

# LLM-Supported Natural Language to Bash Translation (NL2SH)

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### Motivation

Bash commands are <u>complex</u> and <u>difficult to memorize</u>. Incorrect commands can cause <u>system failures</u>. Large Language Models (LLMs) are unreliable for translating natural language to Bash (NL2SH). Improving NL2SH requires (1) clean data, (2) reliable benchmarks, and (3) accurate translation methods.

## Dataset

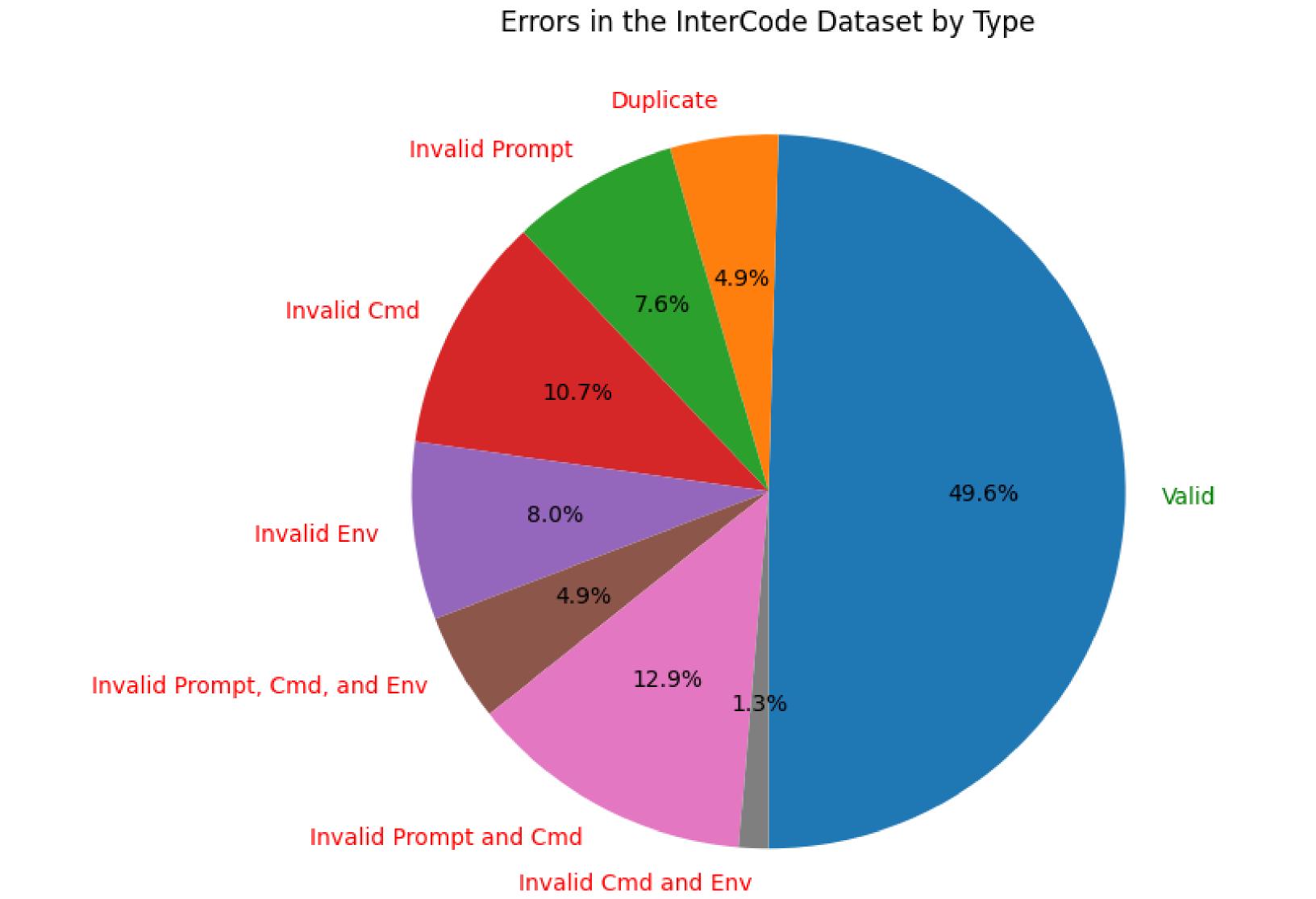
#### Goal:

Given		Determine
Natural Language	Bash	Correct
print the system disk usage	df -h	True
remove a directory named foo	rm foo	False
print the current user's id	id -u	True

Challenge: Datasets contain invalid translations.

Research Question: How can we validate NL2SH datasets to ensure models are evaluated using accurate assessments?

