

# Neuro-Symbolic Compliance: Integrating LLMs and SMT Solver for Automated Financial Legal Analysis

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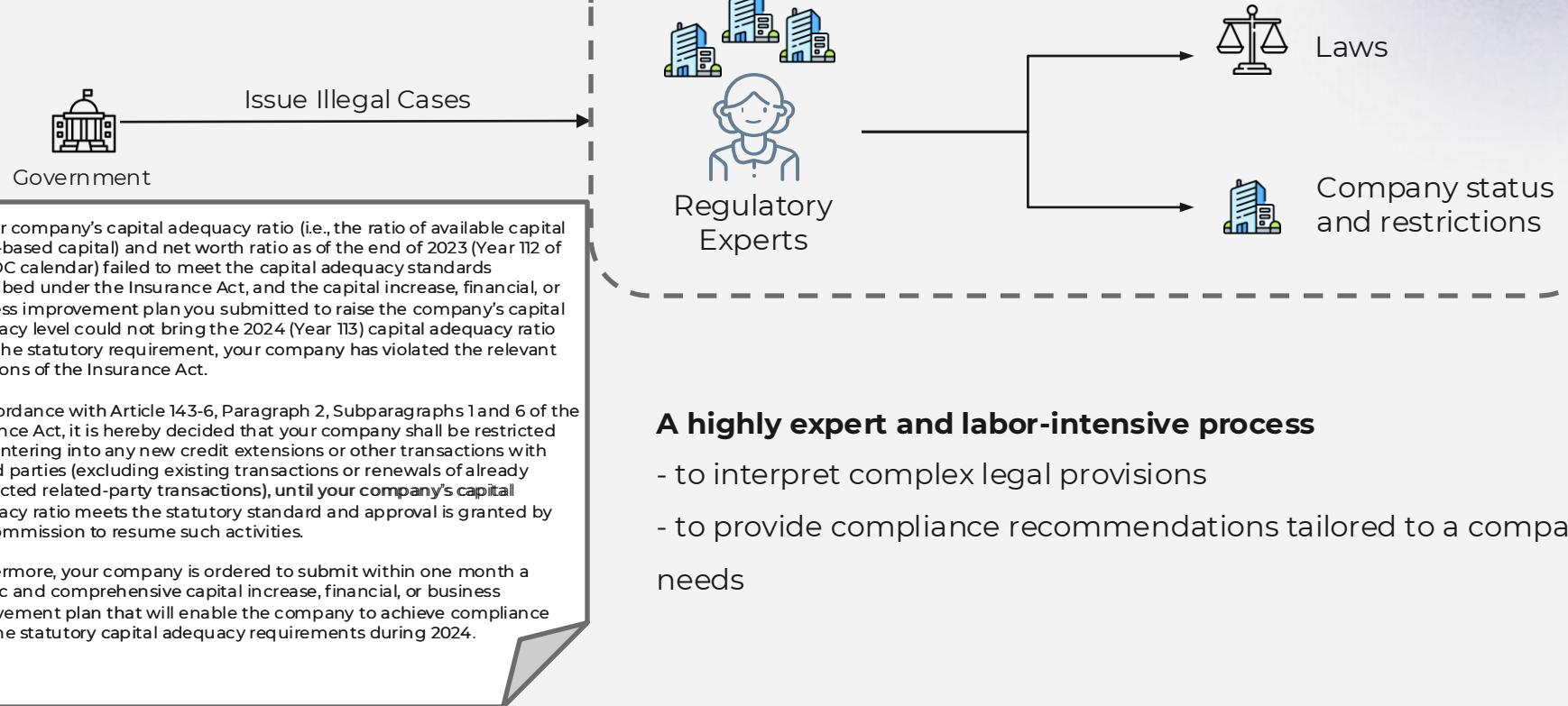
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# Financial Compliance Analysis



# Financial Compliance Analysis with the Aid of LLMs

The appearance of LLMs has made understanding text much faster, and the use of RAG modules can further enhance an LLM's comprehension across different domains.



Regulatory Expert

Base on the illegal case and its related laws, please suggest how to restore legality.

**Article § 143-6 violation: must be done together with raising the capital adequacy ratio**



LLM

You can lift the credit restriction with related parties and ensure the improvement plan is executed to achieve a penalty-free compliant status.



Regulatory Expert

If we can't adjust the status of improvement plan, is there any solution that allows us to make only minimal changes?

