Low-code development and generative Al

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Al-empowered low-code development

- The recent breakthroughs in generative AI through large language models (LLMs) have opened up many new possibilities for text, image, and code generation.
- Low-code development platforms have, therefore, started experimenting with LLMs for program synthesis.
- Program synthesis is the task of automatically finding programs from the underlying programming language that satisfy user intent expressed in some form of constraints.

Al-empowered low-code development

- Program synthesizers typically perform some form of search over the space of programs to generate a program that is consistent with a variety of constraints.
 - These constrains can be in the form of input-output examples, demonstrations, natural language, partial programs, assertions, etc.
- The problem of program synthesis has long been considered the holy grail
 of Computer Science.



The evolution of generative AI models will further lower barriers to using lowcode and no-code development tools, and potentially lead to the birth of a whole new class of intelligent developer technology.















Q ≡

Home > Software Development

Mendix aims to add generative AI to its low-code platform by year-end

The company has been trying out GPT-based features for its low-code platform, planning to add capabilities to streamline app development and offer guidance on best practices.



UNITED STATES .















Microsoft | Power Apps | Product v | Pricing | Partners v | Learn v | Support v | Community v

Announcing a next-generation Al Copilot in Microsoft Power Apps that will transform low-code development



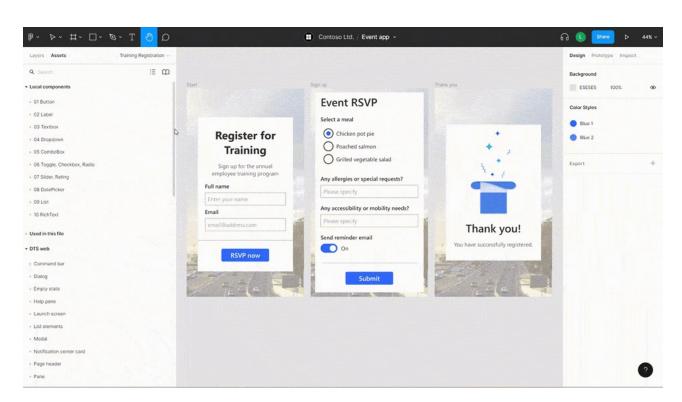


Al-empowered low-code development

 Early analysis¹ indicates that the market for generative AI for application development will be \$4.1 billion by the end of 2023, growing at approximately 32% a year.



From design inputs to working low-code apps







Sketch a form and turn it into an app with Power Apps express design

Conversational Al

- Conversational Al is a type of artificial intelligence (Al) that can simulate human conversation.
- It works by using a combination of natural language processing (NLP) and machine learning (ML).
- The model is **constantly learning** from its interactions and improving its response quality over time.
- Conversational AI has been receiving a lot of attention due to the latest breakthroughs in large language models (LLMs).

I need to write a Python function called count common. Here's a description: 'Write a function to count the most common words in a dictionary.'. Can you give me an example? assert count common(['one', 'two', 'three', 'four', 'five', 'one', 'two', 'one', 'three', 'one']) == [('one', 4), ('two', 2), ('three', 2), ('four', 1)] def count_common(list_of_words): $dict = {}$ for word in list of words: if word in dict: dict[word] += 1 else: dict[word] = 1 return sorted(dict.items(), key=lambda x: x[1]) Close! But can you sort the output in reverse (descending) order? def count_common(list_of_words): $dict = {}$ for word in list of words: tests fail if word in dict: dict[word] += 1 else: dict[word] = 1return sorted(dict.items(), kev=lambda x; x[1], reverse=True) Great. Now notice that the example only returns the top 4 results. def count_common(list_of_words): $dict = {}$ for word in list of words: tests pass if word in dict: dict[word] += 1 else: dict[word] = 1 return sorted(dict.items(), key=lambda x: x[1], reverse=True)[:4] Hooray! 🎉 🎉 🎉