Result: Manual verification of test data improves benchmark reliability. Cleaning and increasing training data improves in-weight learning.



#### Benchmark

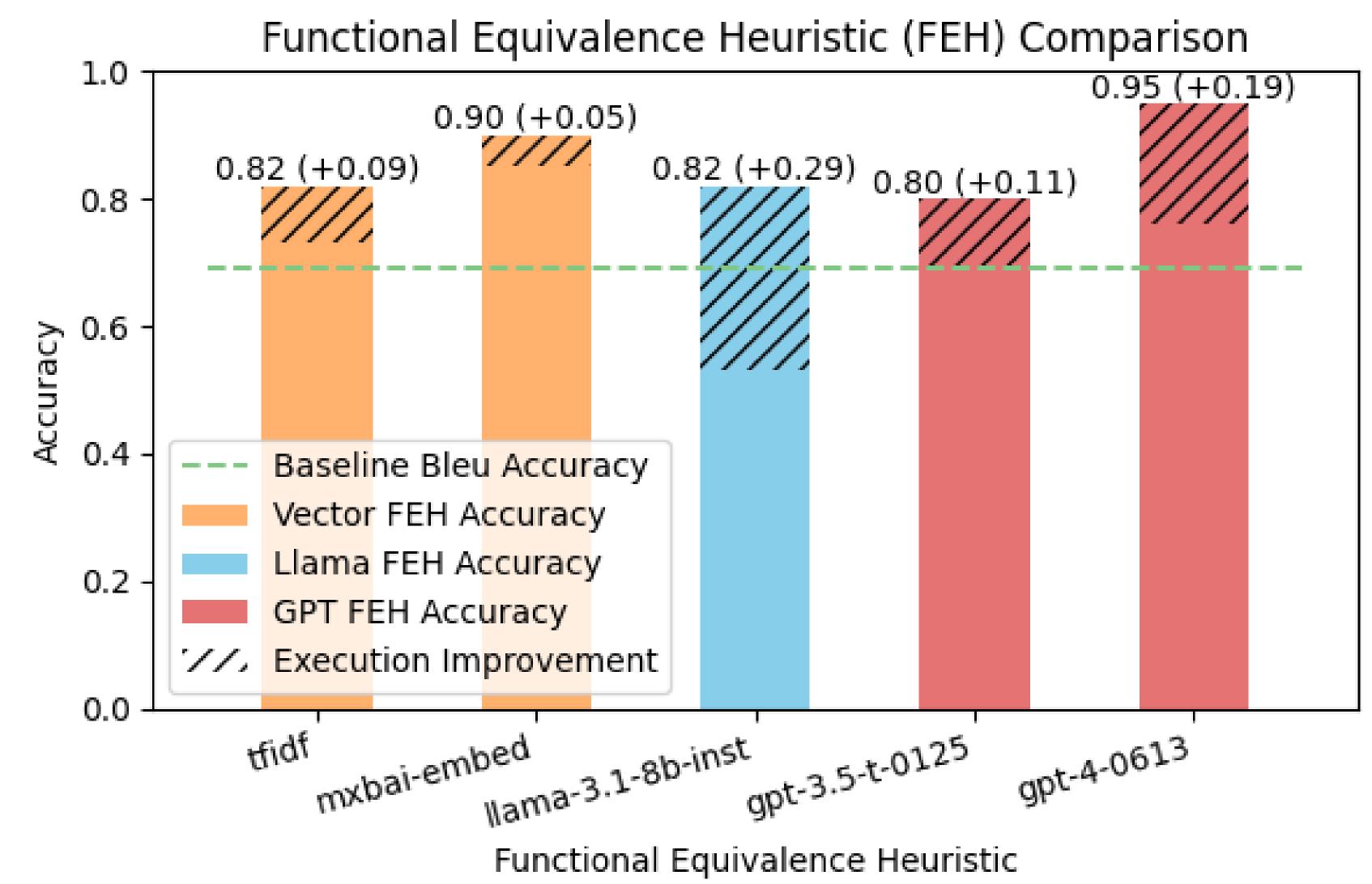
#### Goal:

Given			Determine
Natural Language	GT Command	Model Output	Equivalent
print the system boot time	who -b	uptime -s	True
delete bin\ in the current dir	rm -r ./bin	rm -r /bin	False
list all groups on the system	getent group	cat /etc/group	True

Challenge: Benchmarks cannot determine the functional equivalence of Bash commands.

Research Question: How can we design a functional equivalence heuristic that accurately measures the quality of model translations?

Result: Command execution with language model evaluation of command outputs outperforms previous benchmarking methods.