**Article § 143-6 violation : only required at Capital Level 4, but this case is at Level 3**

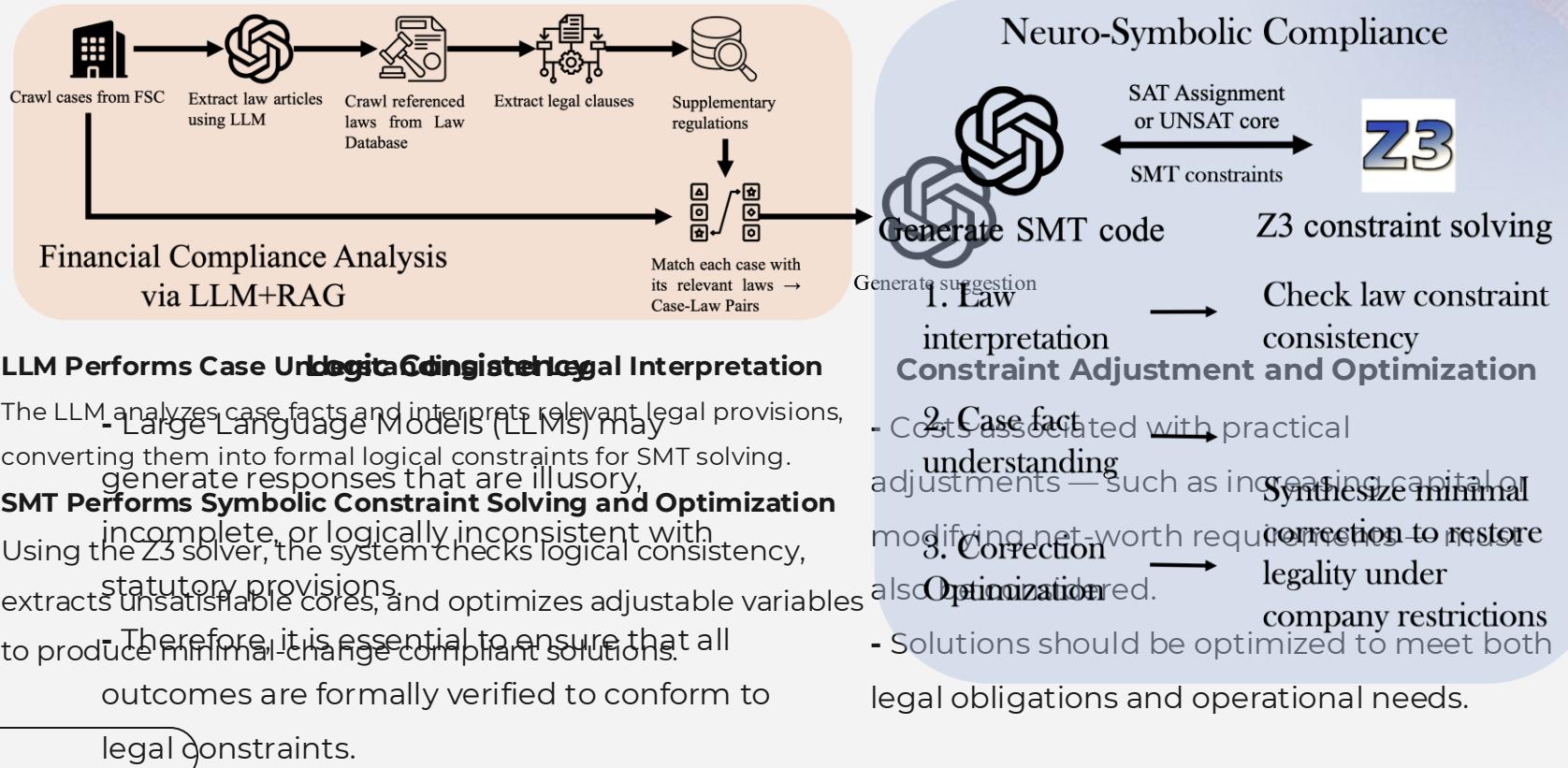


LLM

If the improvement plan status cannot be adjusted, you should accelerate approval for asset disposal and initiate limited capital reinforcement to strengthen its financial position.

**LLMs read fast — but can they reason right?**

# Our Method : Neuro-Symbolic Compliance

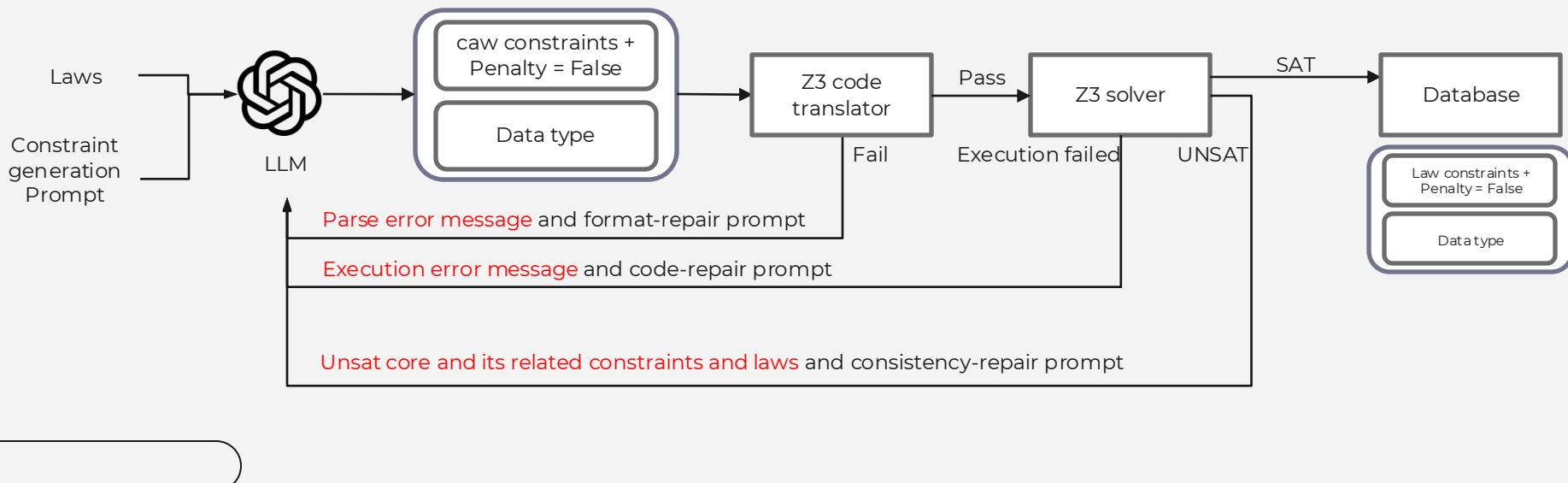


# Law Interpretation and Legality Consistency Checking

The LLM formalizes legal provisions into logical constraints, while Z3 validates their correctness.

Parsing errors, execution failures, and unsat cores are used to refine the constraints.

Only solver-verified constraints are stored as legally consistent representations.

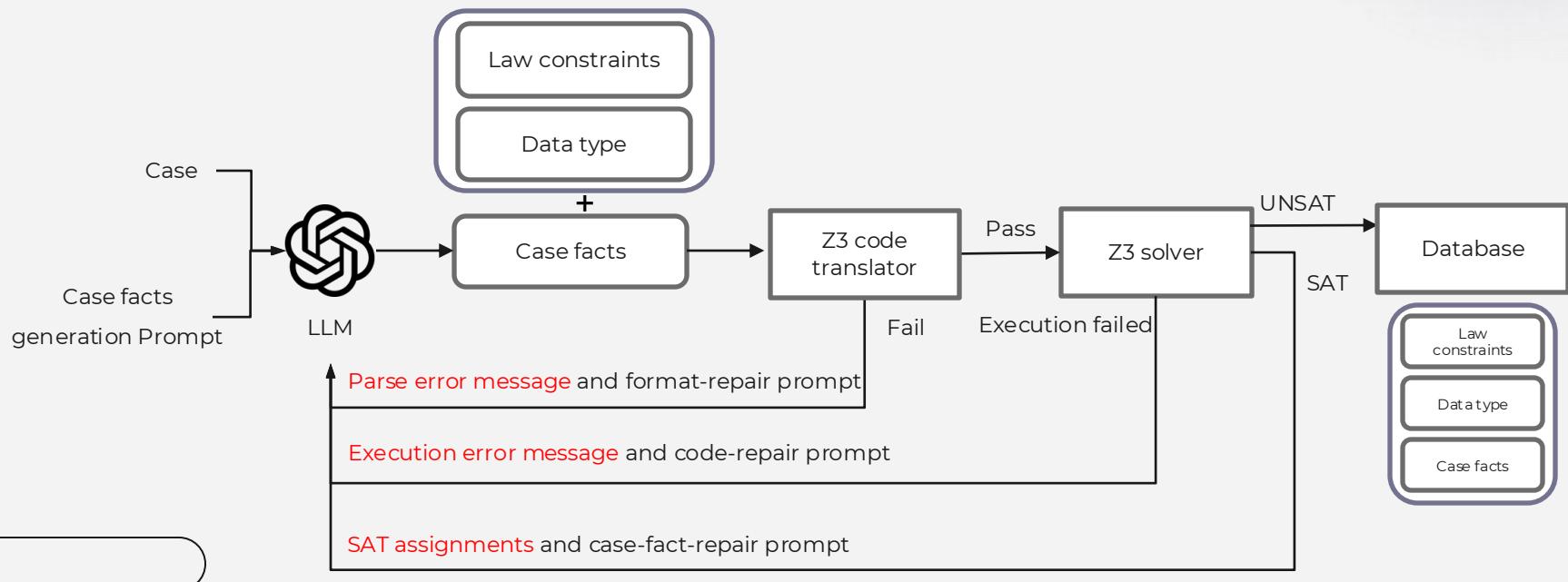


# Case Fact Understanding and Illegality Consistency Checking

The LLM generates case fact assignments, which are translated and checked by the Z3 solver.

Any parsing, execution, or inconsistency errors trigger automated repair.

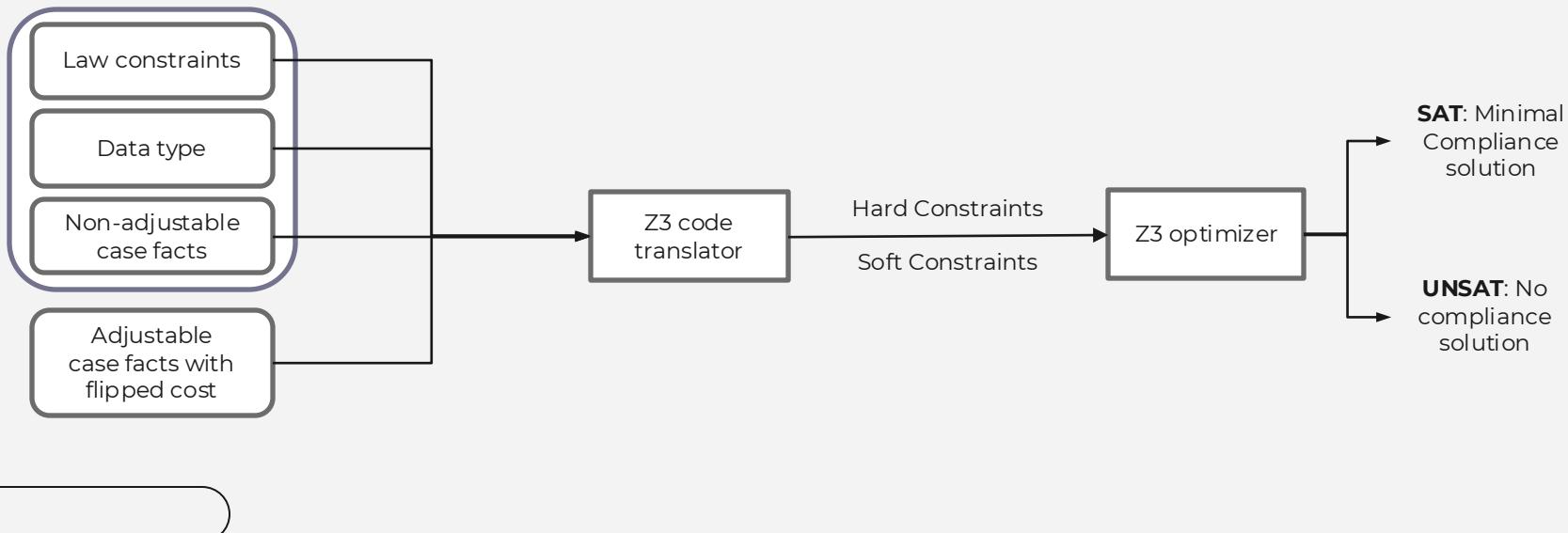
For violation cases, the solver returns **UNSAT**, and the case facts are stored.



# Minimal Compliance Solving using Z3 Optimizer (MaxSMT)

Once law constraints and case facts are generated and validated,  
the Z3 Optimizer is used to compute the minimal compliant adjustment.

