



# Scottish Graduate School of Social Science

Sgoil Cheumnaichean Saidheans  
Sòisealta na h-Alba

## Collecting Digital Data

LLMs as Coding Assistants

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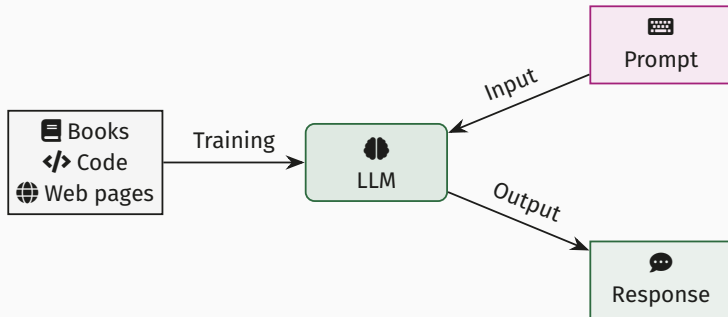
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## **LLMs as Coding Assistants**

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# What Are LLMs?

A **Large Language Model** is a statistical model trained on vast amounts of text and code.



You provide a **prompt** (a question or instruction), and the model generates a **response**.

## Paid vs Open Models

<b>Paid (ChatGPT, Claude)</b>	<b>Open (Llama, Mistral)</b>
Hosted by provider	Run locally or self-host
Subscription / API fees	Free to use
Often more capable	Improving rapidly
Data sent to provider	Data stays with you
Easy to use (chat UI)	Requires some setup

### Tip

For this course, any chat-based LLM will work — use whichever you have access to.

# How to Prompt for Code

## ✓ Good prompt

“Write a Python script using `requests` to download stop-and-search data from the UK Police API for the City of London force. Save the results as a JSON file.”

“Get me some police data.”

## What makes a good prompt?

- Specify the **programming language** and **libraries**
- Provide the **exact URL** or API endpoint
- State the **desired output** format
- Include relevant **parameters** (e.g., force name, date range)

# What LLMs Are Good At

- ✓ **Boilerplate code** — standard request/response patterns, file I/O, data wrangling
- ✓ **Syntax and library usage** — “How do I use `pandas.json_normalize`?”
- ✓ **Explaining error messages** — paste a traceback and ask what went wrong
- ✓ **Converting between languages** — turn a Python script into R (or vice versa)
- ✓ **Generating documentation** — docstrings, comments, README files

## Key insight

LLMs excel at tasks that are **well-documented** and **widely practised** online.






## What LLMs Are Bad At

- ✗ **Up-to-date API endpoints/parameters** — training data has a cut-off date
- ✗ **Complex multi-step logic** — long workflows with many dependencies
- ✗ **Understanding your specific research context** — it doesn't know your data
- ✗ **Guaranteeing correctness** — output looks plausible but may be wrong
- ✗ **Hallucinated function names** — confidently invents functions that don't exist

LLMs are **confident**, not **correct**. Always verify their output.

## Critical Evaluation of LLM-Generated Code

You **must** understand the code yourself. If you can't explain what it does, don't use it.

-  **Always test** LLM-generated code — run it, inspect the output
-  **Check API docs** match what the LLM suggests — endpoints, parameters, authentication
-  **Be wary of hallucinated libraries/functions** — verify they actually exist
-  **Consider reproducibility** — LLM outputs vary between sessions and models
-  **Security risks** — never paste API keys, passwords, or sensitive data into a chat



### Practical 4: LLM Showdown

- Compare different LLMs on the same coding task
- Evaluate the quality, correctness, and usability of generated code
- Apply the critical evaluation criteria from this lecture

Use your preferred LLM (ChatGPT, Claude, Llama, etc.)  
to complete the challenge tasks.

# Thank You!

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[www.brawdata.co.uk](http://www.brawdata.co.uk)



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Course materials on GitHub



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Thank you for attending the **SGSSS Collecting Digital Data** course.

Good luck with your research!