





Kunerva+: An Intelligent Security Policy Generation Framework for Containers

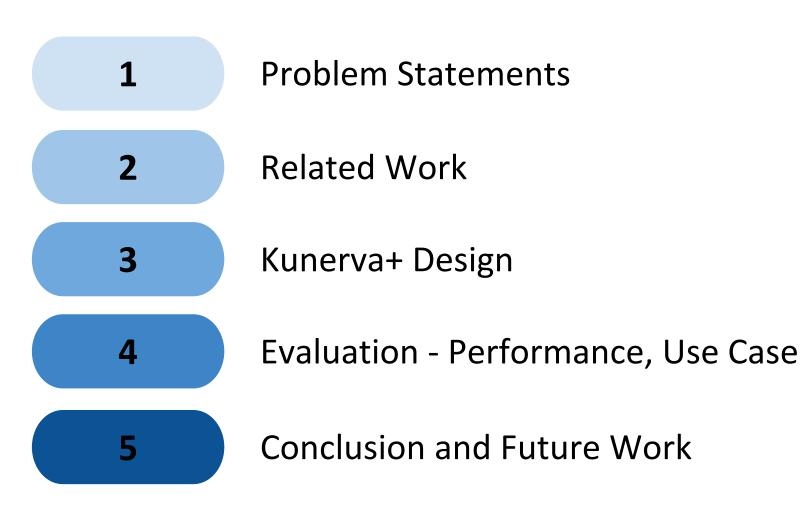
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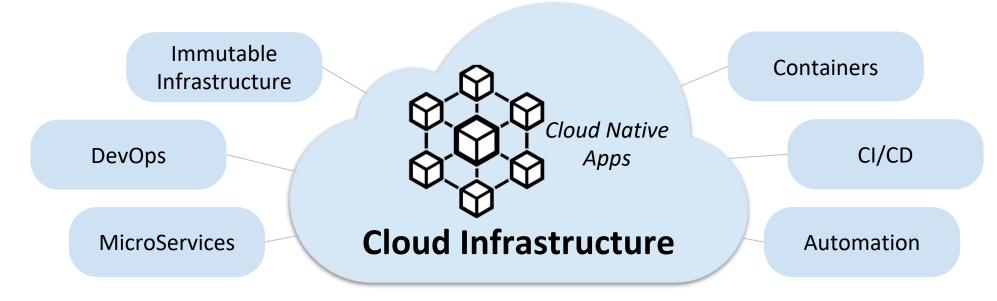
OUTLINE



PROBLEM STATEMENTS

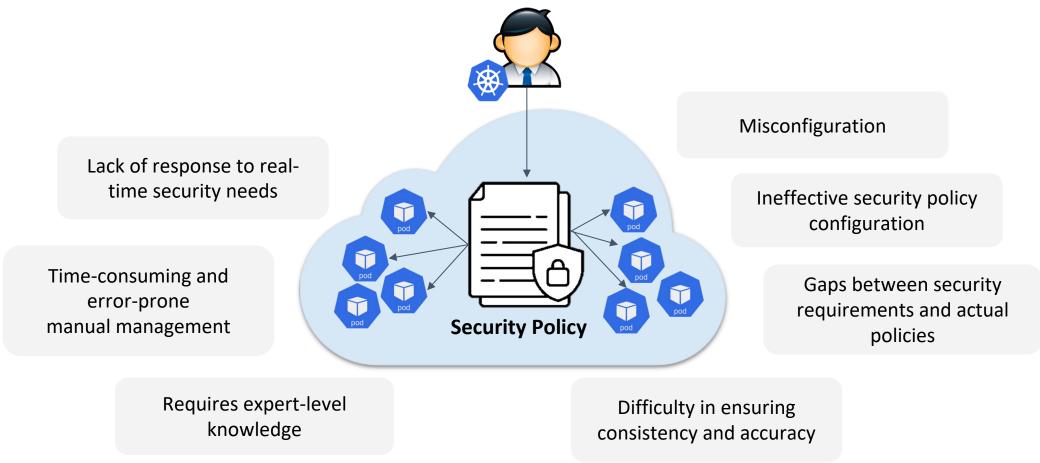
• Increased complexity of security management