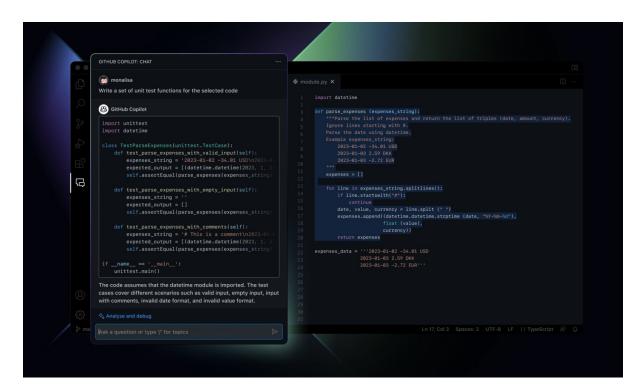
I need to write a Python function called find Max Num. Here's a description: 'Write a python function to find the largest number that can be formed with the given list of digits.'. Can you give me an example? assert find Max Num([1,2,3]) == 321 def find Max Num(test list): largest = 0 for i in range(0,len(test list)): current = test_list[i] for j in range(0,len(test list)): if (current == test list[i]): current = current * 10 else: break if (largest < current): largest = current return largest This function should sort the list in decreasing order and then combine the digits into an integer. def find Max Num(test list): 8 return sorted(test list)[-1] tests fail Join the string representations of the values in the sorted list. def find_Max_Num(test_list): return ".join(map(str, sorted(test list))) tests fail Sort the list from highest to lowest, not lowest to highest. def find Max Num(test list): return ".join(map(str, sorted(test_list, reverse=True))) tests fail Return the int of the string.

return int(".join(map(str, sorted(test_list, reverse=True))))

def find Max Num(test list):



GitHub Copilot X





Low-code: Can we just ask for the program?

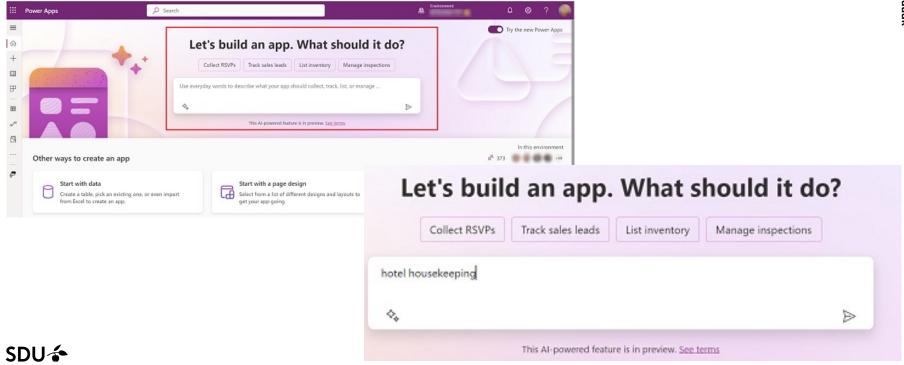
I need an input form with a text box, a submit button, a cancel button, and the Salesforce logo.



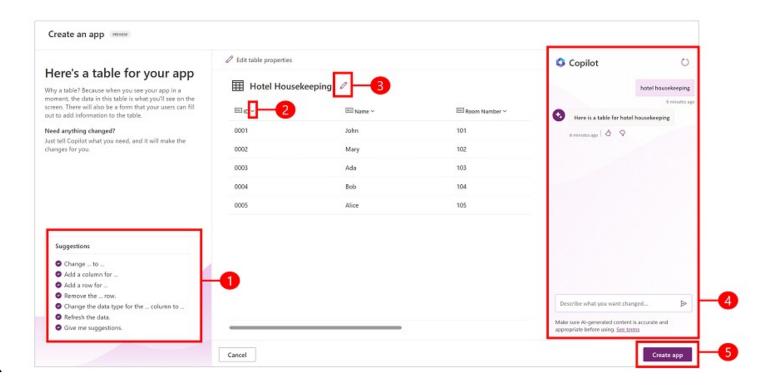




Low-code and conversational Al

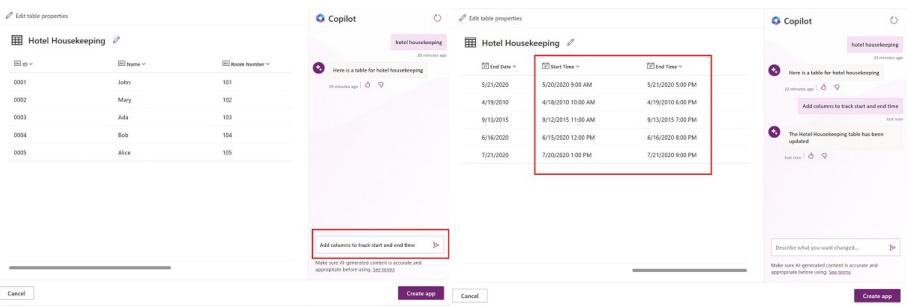


Low-code and conversational Al





Low-code and conversational Al





But....