Non-adjustable facts become hard constraints, while adjustable facts become weighted soft constraints  
for prioritized optimization.



# RQ1: To what extent can LLMs effectively generate SMT constraints in the legal domain?

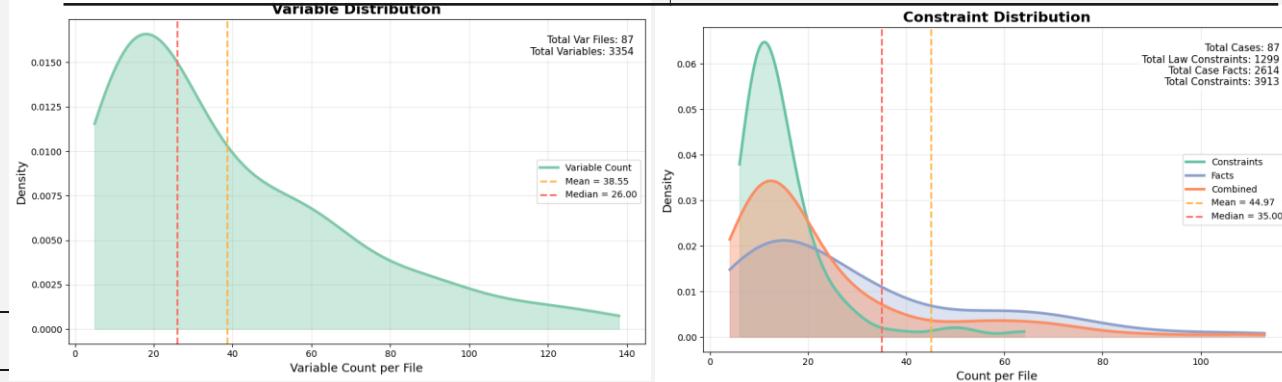
We evaluate our pipeline against 87 real-world regulatory cases from the Taiwan FSC.

Neuro-Symbolic Compliance		Pass	Repair (Needed/Success)	Pass rate(%)
<b>Law Interpretation</b>	Format-repair	87	0/0	100.00
	Code-repair	83	4/4	100.00
	Consistency-repair	87	0/0	100.00
<b>Case Fact Understanding</b>	Format-repair	87	0/0	100.00
	Code-repair	87	0/0	100.00
	Case-fact-repair	87	0/0	100.00
<b>Cost-aware Compliance</b>	Z3-MaxSAT Optimization	87	0/0	100.00

# RQ1: To what extent can LLMs effectively generate SMT constraints in the legal domain?

To further analyze task complexity, we examined the structural diversity of constraints and variables, confirming that the constraint generation task is inherently non-trivial.

	#Var	Real	Int	Bool	#Con	Law	Case	Law Ratio(%)
<b>Avg</b>	<b>38.55</b>	1.23	0.92	36.40	<b>44.97</b>	14.93	30.04	38.61
<b>Min</b>	5	0	0	5	12	6	4	13.33
<b>Max</b>	138	15	11	138	139	64	113	86.36
<b>Std.</b>	30.85	2.60	1.92	29.63	29.99	9.64	24.83	15.90

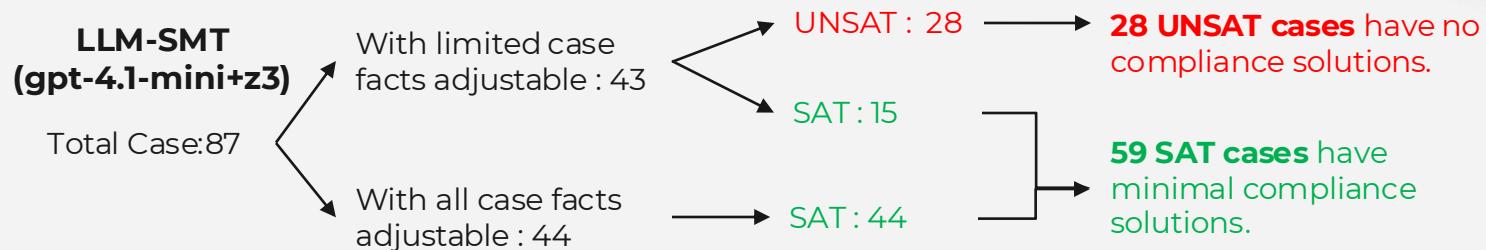


# RQ2: Does LLM-SMT Integration Outperform Standalone LLMs in Finding Legally Compliant Solutions?

For each case, we provide LLM-SMT / Pure LLM

(1) the case description, (2) relevant legal articles, (3) adjustable case facts with flipped cost

The model is required to check whether a feasible (SAT) solution (with minimal cost) exists.



Pure LLM  
(gpt-4.1-mini)

	59 SAT cases	28 UNSAT cases
Have correction	59 (8 / 51)	19
No suggestions	0	9

Accuracy rate drops to 21.8%.  
(8+9 / 87)

# RQ3 Does LLM-SMT Integration Better Identify Minimal Compliant Solutions Than LLMs Alone?

For 8 SAT cases that LLM returns valid suggestions, its suggestion on average increases 33% cost compared to solutions from LLM-SMT (5.25 → 7).

LLM-SMT provides the optimal correction with 30.41% (42.08%) flipped rate for limited (all) adjustable cases.

Cost(#Flips)	LLM (8)	LLM-SMT (8)
<b>Mean</b>	7.00	5.25
<b>Median</b>	4.5	3
<b>Min</b>	2	1
<b>Max</b>	20	15
<b>Std. Dev.</b>	6.78	5.55
<b>Avg. Flip rate</b>	53.23%	42.30%

Cost (#Flips)	Limited case facts LLM-SMT (15)	All case facts LLM-SMT (44)
<b>Mean</b>	4.60	11.14
<b>Median</b>	3	8
<b>Min</b>	1	1
<b>Max</b>	15	56
<b>Std. dev.</b>	3.83	11.45
<b>Avg. flip rate</b>	30.41%	42.08%

# Conclusion & Extension

We propose a Neuro-Symbolic Compliance framework that integrates LLMs and SMT optimization for automated legal reasoning, providing efficient, cost-aware compliance analysis.

Extension:

- Agentic pipeline and platform deployment
- Public available dataset with extensive case study



Demo



Artifacts



Full Paper

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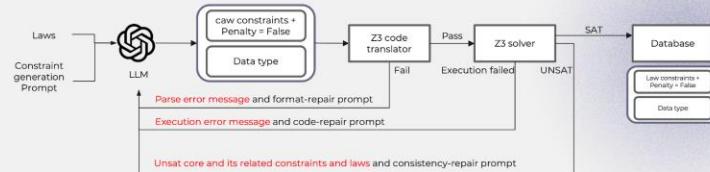


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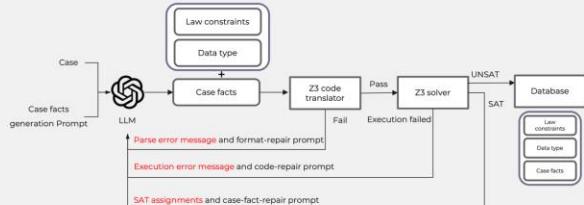
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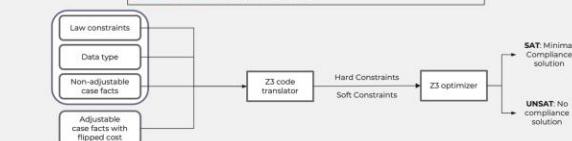
Phase 1:  
Law  
constraints



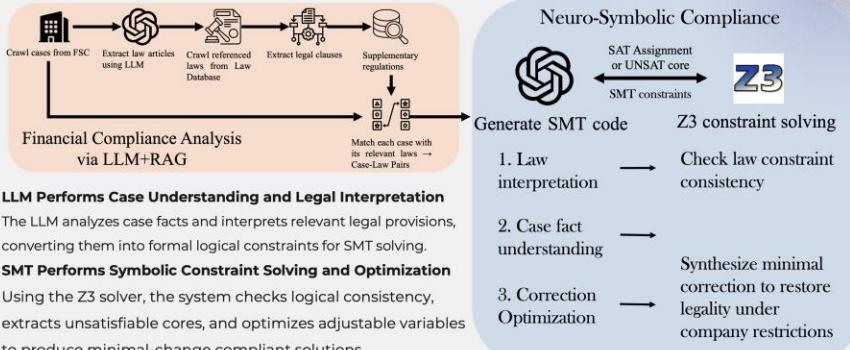
Phase 2:  
Case-facts  
assignment



Phase 3:  
Cost-aware  
correction  
solving



## Our Method : Neuro-Symbolic Compliance



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