事给出你的见解和看法,字数限制在200字以内: **(用户输入的问题)**



鲁迅

请你以鲁迅的口吻,一句话评价以下社会资讯,字数限制在100字以内: {用户输入的问题}





 XPIN GPT

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你好,我是差评君的人工智能助手,你可以向我提问关于差评君的问题。

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我们公司在杭州余杭区的梦想小镇天使村。

你们公司招土木工程专业的学生吗





这个我还不太会,你去问问杭州麻瓜网络科技有限公司。







我们公司主要做差评君的短视频内容,以及数码账号差评硬件部。差评君的短视频内容主要是 短视频更新,而差评硬件部则主要是做科技、数码相关知识的科普。

差评硬件部都做过哪些 品牌的视频





你们公司叫什么名字?

差评硬件部做过美的、索尼、漫步者、北通、绿联、公牛、摩米士、闪魔、绿巨能等品牌的视频。

0 / 100





课后作业

使用ERNIE-Bot、Qianfan-Chinese-Llama-2-7B和另一个自选模型

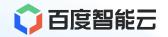
- 1、完成【宿管阿姨npc】复现
- 2、并延展设计一个游戏NPC角色,完成Prompt测试

*将上述作业在【百度智能云干帆社区】进行发布,发布时选择 #大模型实训营 话题



扫码进入 百度智能云千帆社区





答疑与讨论



可添加小助手进入课程群



THANKS