Garbage in, garbage out has never been more true.

So...

Prompt Engineering (asking the right question) has never been so important.

Prompt Engineering

Prompt engineering is the process of designing and optimizing prompts to LLMs for a wide variety of applications and research topics. Prompts are short pieces of text that are used to guide the LM's output. They can be used to specify the task that the LM is supposed to accomplish, provide additional information that the LM needs to complete the task, or simply make the task easier for the LM to understand.

Source: Google Bard



Prompt Engineering

- Prompt engineering is enabled by in-context learning, which is a model's ability to temporarily learn from prompts.
 - In contrast to training and fine tuning for specific tasks, which are not temporary.
- *In-context learning* has been discovered as an *emergent ability* of LLMs (Wei et al., 2022) and it's a form of **meta-learning** or "learning to learn".
- At present, the mechanisms of in-context learning are not well understood and remain mostly an intuition (vonOswald et al., 2023; Garg et al., 2023).



One of the reasons we should care about the prompt: LLMs are non-deterministic!

"Code generation models have achieved impressive performance. However, they tend to be brittle as slight edits to a prompt could lead to very different generations." (Wang et al., 2022)



What is 1 + 1?

```
What's 1 + 1?
>> 1+1=2
What's 1 + 1?
>> 1 + 1 = 2.
What's 1 + 1?
>> 1 + 1 = 2.
What's 1 + 1?
>> 1 + 1 = 2.
What's 1 + 1?
>> I understand that you are asking me what is 1 + 1. The answer is 2.
```

Chat with Google Bard



Engineering the Question

```
What's 1+1 stated as an integer?

>> 1+1 stated as an integer is 2.
An integer is a whole number that can be positive, negative, or zero. It is no
```

Chat with Google Bard (even worse)



Engineering the Question

```
What's 1+1? Only return the numeric answer.

>> 2
What's 1+1? Only return the numeric answer.

>> 2
What's 1+1? Only return the numeric answer.

>> 2

What's 1+1? Only return the numeric answer.

>> 2
```

Chat with Google Bard (better, more deterministic)



Prompt Engineering Techniques



Prompt Engineering Techniques

- Zero-Shot Prompting
- Few-Shot Prompting
- One-Shot Prompting
- Chain-of-Thought (CoT) Prompting
- Zero-shot COT Prompting
- Generated Knowledge Prompting

- Least-to-most prompting
- Self-consistency decoding
- Complexity-based prompting
- Self-refine
- Tree of Thoughts
- Maieutic prompting
- Retrieval Augmented Generation
- Automatic Reasoning and Tool-use
- Automatic Prompt Engineer
- Active-Prompt
- Directional Stimulus Prompting
- ReAct
- Multimodal CoT
- Graph Prompting
- .



Zero-Shot Prompting

 In zero-shot prompting, no examples are provided to the model, and only the task request is provided.

```
Prompt:
```

```
Classify the text into neutral, negative or positive.
Text: I think the vacation is okay.
Sentiment:
```

Output:

Neutral



Few-Shot Prompting

- While LLMs demonstrate remarkable zero-shot capabilities, they still fall short on more complex tasks when using the zero-shot setting.
- Few-shot prompting can be used as a technique to enable in-context
 learning where we provide demonstrations in the prompt to steer the model
 to better performance.
- The demonstrations serve as **conditioning** for subsequent examples where we would like the model to generate a response.

Few-Shot Prompting

 The user includes several examples in the prompt that demonstrate the expected answer format and content.

```
Copy
Convert the questions to a command:
Q: Ask Constance if we need some bread
A: send-msg `find constance` Do we need some bread?
Q: Send a message to Greg to figure out if things are ready for Wednesday.
A: send-msg `find greg` Is everything ready forWednesday?
Q: Ask Ilya if we're still having our meeting this evening
A: send-msq `find ilya` Are we still having a meetingthis evening?
Q: Contact the ski store and figure out if I can getmy skis fixed before I leave on Thursday
A: send-msg `find ski store` Would it be possible toget my skis fixed before I leave on Thursday
0: Thank Nicolas for lunch
A: send-msg `find nicolas` Thank you for lunch!
Q: Tell Constance that I won't be home before 19:30tonight - unmovable meeting.
A: send-msg `find constance` I won't be home before19:30 tonight. I have a meeting I can't move.
0: Tell John that I need to book an appointment at10:30
A:
```



One-Shot Prompting

 This case is the same as the few-shot approach except only one example is provided.

```
Convert the questions to a command:
Q: Ask Constance if we need some bread
A: send-msg `find constance` Do we need some bread?
Q: Send a message to Greg to figure out if things are ready for Wednesday.
A:
```



Zero-Shot Prompting

Convert the question to a command:
Q: Ask Constance if we need some bread
A:



Chain-of-Thought (CoT) Prompting (Wei et al., 2022)

- Chain-of-Thought (CoT) prompting enables complex reasoning capabilities through intermediate reasoning steps.
- Can be combined with few-shot prompting to get better results on more complex tasks that require reasoning before responding.



Chain-of-Thought (CoT) Prompting (Wei et al., 2022)

Standard Prompting

Model Input

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 tennis balls. How many tennis balls does he have now?

A: The answer is 11.

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

Model Output

A: The answer is 27.



Chain-of-Thought Prompting

Model Input

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 tennis balls. How many tennis balls does he have now?

A: Roger started with 5 balls. 2 cans of 3 tennis balls each is 6 tennis balls. 5 + 6 = 11. The answer is 11.

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

Model Output

A: The cafeteria had 23 apples originally. They used 20 to make lunch. So they had 23 - 20 = 3. They bought 6 more apples, so they have 3 + 6 = 9. The answer is 9. <



Zero-shot COT Prompting (Kojima et al., 2022)

Essentially involves adding "Let's think step by step" to the original prompt.

(a) Few-shot

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 tennis balls. How many tennis balls does he have now?

A: The answer is 11.

Q: A juggler can juggle 16 balls. Half of the balls are golf balls, and half of the golf balls are blue. How many blue golf balls are there?

A:

(Output) The answer is 8. X

(c) Zero-shot

Q: A juggler can juggle 16 balls. Half of the balls are golf balls, and half of the golf balls are blue. How many blue golf balls are there?

A: The answer (arabic numerals) is

(Output) 8 X

(b) Few-shot-CoT

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 tennis balls. How many tennis balls does he have now?

A: Roger started with 5 balls. 2 cans of 3 tennis balls each is 6 tennis balls. 5 + 6 = 11. The answer is 11.

Q: A juggler can juggle 16 balls. Half of the balls are golf balls, and half of the golf balls are blue. How many blue golf balls are there?

A:

(Output) The juggler can juggle 16 balls. Half of the balls are golf balls. So there are 16 / 2 = 8 golf balls. Half of the golf balls are blue. So there are 8 / 2 = 4 blue golf balls. The answer is 4.

(d) Zero-shot-CoT (Ours)

Q: A juggler can juggle 16 balls. Half of the balls are golf balls, and half of the golf balls are blue. How many blue golf balls are there?

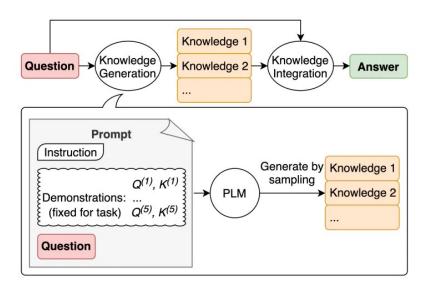
A: Let's think step by step.

(Output) There are 16 balls in total. Half of the balls are golf balls. That means that there are 8 golf balls. Half of the golf balls are blue. That means that there are 4 blue golf balls.



Generated Knowledge Prompting (Liu et al., 2021)

 Generated Knowledge Prompting involves incorporating knowledge or information to help the model make more accurate predictions.





LLM Prompting for Programming Tasks

LLM prompting for programming

An example from Martin Fowler and Xu Hao, Thoughtworks.

Source: https://martinfowler.com/articles/2023-chatgpt-xu-hao.html

- The objective is to build Self-Testing Code.
- It starts with a prompt that sets the context for the application and how you want the code to be structured.
- Even though the final goal is to generate code and its tests, ChatGPT is initially instructed: "Don't generate code".



SDU Software Engineering

The current system is an online whiteboard system. Tech stack: typescript, react, redux, konvajs and react-konva. And vitest, react testing library for model, view model and related hooks, cypress component tests for view.

All codes should be written in the tech stack mentioned above. Requirements should be implemented as react components in the MVVM architecture pattern.

There are 2 types of view model in the system.

- Shared view model. View model that represents states shared among local and remote users.
- 2. Local view model. View model that represents states only applicable to local user

Here are the common implementation strategy:

- 1. Shared view model is implemented as Redux store slice. Tested in vitest.
- Local view model is implemented as React component props or states(by useState hook), unless for global local view model, which is also implemented as Redux store slice. Tested in vitest.
- Hooks are used as the major view helpers to retrieve data from shared view model. For most the case, it will use 'createSelector' and 'useSelector' for memorization. Tested in vitest and react testing library.
- 4. Don't dispatch action directly to change the states of shared view model, use an encapsulated view model interface instead. In the interface, each redux action is mapped to a method. Tested in vitest.

View is consist of konva shapes, and implemented as react component via react-konva.Tested in cypress component tests

Here are certain patterns should be followed when implement and test the component

- 1. When write test, use describe instead of test
- 2. Data-driven tests are preferred.
- 3. When test the view component, fake view model via the view model interface

Awareness Layer

Requirement:

Display other users' awareness info(cursor, name and online information) on the whiteboard.

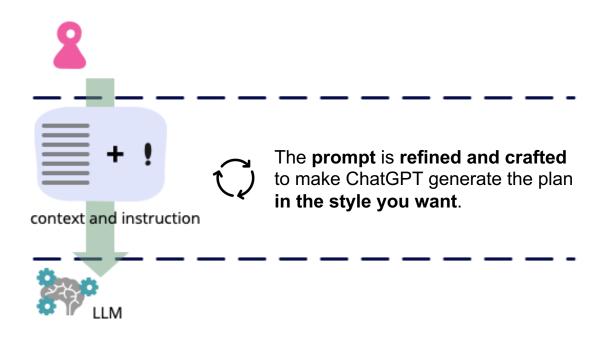
AC1: Don't display local user

AC2: When remote user changes cursor location, display the change in animation.

Provide an overall solution following the guidance mentioned above. Hint, keep all awareness information in a Konva layer, and an awareness info component to render cursor, and name. Don't generate code. Describe the solution, and breaking the solution down as a task list based on the guidance mentioned above. And we will refer this task list as our master plan.



Initial Prompt to ChatGPT: Design Guidelines



LLM prompting for programming

An example from Martin Fowler and Xu Hao, Thoughtworks.

Source: https://martinfowler.com/articles/2023-chatgpt-xu-hao.html

- Initial prompt primes the LLM with an implementation strategy and encourages the model to explain its reasoning (chain of thought prompting).
- Prompt initially asks for an implementation plan rather than code (generated knowledge prompting).
- The plan is used later to refine the implementation and generate useful sections of code.



LLM prompting for programming

An example from Martin Fowler and Xu Hao, Thoughtworks.

Source: https://martinfowler.com/articles/2023-chatgpt-xu-hao.html

- This prompt technique is called Generated Knowledge¹.
 - ❖ First, we ask the LLM to generate some useful information about the problem, and then
 - ❖ We **feed that information back** into the LLM to generate the final product.



Solution:

To implement the awareness layer feature, we can follow the guidelines mentioned in the problem statement and use the existing tech stack.

Task List:

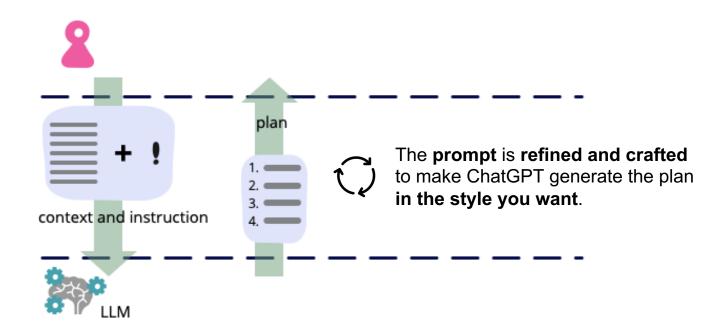
- 1. Create a new component named "AwarenessInfo" to render cursor and name for remote users.
- 2. Create a new Konva layer to hold all the "AwarenessInfo" components.
- 3. Add the Konva layer to the main whiteboard component.
- 4. Create a new Redux slice to store the awareness information for remote users.
- 5. Create an encapsulated view model interface for the awareness Redux slice.
- 6. Implement the "AwarenessInfo" component to retrieve and display the awareness information for each remote user from the view model.
- 7. Implement a method in the encapsulated view model interface to update the awareness information for each remote user.
- 8. When a remote user changes the cursor location, dispatch the update method in the encapsulated view model interface to update the corresponding awareness information.
- 9. In the "AwarenessInfo" component, use the new location to animate the cursor change.
- 10. Write tests for the new "AwarenessInfo" component, including data-driven tests to ensure it properly displays the awareness information for all remote users.
- 11. Write tests for the encapsulated view model interface to ensure it properly updates the awareness information for remote users.
- 12. Write end-to-end Cypress tests to ensure the awareness layer is properly displayed and updated when remote users change their cursor location.

