

# NFN – Recent Work on Expression Forwarding

**Christopher Scherb · Claudio Marxer · Christian Tschudin**

*University of Basel, Switzerland*

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# NFN Mindset

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ICN generalization: Deliver cooked results instead of raw data

## Idea:

- User defines computation / workflow
- Network finds execution location
- Similar to Serverless Computing, but offers workflow orchestration
- Implemented using Interest / Named Data Object Transport Layer

# PiCN - pure Python Implementation of NFN

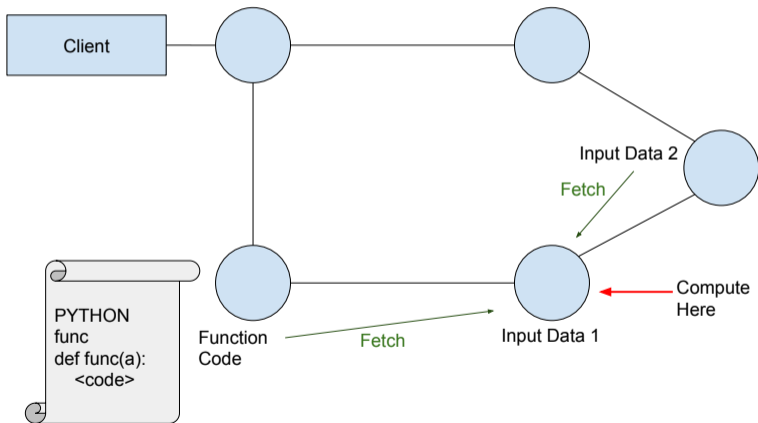
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- code available at: <https://github.com/cn-uofbasel/PiCN>
- includes an ICN Forwarder
- executes Python code under network control
  - decomposes complex expressions “in the network”
  - optimizes by migrating either code and/or data  
(more about this in this talk)
- since recently also for native Intel x86 code
- previous Scala (JVM) support could be added again

# Example

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Request: <lambda expression>



## PiCN (cont.)

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- Software contains:
  - NDN and NFN Forwarder
  - client and command line tools
  - repo
- modular, easy to extend
- simple simulation and debugging system

# Qutline

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- Basic NFN Principles (what we want to do)
- Layering Overview
- Core of NFN: **expression rewriting**
- Example: expression rewriting based on **FIB information**
- Example: expression rewriting based on **mobility patterns**
- Example: expression rewriting based on **pricing**
- Orchestration of **"Plans"** and creating **"Templates"**
- More research topics: result authenticity, compute privacy

## Architecture / Core Components

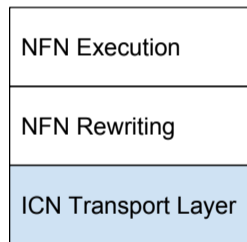
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NFN Execution	Execute Function Code
NFN Rewriting	Decide where to compute
ICN Transport Layer	Forward Messages

# CCN/NDN Layer

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- Basic CCNx/NDN Forwarding
- **Hierarchical** Names
- Forwarding based on **Longest Prefix Matching**
- **Name** in Interest Message contains reference to single **Named Data Object**

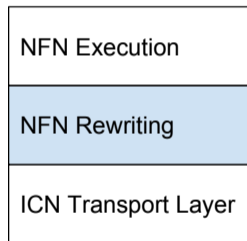




# NFN Rewriting 1

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- Workflow - consisting of **Function Calls** on **Input Data**
- Encoding the Workflow in CCN/NDN **Names**
- Using  $\lambda$ -**calculus** to describe workflow
- Name in Interest Message contains references to multiple **Named Data Object**
- Magic: **Longest Prefix Matching** and **Name Rewriting**



# NFN Rewriting 2: Name Encoding Example

Workflow:

`/func/f1 (/data/d1, /func/f1 (/data/d2) )`

Names:

`/func/f1`

`/data/d1`

`/func/f2`

`/data/d2`



Choose One Name to Prepend

`/data/d1`



NFN - Name Encoding

Name Components:

`data`

`d1`

`λ x. /func/f1(x, /func/f2( /data/d2 )`

`NFN`

NFN Execution

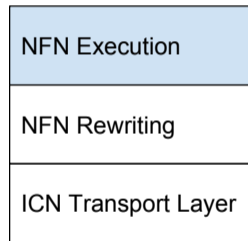
NFN Rewriting

ICN Transport Layer

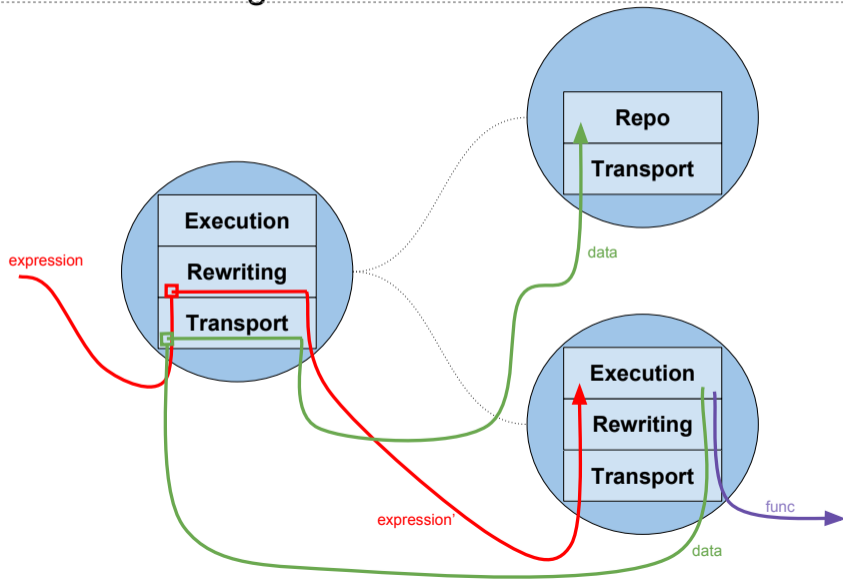
# NFN Execution

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- Store **Function Code** in **Named Data Objects**
- Any **NFN node** can execute **Function Code** by fetching it over ICN
- Requesting missing **Named Data Objects** and **Function Code**
- Requires Safe Execution Environment (Sandboxing)



# NFN Interest Handling



# Rewriting Decisions

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Interesting Part:

Which Name to **prepend** in front of the computation (**Rewriting**)

- Because: Influences where to **forward** a computation
- Defined by a Rewriting Strategy

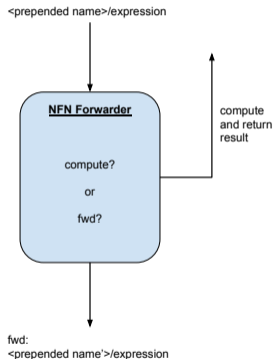
Should a node forward a computation or execute it locally?

- Determines where to compute a result.
- A node can split a computation into subcomputations  $\Rightarrow$  parallel/distributed execution

# Rewriting Strategy

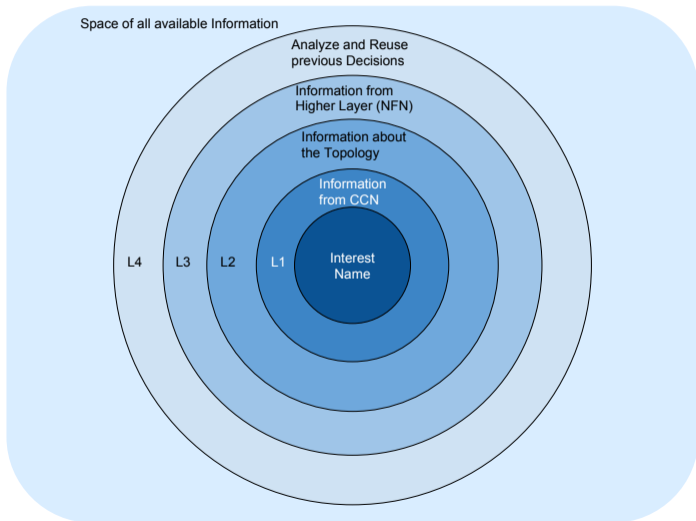
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- Simple Rewriting Strategy - inspired by Hadoop
- Goal: Reduce **load** on links
- Forward a **computation request** to the **input data** (prepend input data)
- Start computation if the prepend data are available on the node.
- Transport **function code** to **input data**



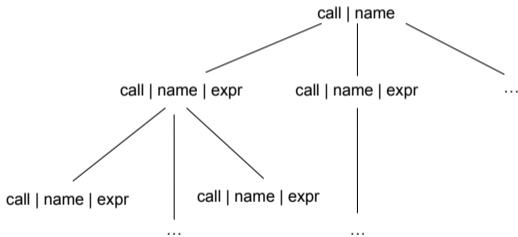
# Advanced Expression Rewriting for NFN

- Add additional information to improve the **rewriting decision**
- Scenario **dependent** and **independent**