- Frequent Creation and Deletion of Containers
- Dynamic Network Configuration Changes
- Increased risk of security vulnerabilities



PROBLEM STATEMENTS

• Limitations of Passive Security Policy Management





RELATED WORK

- Jacobs et al. (LUMI) [1] proposed using natural language to express network management intent with Nile.
- But this approach utilizes an intermediate form of policy, rather than using natural language verbatim, and is primarily focused on network configuration.

- Li et al. (AUTOARMOR) [2] developed an automatic policy generation method for interservice access control in microservices.
- Although this method effectively handles static analysis and policy updates, it is not comprehensive enough to integrate both network and system security policies in cloud-native environments.

[1] Jacobs, Arthur S., et al. "Hey, lumi! using natural language for {intent-based} network management." 2021 USENIX Annual Technical Conference (USENIX ATC 21). 2021.
[2] Li, Xing, et al. "Automatic policy generation for {Inter-Service} access control of microservices."30th USENIX Security Symposium (USENIX Security 21). 2021.

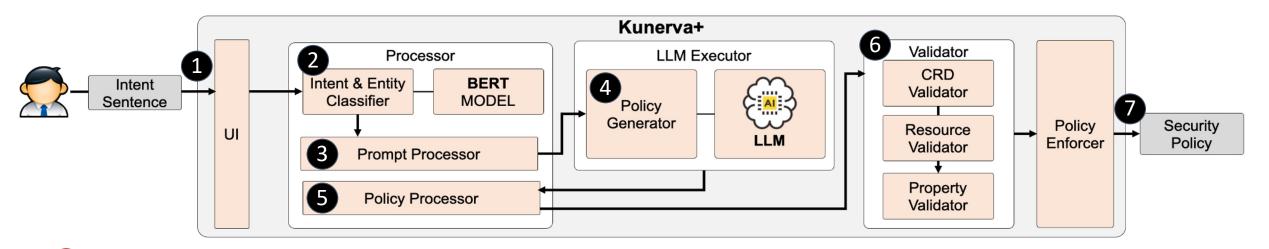


Kunerva+ DESIGN: Architecture

 Aim to enhance the efficiency of security management in cloud-native by automatically generating and validating network and system security policies based on natural language input.

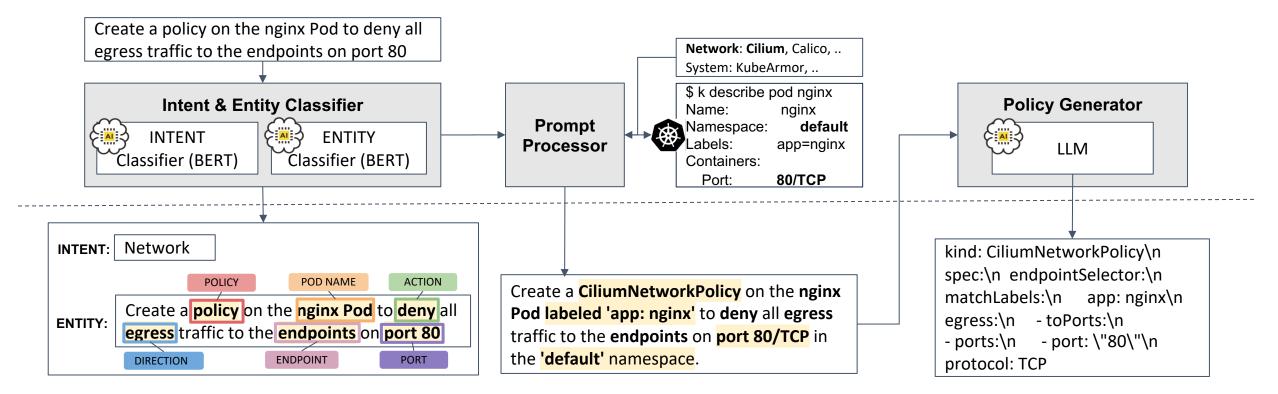
1. <u>Accurate interpretation of natural language input</u>: Be able to accurately analyze natural language input from users and translate it into security policies.</u>

2. <u>Validate and enforce automated policies</u>: Automate the process of validating and enforcing the generated security policies.



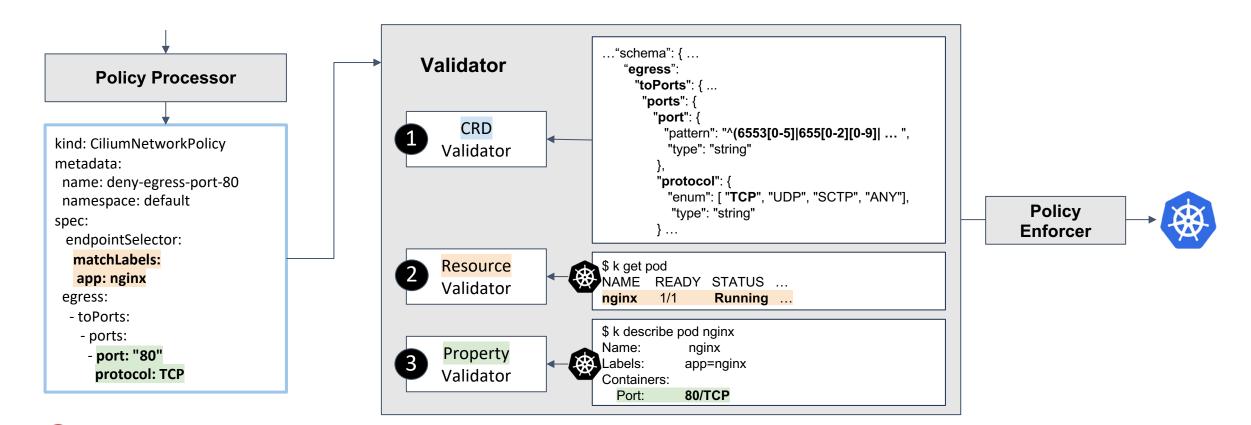
Kunerva+ DESIGN: Policy Creation

- Use BERT-based classifiers to identify intents and entities from user input.
- Transform the extracted information into detailed prompts for the policy generator.



Kunerva+ DESIGN: Policy Validation and Enforcement

- Verify CRD syntax, resource existence, and property conditions.
- Ensure that the policy is correctly configured and applicable.



Kunerva+ DESIGN: Select Dataset and Model

- Scraped policy files *from GitHub*, modified fields, and generated instructions.
- Choose an *open text-to-text model* for policy creation:
 - 1. Models with small size and low parameter count for high accuracy
 - 2. Models with above-average policy performance for a wide range of generation requirements

Dataset Name	Туре	Size		Model Name	Size (params)	Purpose
Network Policy	JSON Lines	166,064개	187.3 MB	bert-intent-classification	425 MB	Intent Classification
System Policy		2,914개	3.27 MB	bert-ner-classification	425 MB	NER Classification
Network Policy	CSV	1,500개	4.10 MB (3,0007∦)	Meta/Meta-Llama-3-8B-Instruct	16GB (8.03B)	Policy Generation
System Policy		1,500개		DeepSeek/deepseek-coder-7b-instruct-v1.5	14GB (6.91B)	Policy Generation
Table 1. Summary of Datasets				MistralAI/Mistral-7B-Instruct-v0.2	15GB (7.24B)	Policy Generation
				Google/codegemma-7b-it	17GB (8.54B)	Policy Generation
				Meta/codeLlama-7b-Instruct-hf	14GB (6.74B)	Policy Generation
				Table 2. Summary of Model Features		



EVALUATION: Performance of models (BLEU)

• On average, the fine-tuned model outperformed the baseline model by approximately 27%.

33.54 33.74

codeLlama

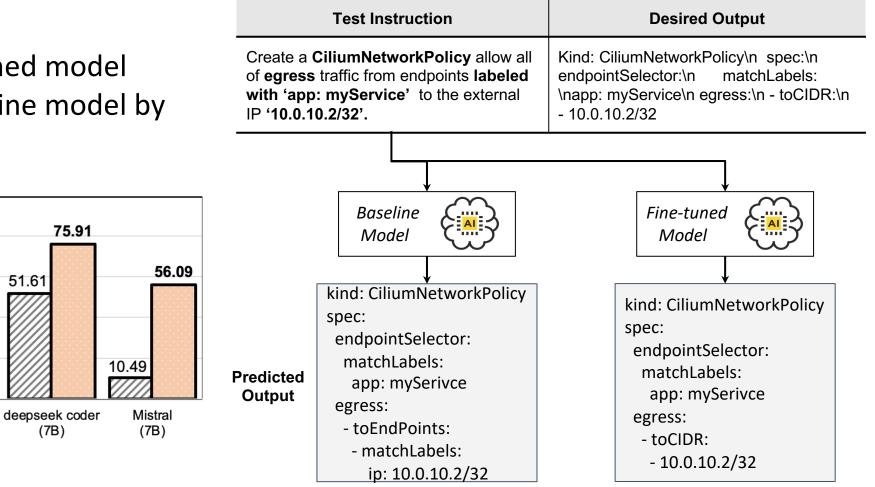
(7B)

51.61

(7B)

72.23

52.83



codegemma

(7B)

Baseline Model

Finetuned Model

47.98

80

60

20

Ο

19.45

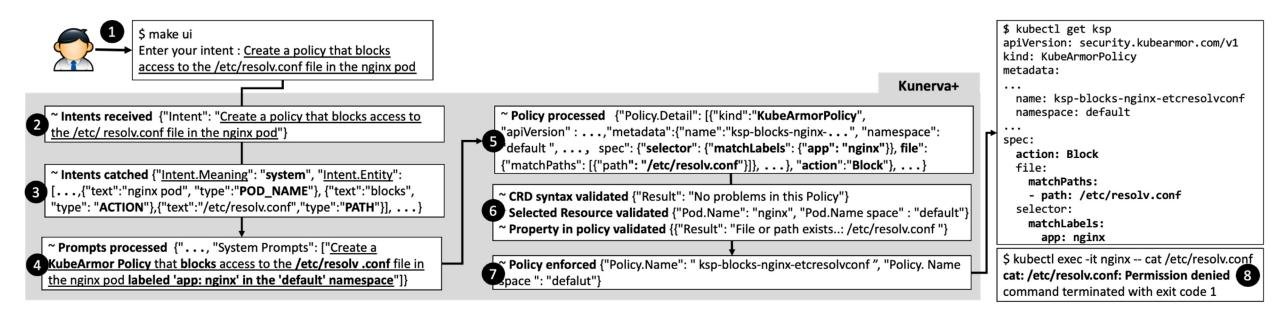
Meta Llama3

(8B)

Score(%)

EVALUATION: Use Case

- **Example:** <u>Blocking access to the /etc/resolv.conf file in the nginx pod.</u>
- Demonstrated policy creation, validation, and enforcement process to show the ability of Kunerva+ to effectively manage security policies in real-world scenarios.





CONCLUSION AND FUTURE WORK

- Provide an *intelligent security policy generation framework* to reduce complexity and human error in containerized environments.
- Propose an *automated policy validation and enforcement mechanism* to ensure reliability and accuracy in dynamic cloud-native environments.
- Demonstrate the *practical utility of AI in security management by using fine-tuned LLMs* to effectively translate natural language input into accurate security policies.
- Future work <u>focuses on automatically inferring optimal security policies based solely</u> <u>on resource configuration files</u> to further reduce user input while improving the system's adaptability to security needs.





YOU