By following these tasks, we can implement the awareness layer feature in the online whiteboard system.

The key element of the response is an ordered task list, with numbers for each step. We can use these numbers to refer to these tasks for the rest of the session.



#sdudk



SDU Software Engineering

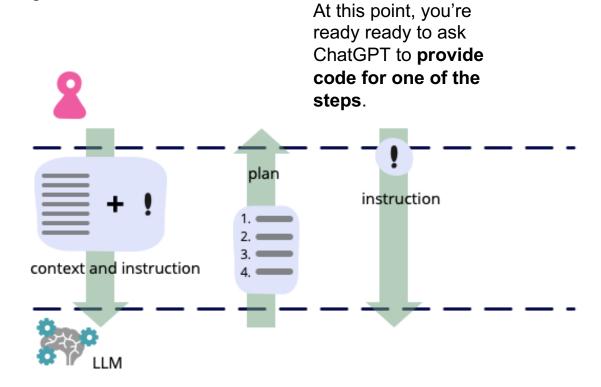
Rewrite the master plan, this time with detailed component names, methods, and props

Sure, here's a revised master plan with detailed component names, methods, and props:

Task List:

- Create a new component named "AwarenessInfo" to render cursor and name for remote users. This component will receive props such as user name, cursor position, and online status.
- Create a new Konva layer named "AwarenessLayer" to hold all the "AwarenessInfo" components.
- 3. Add the "AwarenessLayer" to the main whiteboard component.
- Create a new Redux slice named "awareness" to store the awareness information for remote users.
- 5. Create an encapsulated view model interface named "AwarenessViewModel" for the "awareness" Redux slice. This interface will contain a method named "updateUserAwarenessInfo" to update the awareness information for each remote user.
- 6. Implement the "AwarenessInfo" component to retrieve and display the awareness information for each remote user from the view model. The component will use the "useSelector" hook to retrieve the relevant information.
- 7. Implement the "updateUserAwarenessInfo" method in the "AwarenessViewModel" interface to dispatch a Redux action to update the awareness information for each remote user. This method will be called by the "MainWhiteboard" component.
- 8. When a remote user changes the cursor location, dispatch the "updateUserAwarenessInfo" method in the "AwarenessViewModel" interface to update the corresponding awareness information.
- In the "AwarenessInfo" component, use the new location to animate the cursor change. This can be accomplished using the Konva "Tween" animation.
- 10. Write tests for the new "AwarenessInfo" component, including data-driven tests to ensure it properly displays the awareness information for all remote users. The tests will be written using the React Testing Library.
- 11. Write tests for the encapsulated view model interface to ensure it properly updates the awareness information for remote users. These tests will use the Vitest framework.
- 12. Write end-to-end Cypress tests to ensure the awareness layer is properly displayed and updated when remote users change their cursor location. The tests will simulate user interactions with the whiteboard and verify the state of the "AwarenessLayer" and "AwarenessInfo" components.





Sure, let's start with writing tests for the awareness slice and AwarenessViewModel:

const initialState: AwarenessState = {
 cursorLocation: { x: 0, y: 0 },
 name: '',

cursorLocation:
name: '',
isOnline: false,
;

Awareness slice:

name: string;

isOnline: boolean;

interface AwarenessState {

cursorLocation: { x: number; y: number };

};
const awarenessSlice = createSlice({
 name: 'awareness',
 initialState,

initialState,
reducers: {
 setCursorLocation: (state, action: PayloadAction<{ x: number; y: number }>) => {
 state.cursorLocation = action.payload;
 },
 setName: (state, action: PayloadAction) => {

state.name = action.payload;
),
setOnline: (state, action: PayloadAction) => (
 state.isOnline = action.payload;
),
),

},
});
export const (setCursorLocation, setName, setOnline) = awarenessSlice.actions;

export const { setCursorLocation, setNote
export default awarenessSlice.reducer;
AwarenessViewModel:

AwarenessViewModel:

class AwarenessViewModel {
 private store: Store;

constructor(store: Store) {
 this.store = store;
}
setCursorLocation(cursorLocation: { x: number; y: number }) {

this.store.dispatch(setCursorLocation(cursorLocation));
}
setName(name: string) {
 this.store.dispatch(setName(name));
}

}
setOnline(isOnline: boolean) {
 this.store.dispatch(setOnline(isOnline));
}

Note: In this implementation, I assumed that the AwarenessViewModel is used to set the values in the store. If it is also used to retrieve the values from the store, we can add corresponding methods to the AwarenessViewModel class

DU∜

Test for Awareness Slice:

describe('awareness slice', () => {

provide me example tests and implementation for the awareness slice and

let store: MockStore;
beforeEach(() => {
 store = configureMockStore()(initialState);
));

it('should set the cursor location', () => {
 store.dispatch(setCursorLocation({ x: 100, y: 200

AwarenessViewModel. Test first.

it('should set the cursor location', () => {
 store.dispatch(setCursorLocation((x: 100, y: 200)));
 const state = store.getState().awareness;
 expect(state.cursorLocation).toEqual((x: 100, y: 200));
));

it('should set the name', () => {

store.dispatch(setName('John Doe'));
const state = store.getState().awareness;
expect(state.name).toBe('John Doe');
));
it('should set the online status', () => {
 store.dispatch(setOnline(true));
 const state = store.getState().awareness;
 synch('state.isonline).toBe('true).

expect(state.isOnline).toBe(true);
));
));
Test for AwarenessViewModel:
describe('AwarenessViewModel', () => {
let store: MockStore:

let awarenessViewModel: AwarenessViewModel;

beforeEach(() => {
 store = configureMockStore()(initialState);
 awarenessYiewModel = new AwarenessYiewModel(store);
));

it('should set the cursor location', () => {
 awarenessYiewModel.setCursorLocation((x: 100, y: 200));
 const state = store.getState().awareness;
 expect(state.cursorLocation).toEqual((x: 100, y: 200));
));

expect(state.name).toBe('John Doe');
));

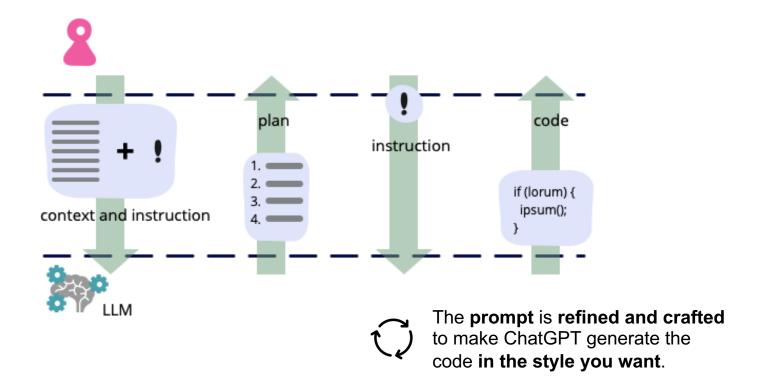
it('should set the online status', () => {
 awarenessViewModel.setOnline(true);
 const state = store.getState().awareness;
 expect(state.isOnline).toBe(true);

awarenessViewModel.setName('John Doe'):

const state = store.getState().awareness;

it('should set the name', () => {

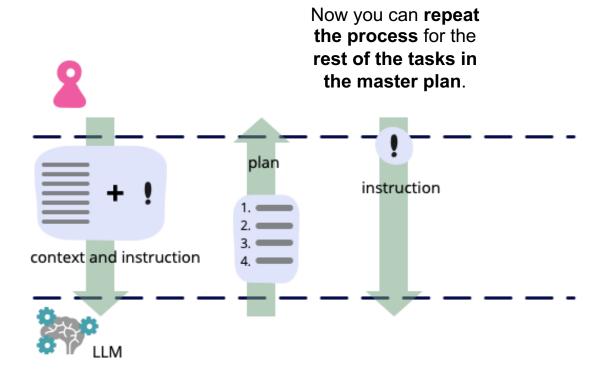
const state = store.getState().awareness;
expect(state.isOnline).toBe(true);
});
});
Now, let's implement the Awareness Slice and AwarenessViewModel:



