

# The EPI Framework

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## Before we begin

- I'm a computer scientist
- But I'll try to keep it nice and concrete :)
- Except when you see Ô
  - Then it does get technical

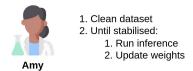
- Also note, the **EPI Framework** used to be called **Brane**, so I'll use them interchangeably!

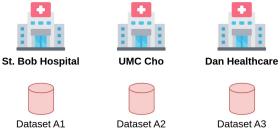
## I. Introduction

## Background

- There is a distributed dataset
  - Spread out over multiple hospitals (domains)
- Amy is a data scientist who wants to analyse it
- However: data is super-sensitive!

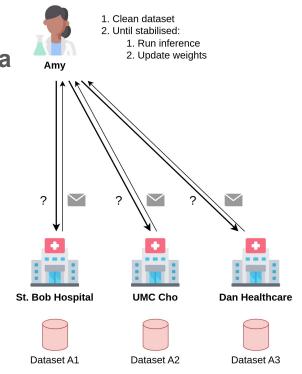
- How can Amy safely analyse the data?





## Naive approach

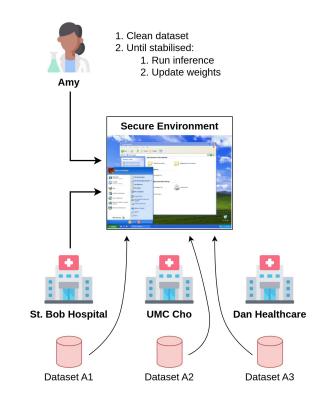
- Amy might request all domains for access to data
- Domains then send back the data in full
- She performs her steps locally
- Evaluation
  - Requires a lot of trust in Amy!
    - No oversight to what she does
  - Very tedious to arrange
    - Manual decision making
    - Manual preprocessing
  - Might be hard to share data securely



## Better approach

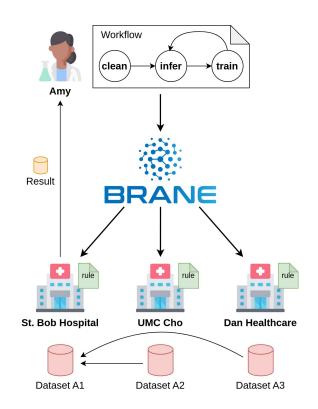
- One hospital creates **secure environment** 
  - Hospitals make data available in this environment
- Amy can perform her work "on-site"
- Evaluation
  - Still requires trust in Amy
    - Better, but still limited oversight
  - Still tedious to arrange
    - Manual decision making still occurs
    - Per-researcher setup

Might be hard to share data securely



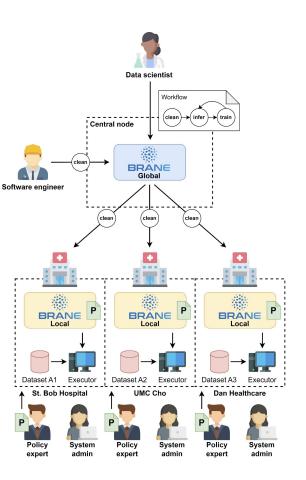
## **EPI** approach

- Amy can formalise her steps (workflow)
- Domains can formalise their regulations (policy)
- EPI Framework can perform the computation
  - While ensuring regulations are not violated
- Evaluation
  - Requires minimal trust in Amy
    - Her plan can be analysed before computation
  - Less tedious to arrange
    - Manual decision making only required sometimes
    - Policies pre-defined (easily applied)



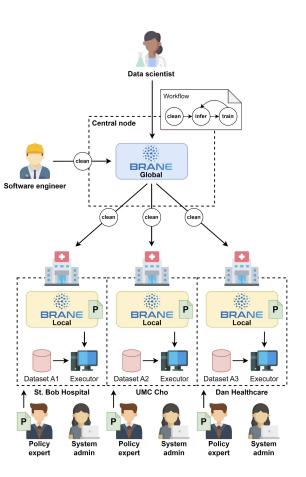
## The EPI Framework

- Built to share data for workflow, policy-compliant
- The framework consists of two components:
  - The **central component** has the overview and accepts incoming work
  - The **local components** (owned by hospitals) has the data and performs the work



## Separation of concerns

- Two central-side roles
  - Data scientists
  - Software engineer
- Two local-side roles
  - System administrator
    - Manages the components
  - Policy expert



## II. Workflows

Data scientists & Software engineers

## Recipes as workflows

- Recipes (as in cooking) are perfect examples of workflows
- Defines a series of steps/tasks to perform
- Mentions only relevant details
  - e.g., it doesn't say which chef executes a task
- Tasks may be dependent on each other

#### **Pasta Broccolo**

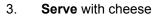
- 1. **Clean** the broccoli and cut it
- 2. **Boil** the broccoli pieces for 5 minutes, and the pasta for 12 minutes
- 3. Serve with cheese

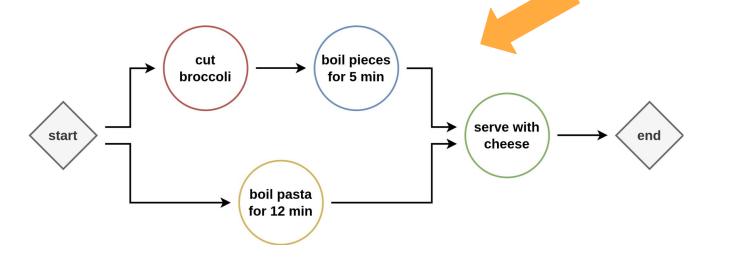
## Formalising recipes

- Recipes (workflows) can be represented as graphs
- Nodes are tasks
- Edges are dependencies



- 1. **Clean** the broccoli and cut it
- 2. **Boil** the broccoli pieces for 5 minutes, and the pasta for 12 minutes



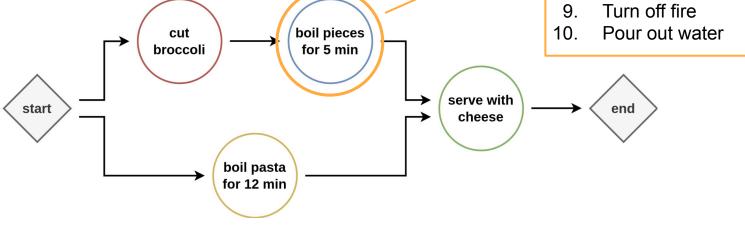


## Formalising recipes

- Recipes (workflows) can be represented as graphs
- Nodes are tasks
- Edges are dependencies

#### **Boiling broccoli pieces**

- 1. Get pan
- 2. