

Integrating AI in Stata Programming: Perspectives on Enhancement and Constraints Across Skill Levels

Ricardo Mora
Dept. Economics, UC3M

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WORKSHOP ON EMPIRICAL RESEARCH IN THE AI ERA
BPLIM

Introduction

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This study focuses on **ChatGPT** “current” capabilities in assisting **Stata** (17) users.

- ChatGPT is widely popular and is acknowledged to exemplify the best of LLM abilities.
- Stata is widely used for data analysis in social sciences.

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- **Programming tasks:** interactively writing and debugging Stata code to create software implementations.
LLMs help highly skilled users the most.
- **Troubleshooting errors:** identify and resolve multiple issues within the code simultaneously, without iterative debugging or gradual corrections.
This presentation focuses mainly on this last point.

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- **Autonomy** within cooperation.
- Cognitive **flexibility**

Experiment design

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- **ChatGPT API:** Python is used to interact with ChatGPT in batch mode.
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- **User levels:**
 - Beginner: Foundational tasks, focusing on simple data manipulation and regression analysis (OLS). [▶ Original](#)
 - Intermediate: Efficient data handling with loops and conditional analyses; handles multiple (IV) models with organized output. [▶ Original](#)
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 - Model 4o: Released in 2024, more accurate.
- **Openbook (closed book)** AI model is (not) provided with additional external resources.
 - In openbook mode, the API request includes a Stata log file. In closed-book mode, the request only includes the do file.

Sample

Sample

- 225 ChatGPT API requests per user level
 - $225 = 75 \text{ do files} \times (3.5\text{-turbo in closed-book mode} + 3.5\text{-turbo in openbook} + 40 \text{ in openbook})$
 - **675** ($= 225 \times 3 \text{ user levels}$) requests in total.

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- Erroneous scripts:
 - **Typographical** errors ($\approx 1/3$): mistakes made during the manual entry of code, such as misspelled words or incorrect punctuation.
 - Nonexistent **Commands/options** ($\approx 1/3$): user attempts to execute commands/options that are foreign to Stata.
 - **Syntax** issues ($\approx 1/3$):
 - Command/option code does not follow Stata grammar.
 - Erroneous overall structure of the code leads to program abortion.

► Detailed results

ChatGPT API Requests

First Request

- Closed-book mode

You are an expert in Stata programming. For the uploaded do file, provide a brief explanation of why Stata gives an error message and suggest a code solution to prevent the error and avoid interruption of the do file. Be concise.

- Openbook mode

You are an expert in Stata programming. For the uploaded do file and its corresponding log file, provide a brief explanation of why Stata gives an error message and suggest a code solution to prevent the error and avoid interruption of the do file. Be concise.

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Second request (only in openbook code)

You are an expert in Stata programming. You are required to provide a response in one of the following formats ONLY: 'Yes', 'No', or 'I do not know'. Your task is to answer the question: 'Do you know what the error is?' based on the provided Stata do file and its corresponding log file.

Evaluation metrics

With the sample of erroneous API requests

- **Solution effectiveness:** Percentage of requests resulting in acceptable Stata solutions.
- **Hallucination incidence:** Percentage of requests using arguments that invoke erroneous Stata behavior.
- **Illusory expertise:** Percentage of requests triggering erroneous Stata solutions conditional to ChatGPT stating that it knows the answer.

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With the full sample

- **Hammer effect:** The effect on solution effectiveness of being requested “(to) provide a brief explanation of why Stata gives an error message” when there is actually no error in the script.
 - Saying that there is no error in the script counts as acceptable ‘solution’.

Solution effectiveness

ChatGPT performance by openbook mode and user's level

	Closed-book mode		Openbook mode	
	Failure rate	Success rate	Failure rate	Success rate
Beginner	81.3	18.7	16.0	84.0
Intermediate	81.3	18.7	26.0	74.0
Advanced	84.0	16.0	58.7	41.3
All users	82.2	17.8	33.6	66.4

Notes: Sample of API requests with erroneous code. Failure and success rates are percentages of requests where ChatGPT does not provide an acceptable solution (Failure) or otherwise). Request under ‘Closed-book mode’ are requests without access to the error log file.

Success rates by error type, openbook mode, model and user's level

	Typographical errors			Command errors			Syntax errors		
	Close book		Openbook	Close book		Openbook	Close book		Openbook
	3.5-turbo	3.5-turbo	4o	3.5-turbo	3.5-turbo	4o	3.5-turbo	3.5-turbo	4o
Beginner	45.5	95.5	100.0	10.0	90.0	100.0	6.7	60.0	83.3
Intermediate	31.8	72.7	90.9	4.0	68.0	88.0	22.2	66.7	66.7
Advanced	29.2	37.5	62.5	11.8	35.3	82.4	10.7	17.9	35.7
All users	35.3	67.6	83.8	8.1	66.1	90.3	12.9	48.2	62.4

Notes: Sample of API requests with erroneous code. Success rates are percentages of the requests correctly solved by ChatGPT.

Hallucination incidence

Some hallucinations in gpt 3.5-turbo

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- Stata does not allow generating binary variables directly from logical expressions like ‘fte>=20’ (...).
- (T)he ‘foreach’ loop is not properly closed with an ‘end’ command.
- Stata cannot calculate the average of variables that have missing values (...).

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- (The) ‘generate()’ option is not available with the ‘tabulate’ command in Stata (...).
- (L)ocal macro ‘depvar’ is not correctly interpolated within the ‘run _regressions’ command inside the loop (...).

Hallucination incidence by user's level and openbook mode

	Closed-book mode		Openbook mode	
	True argument	False argument	True argument	False argument
Beginner	30.7	69.3	94.7	5.3
Intermediate	62.7	37.3	81.3	18.7
Advanced	30.7	69.3	57.3	42.7
All	41.3	58.7	77.8	22.2

Notes: Sample of API requests with erroneous code. ‘True argument’ columns show percentages of requests with ChatGPT responses without hallucinations (i.e., using correct arguments). ‘False argument’ columns show percentage of hallucinations (i.e., responses that use factually wrong arguments related to Stata’s behavior).

Hallucination incidence by error type, openbook mode, model and user's level

	Typographical errors			Command errors			Syntax errors		
	Close book	Openbook		Close book	Openbook		Close book	Openbook	
	3.5-turbo	3.5-turbo	4o	3.5-turbo	3.5-turbo	4o	3.5-turbo	3.5-turbo	4o
Beginner	36.4	0.0	0.0	80.0	5.0	5.0	83.3	10.0	0.0
Intermediate	36.4	18.2	4.5	40.0	32.0	12.0	33.3	25.9	11.1
Advanced	58.3	41.7	29.2	58.8	41.2	23.5	82.1	75.0	25.0
All users	44.1	20.6	11.8	58.1	25.8	12.9	67.1	36.5	11.8

Notes: Sample of API requests with erroneous code. Cells report hallucination rates (percentages) where ChatGPT makes factually incorrect statements about Stata behavior.

Illusory expertise

Percentages of failures conditional on expertise self-report

	Do you know what the error is?		
	(Model 3.5-turbo)		(Model 4o)
	No	Yes	Yes
Beginner	16.1	37.5	6.9
Intermediate	33.3	25.0	18.9
Advanced	75.0	61.9	44.1
All users	39.9	42.1	22.9

Notes: Sample of API requests with erroneous code. The table shows failure rates (percentages) based on user level, model, and self-reported expertise. Self-reported expertise collected through an independent request. Results for Model 4o under low self-reported expertise are excluded due to small sample sizes.

Illusory expertise under difficult tasks

	Do you know what the error is?		
	(Model 3.5-turbo)	(Model 4o)	
	No	Yes	Yes
Intermediate	36.8	21.4	23.1
Advanced	78.8	66.7	46.7
All users	56.3	42.3	34.0

Notes: Sample of API requests with erroneous code from Intermediate and Advanced users who commit Command and Syntax error codes. Self-reported expertise collected through an independent request. The table shows failure rates (percentages) based on user level, model, and self-reported expertise. Results for Model 4o under low self-reported expertise are excluded due to small sample sizes.

The hammer effect

To the hammer, everything looks like a nail

- Regardless of whether the request is in openbook or closed-book mode, I prime ChatGPT to assume that the do file contains one error.

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- Regardless of whether the request is in openbook or closed-book mode, I prime ChatGPT to assume that the do file contains one error.
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- Regardless of whether the request is in openbook or closed-book mode, I prime ChatGPT to assume that the do file contains one error.
- Given the small sample size, I cannot estimate within-cell probabilities, but I can estimate a binary mode to evaluate the marginal effect of sending a conflicting instruction to ChatGPT.
- If ChatGPT has autonomous reasoning, the fact that there actually was not error in the do file should not affect the probability of appropriate answer.
 - The right answer when there is no error is something alike to “The do file can run entirely without any interruption.”
 - Of course, a script that runs without interruption is not guaranteed to render the desired results. Hence, it is a legitimate concern to offer ways to debug the script, even though it were not to stop.

Appropriate answer. Probit ML estimates.

	Full sample		Model 4o		
	Unconditional	Conditional 1	Conditional 2	Unconditional	Conditional 3
No error	-1.012*** (0.277)	-1.292*** (0.317)	-1.376*** (0.342)	-0.746* (0.407)	-0.784* (0.419)
Openbook		1.150*** (0.133)	1.190*** (0.143)		
gpt 4o		0.578*** (0.129)	0.674*** (0.181)		
Command/Syntax		-0.588*** (0.117)	-0.585*** (0.118)		-0.951*** (0.239)
Knows answer			-0.141 (0.186)		
N. obs.	675	675	675	225	225
AME(No error)	-0.397 (0.105) [0.000]	-0.390 (0.092) [0.000]	-0.415 (0.099) [0.000]	-0.228 (0.122) [0.062]	-0.222 (0.116) [0.055]

Notes: *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$. Standard errors in parenthesis and p -values in square brackets. Probit Maximum Likelihood estimates. Full sample includes all 625 requestes. Output variable is a dummy binary for appropriate answer. Variable 'No error' is a dummy variable that takes value 1 if the request involves a script that has no errors.AME(No error) is the estimated Average Marginal Effect of 'No error'. 'No error' predicts failure perfectly in the requests to gpt 3.5-turbo.

Conclusions

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- **Solution effectiveness:** Improvements from 3.5-turbo to 4o
 - Still, large failure rates in complex do files ($\approx 50\%$) even under open book.
- Openbook model drastically reduces **Hallucination incidence**.
 - Specially true for gpt 4o. Percentage of requests using arguments that invoke erroneous Stata behavior.
- **Illusory expertise** is still a problem in 4o, especially under advanced programming.

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- **Illusory expertise** is still a problem in 4o, especially under advanced programming.
- **Hammer effect:** Feeding ChatGPT with strict guidelines that challenge the openbook facts negatively influences solution effectiveness.

Number of observations by user level and error type

	None	Typographical	Command	Syntax	All requests
Beginner	9	66	60	90	225
Intermediate	3	66	75	81	225
Advanced	18	72	51	84	225
All users	30	204	186	255	675

Notes: Sample of API requests. The table shows the number of API requests categorized by user proficiency level ('Beginner', 'Intermediate', 'Advanced') and error type ('None': no errors, 'Typographical': minor errors in text or formatting, 'Command': usage of non-existent commands or options, and 'Syntax': structural programming errors).

► back

The Beginner user do file

```
*Untitled Document 1
1 clear
2 cd "/home/ricardo/AAPAPERS/Chat/Stata_GPT_4/"
3 log using "output/beginner.log", replace
4 import delimited "data/CardKrueger2.csv", clear case(preserve)
5 sort id month
6 merge 1:1 id month using "data/CardKrueger1.dta"
7 drop if _merge != 3
8 summarize fte month state Post Treated
9 gen Post_Treated_Control = Post * Treated
10 egen mean_fte = mean(fte)
11 replace fte = mean_fte if missing(fte)
12 tabulate month, m
13 save "data/CardKrueger_merged.dta", replace
14 regress fte Post Treated Post_Treated, robust
15 gen Dfte20 = fte >= 20
16 logit Dfte20 Post Treated Post_Treated
17 histogram fte, bin(50)
18 log close
19
```

► back

The Intermediate user do file

```

1 clear
2 capture log close
3 cd "/home/ricardo/AAPAPERS/Chat/Stata_GPT_4/"
4 log using "output/intermediate.log", replace
5 use "data/ColombiaDHS.dta", clear
6 keep year_int region nonsevere_violence severe_violence sexual_violence age num_children age_1child educ_yr father_mother_violence
7 local depvars nonsevere_violence severe_violence sexual_violence
8 local controls age num_children age_1child
9 foreach var of local controls {
10     gen squared_`var' = `var'^2
11 }
12 preserve
13 collapse (mean) `controls', by(year_int)
14 sort year_int
15 list year_int `controls'
16 codebook `controls'
17 restore
18 tab year_int, generate(Dtime)
19 tab region, generate(Dregion)
20 qui foreach depvar of local depvars {
21     regress `depvar' educ_yr, robust
22     estimates store Uncond
23     regress `depvar' educ_yr `controls' squared_age, robust
24     estimates store Cond
25     regress `depvar' educ_yr `controls' squared_age Dtime* Dregion*, robust
26     estimates store TimeRegion
27     ivregress 2sls `depvar' (educ_yr = father_mother_violence) `controls' squared_age Dtime* Dregion*, robust
28     estimates store IV
29     noi dis newline as txt "Dependent variable: " in y "`depvar'"
30     noi estimates table Uncond Cond TimeRegion IV, b(%7.4f) keep(educ_yr `controls' squared_age) stats(N r2_a) star
31 }
32 log close
33

```

► back

The Advanced user do file

```

1 clear
2 capture log close
3 capture program drop run_regressions
4 program define run_regressions
5 syntax , DEP(varname) EXOG(varlist) ENDOG(varname) INSTR(varlist)
6 qui {
7   gettoken z1 z : exog
8   regress `dep' `endog', robust
9   estimates store Uncond
10  regress `dep' `endog' `exog' c.`z1'#c.`z1', robust
11  estimates store Cond
12  regress `dep' `endog' `exog' c.`z1'#c.`z1' i.year_int i.region, robust
13  estimates store TimeRegion
14  ivregress 2sls `dep' (`endog' = `instr') `exog' c.`z1'#c.`z1' i.year_int i.region, robust
15  estimates store IV
16  noi dis newline as txt "Dependent variable: " in y "`dep'"
17  noi estimates table Uncond Cond TimeRegion IV, b(%7.4f) drop(i.year_int i.region) stats(N r2_a) star
18 }
19 end
20 cd "/home/ricardo/AAPAPERS/Chat/Stata GPT_4/"
21 log using "output/advanced.log", replace
22 use "data/ColombiaDHS.dta", clear
23 keep year int region nonsevere_violence severe_violence sexual_violence age num_children age_1child ///
24     educ yr father_mother_violence
25 local depvars nonsevere_violence severe_violence sexual_violence
26 local controls age num_children age_1child
27 qui foreach depvar of local depvars {
28   noi run_regressions, dep(`depvar') exog(`controls') endog(educ_yr) instr(father_mother_violence)
29 }
30 log close
31

```

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Typographical errors

<pre> 1_Beginner00.do 1 clear 2 capture log close 3 cd "/home/ricardo/AAPAPERS/Chat/Stata GPT_4/" 4 log using "output/beginner.log", replace 5 import delimited "data/CardKrueger2.csv", clear case(preserve) 6 sort id month 7 merge 1:1 id month using "data/CardKrueger1.dta" 8 drop if _merge != 3 9 keep id fte month state Post Treated 10 summarize fte month state Post Treated 11 gen Post_Treated = Post * Treated 12 egen mean_fte = mean(fte) 13 replace fte = mean_fte if missing(fte) 14 tabulate month, m 15 save "data/CardKrueger merged.dta", replace 16 regress fte Post Treated Post_Treated, robust 17 gen Dfte20 = fte>=20 18 logit Dfte20 Post Treated Post_Treated 19 histogram fte, bin(50) 20 log close 21 </pre>	<pre> 1_Beginner73.do 1 clear 2 capture log close 3 cd "/home/ricardo/AAPAPERS/Chat/Stata GPT_4/" 4 log using "output/beginner.log", replace 5 import delimited "data/CardKrueger2.csv", clear case(preserve) 6 sort id month 7 merge 1:1 id month using "data/CardKrueger1.dta" 8 drop if _merge != 3 9 keep id fte month state Post Treated 10 summarize fte month state Post Treated 11 gen Post_Treated = Post * Treated 12 egen mean_fte = mean(fte) 13 replace fte = mean_fte if missing(fte) 14 tabulate month, m 15 save "data/CardKrueger merged.dta", replace 16 regress fte Post Teated Post_Treated, robust 17 gen Dfte20 = fte>=20 18 logit Dfte20 Post Teated Post_Treated 19 histogram fte, b(50) 20 log close 21 </pre>
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Command/options conflicts

1_Beginner00.do	1_Beginner50.do
1 clear	1 clear
2 capture log close	2 capture log close
3 cd "/home/ricardo/AAPAPERS/Chat/Stata_GPT_4/"	3 cd "/home/ricardo/AAPAPERS/Chat/Stata_GPT_4/"
4 log using "output/beginner.log", replace	4 log using "output/beginner.log", replace
5 import delimited "data/CardKrueger2.csv", clear case(preserve)	5 import delimited "data/CardKrueger2.csv", clear case(preserve)
6 sort id month	6 sort id month
7 merge 1:1 id month using "data/CardKrueger1.dta"	7 merge 1:1 id month using "data/CardKrueger1.dta"
8 drop if _merge != 3	8 drop if _merge != 3
9 summarize fte month state Post Treated	9 summarize fte month state Post Treated
10 gen Post_Treated = Post * Treated	10 gen Post_Treated = Post * Treated
11 egen mean_fte = mean(fte)	11 egen mean_fte = mean(fte)
12 replace fte = mean_fte if missing(fte)	12 replace fte = mean_fte if missing(fte)
13 tabulate month, m	13 tabulate month, m
14 save "data/CardKrueger_merged.dta", replace	14 save "data/CardKrueger_merged.dta", replace
15 regress fte Post_Treated Post_Treated, robust	15 regress fte Post_Treated Post_Treated, robust
16 gen Dfte20 = fte>=20	16 gen Dfte20 = fte>=20
17 logit Dfte20 Post_Treated Post_Treated	17 logit Dfte20 Post_Treated Post_Treated
18 histogram fte, bin(50)	18 histogram fte, b(75)
19 log close	19 close log
20	20

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Syntax issues

```
1 clear
2 cd "/home/ricardo/AAPAPERS/Chat/Stata GPT 4/"
3 log using "output/beginner.log", replace
4 import delimited "data/CardKrueger2.csv", clear case(preserve)
5 sort id month
6 merge 1:1 id month using "data/CardKrueger1.dta"
7 drop if _merge != 3
8 summarize fte month state Post Treated
9 gen Post_Treated_Control = Post * Treated
10 egen mean_fte = mean(fte)
11 replace fte = mean_fte if missing(fte)
12 tabulate month, m
13 save "data/CardKrueger_merged.dta", replace
14 regress fte Post Treated Post_Treated, robust
15 gen Dfte20 = fte>=20
16 logit Dfte20 Post Treated Post_Treated
17 histogram fte, bin(50)
18 log close
19
```

```
1 clear
2 cd "/home/ricardo/AAPAPERS/Chat/Stata GPT 4/"
3 import delimited "data/CardKrueger2.csv", clear case(preserve)
4 sort id month
5 merge 1:1 id month using "data/CardKrueger1.dta"
6 drop if _merge != 3
7 summarize fte month state Post Treated
8 gen Post_Treated_Control = Post * Treated
9 egen mean_fte = mean(fte)
10 replace fte = mean_fte if missing(fte)
11 tabulate month, m
12 save "data/CardKrueger_merged.dta", replace
13 regress fte Post Treated Post_Treated, robust
14 gen Dfte20 = fte>=20
15 logit Dfte20 Post Treated Post_Treated
16 histogram fte, bin(50)
17 log close
18
```

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