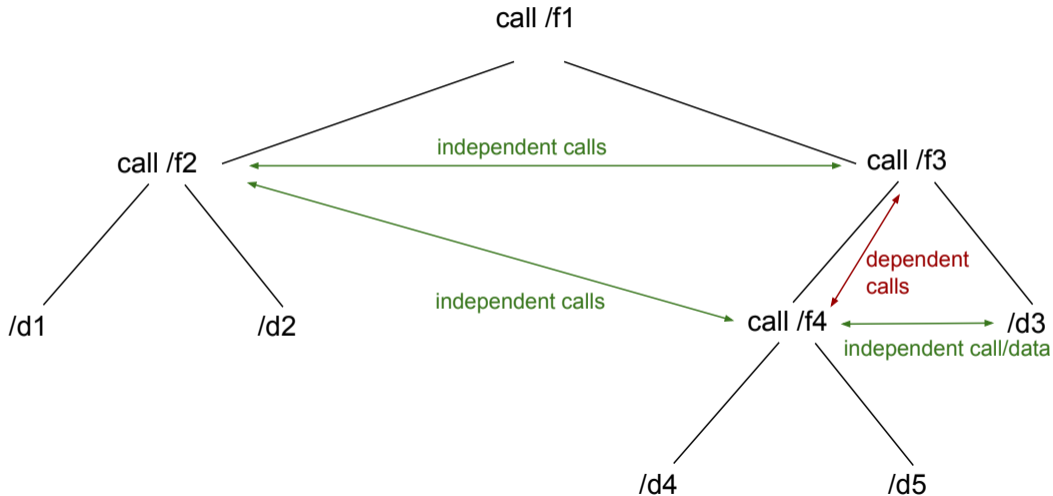
# NFN-Expression Rewriting based on **FIB** Information - 1

- Use the **FIB** to decide which name should be prepended
- Create **AST** and search for independent subcomputations
- Split the computation if subcomputations are forwarded to different nodes
- Good for Map Reduce and Parallel Execution



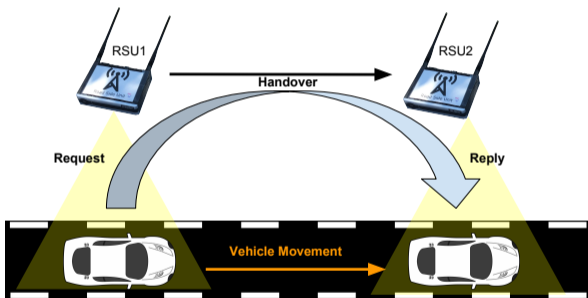


## NFN-Expression Rewriting based on **FIB** Information - 2



# NFN-Expression Rewriting based on Mobility Patterns

- Use information about **mobility patterns**
- e.g. Node is **Edge Computing Node**
- e.g. **Neighbor Node** has no computational capabilities
- Execute even if **prepended data** are **not** on the node



# NFN-Expression Rewriting based on the Price - 1

- Request **Meta Information** about **file size**, **bandwidth** or **load**
- Compute a **Plan** for the **cheapest** execution possibility
- Storing **Plans** in **Named Data Objects**
- **Caching** and **Reusing** of Plans (for subcomputations)

Plan:  
Name: <NFN - expression>

Actions:

1. FWD: prepend <name-1>
2. FWD: prepend <name-1>
3. FWD: prepend <name-2>
4. SPLIT: <Level>

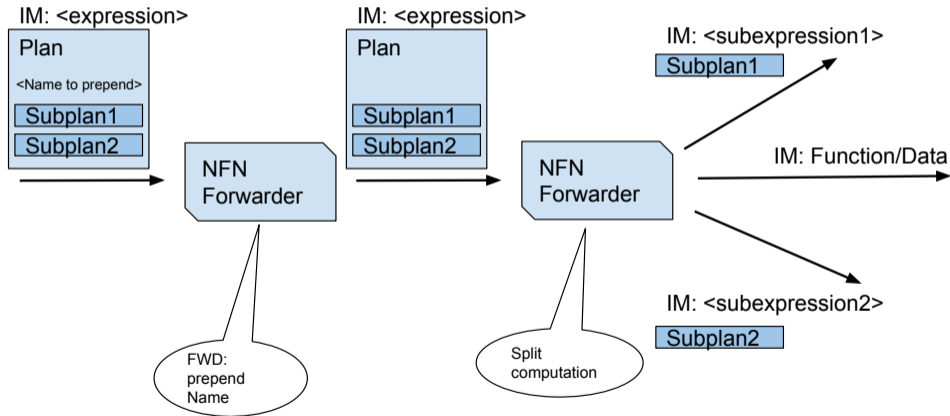
Subplan 1:  
Name <NFN - subexpression-1>