});

```
Sure, here's an example implementation of the tests for the AwarenessViewModel and the
awareness slice without using redux-mock-store:
import { createAwarenessSlice, AwarenessState, updateAwareness } from './awarenessSlice';
import { AwarenessViewModel } from './AwarenessViewModel';
describe('AwarenessViewModel', () => {
 const mockDispatch = jest.fn();
 const initialState: AwarenessState = {
   users: {
     'user1': {
       name: 'User 1'.
       cursorPosition: { x: 0, y: 0 },
       online: true,
      'user2': {
       name: 'User 2',
       cursorPosition: { x: 100, y: 100 },
       online: true,
     }.
 }:
 let viewModel: AwarenessViewModel;
 beforeEach(() => {
   viewModel = new AwarenessViewModel(initialState, mockDispatch);
 });
  describe('getUserNames', () => {
   it('returns the list of user names', () => {
     const names = viewModel.getUserNames();
     expect(names).toEqual(['User 1', 'User 2']);
   });
 }):
  describe('getCursorPosition', () => {
   it('returns the cursor position for a given user ID', () => {
     const cursorPosition = viewModel.getCursorPosition('user1');
     expect(cursorPosition).toEqual({ x: 0, y: 0 });
   }):
   it('returns undefined for an unknown user ID', () => {
     const cursorPosition = viewModel.getCursorPosition('unknown');
     expect(cursorPosition).toBeUndefined();
   3);
 });
 describe('updateCursorPosition', () => {
   it('dispatches the updateAwareness action with the new cursor position', () => {
     viewModel.updateCursorPosition('user1', { x: 50, y: 50 });
     expect(mockDispatch).toHaveBeenCalledWith(updateAwareness({
       user: 'user1'.
       cursorPosition: { x: 50, y: 50 },
     }));
   }):
 });
```

```
And here's an example implementation of the awarenessSlice without using redux-mock-
store:
import { createSlice, PayloadAction } from '@reduxis/toolkit':
export type CursorPosition = { x: number, v: number }:
export type UserAwareness = {
  name: string;
 cursorPosition: CursorPosition;
  online: boolean;
};
export type AwarenessState = {
  users: Record<string, UserAwareness>;
};
const initialState: AwarenessState = {
  users: {},
};
const awarenessSlice = createSlice({
  name: 'awareness',
  initialState.
  reducers: {
    updateAwareness: (state, action: PayloadAction<{ user: string, cursorPosition: CursorPosition
      const { user, cursorPosition } = action.payload;
      state.users[user].cursorPosition = cursorPosition;
    }.
    setOnline: (state, action: PayloadAction<{ user: string, online: boolean }>) => {
      const { user, online } = action.payload;
      state.users[user].online = online;
  },
});
export const { updateAwareness, setOnline } = awarenessSlice.actions;
export const awarenessReducer = awarenessSlice.reducer;
Note that since we're not using redux-mock-store, we need to pass a mock dispatch
function to the view model constructor, and in the test for updateCursorPosition we need
to check that the correct action was dispatched using jest.fn().
```



LLM prompting for programming

An example from Martin Fowler and Xu Hao, Thoughtworks.

Source: https://martinfowler.com/articles/2023-chatgpt-xu-hao.html

- You can have problems with the limited context window (also called "token limit").
 - ChatGPT may stop generating content: just say something like "you are not finished" or "go on".
 - ❖ If the overall prompt is too big, then ChatGPT will simply return a fatal error and we have to start a new conversation.
 - ChatGPT may start forgetting and losing context. This is where the approach of getting a master plan and breaking down the task into separate elements helps.

Other examples: GPT Engineer

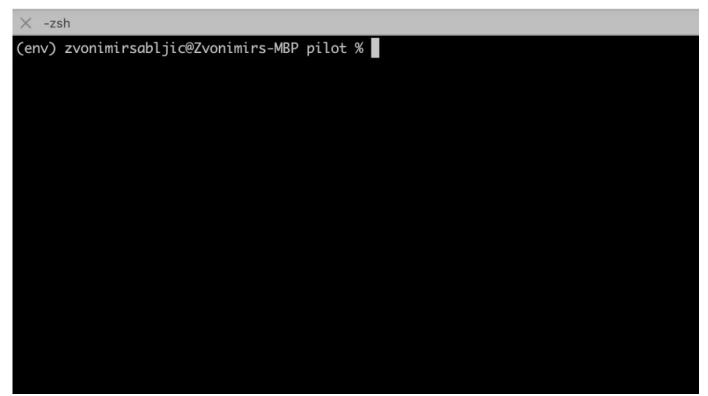
https://github.com/AntonOsika/gpt-engineer

```
(base) → gpt_engineer git:(main) x
```



Other examples: GPT PILOT

https://github.com/Pythagora-io/gpt-pilot





Low-code development and generative Al

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