## **Translation Method**

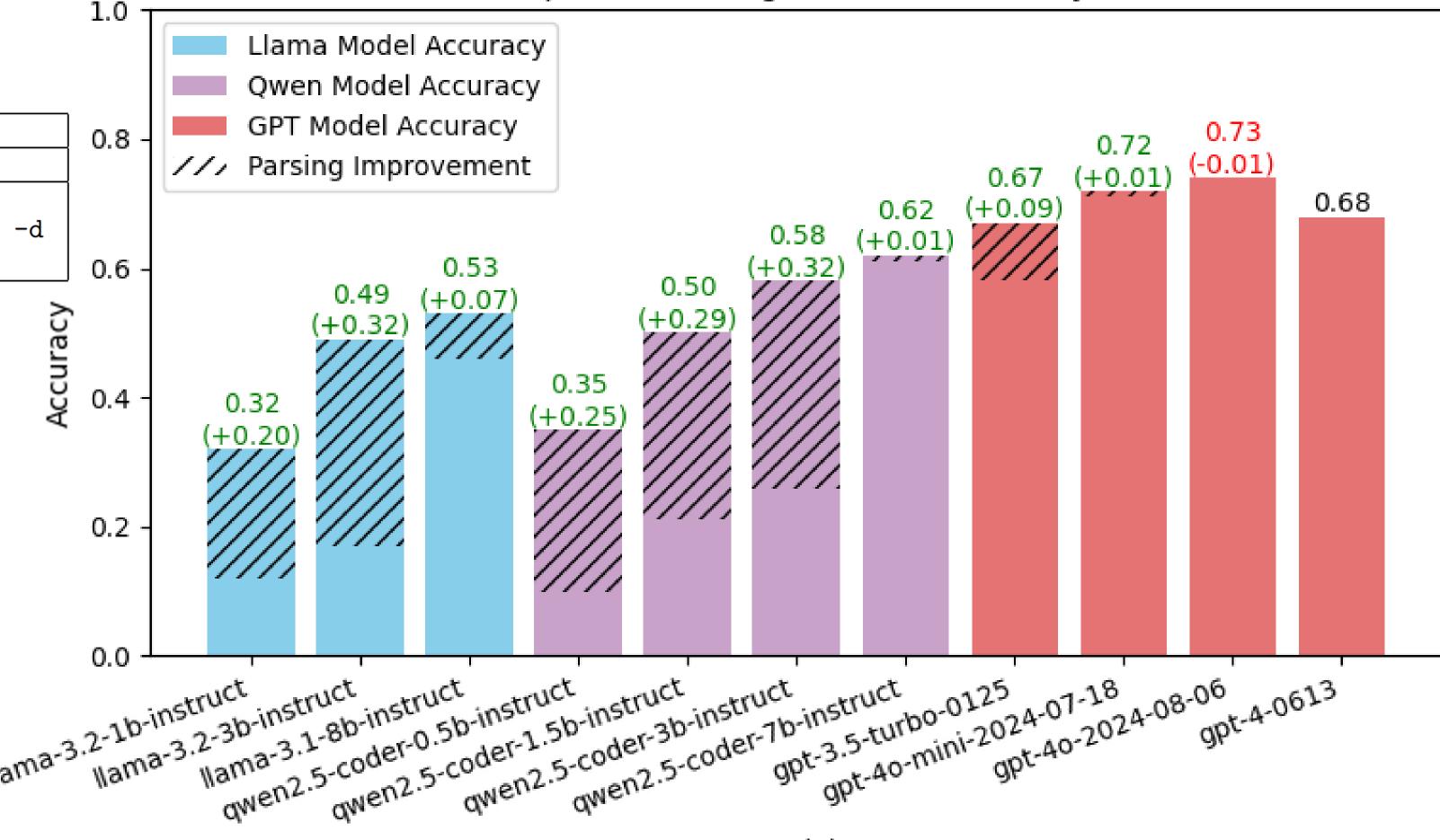
#### Goal:

Given	Determine	
Natural Language	Bash Command	(
List files in /workspace accessed over an hour ago base64 decode aGVsbG8 Sort and print only group names from /etc/group	find /workspace -type f -amin +60 echo 'aGVsbG8='   openssl enc -base64 -d cut -d: -f1 /etc/group   sort	
	>	`

Challenge: LLMs have poor NL2SH performance.

Research Question: How can we improve the accuracy of NL2SH models as measured by a reliable benchmark?

Result: Parsing, in-context learning, in-weight learning and constrained decoding can improve NL2SH accuracy by up to 32%.



Model

Impact of Parsing on Model Accuracy