Actions:  
1. Exec

Subplan 1:  
Name <NFN - subexpression-2>

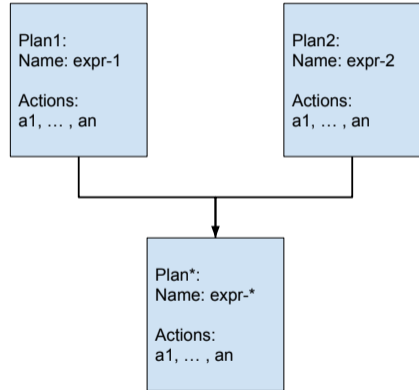
Actions:  
1. FWD: prepend <name-3>  
2. Exec

## NFN-Expression Rewriting based on the Price - 2



# NFN-Expression Rewriting based on Templates - 1

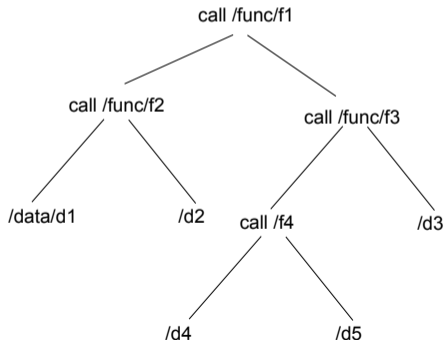
- Create **Templates** based on the **Planning Process**
- Idea: Instead of creating a new **Rewriting Strategy**, let the network learn from previous situations (or simulation)
- Compare **AST** structure and names within the **AST**
- Introduce **Wildcard** Names into the **Plans**
- Based on Observation and Statistics:
  - $\langle \text{prefix} \rangle / \text{name-1} \dots \langle \text{prefix} \rangle / \text{name-n}$   
→ Action  $a$
  - if  $n > x$  → reduce  $\langle \text{prefix} \rangle / *$
- In future with Machine Learning?



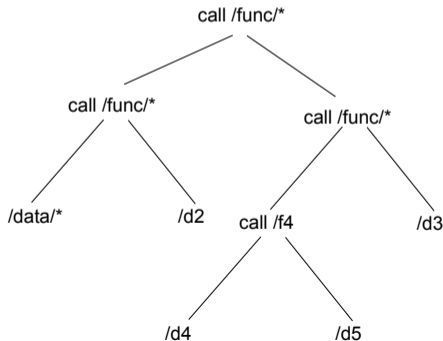
## NFN-Expression Rewriting based on Templates - 2

- A Template is a Tuple:  $\langle AST^*, Action \rangle$
- A request is matched against the  $AST$  containing the Wildcards

Request



Template



# Result Authenticity

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- Data are secured by **signatures**, what about **results**
- Question: How can you know the result is correct?
- Idea: Signed by execution node.
- Add Signatures of the Input Data and the Function Code
- Item: No Proof, but enables users to find out which nodes are fooling

# Data Security

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- Question: How to deal with sensible input data?
- Intel SGX may be a way
- Homomorphic Computing (increases runtime)
- Only real solution: Do it local



# Software

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- PiCN
- <https://github.com/cn-uofbasel/PiCN>
- ICN and NFN Forwarder, Client Tools, Repository
- Modular, easy to extend
- NFN Rewriting Strategies as Plugins
- Simple Simulation System

## References

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- <https://github.com/cn-uofbasel/PiCN>
- An information centric network for computing the distribution of computations (M. Sifalakis, B. Kohler, C. Scherb, C. Tschudin)
- Access-controlled in-network processing of named data (C. Marxer, C. Scherb, C. Tschudin)
- Resolution Strategies for Networking the IoT at the Edge via Named Functions (C. Scherb, D. Grewe, M. Wagner, C. Tschudin)
- Execution State Management in Named Function Networking (C. Scherb, B. Faludi, C. Tschudin)
- A Packet Rewriting Core for Information Centric Networking (C. Scherb, M. Sifalakis, C. Tschudin)
- Smart Execution Strategy Selection for Multi Tier Execution in Named Function Networking (C. Scherb, C. Tschudin)

Question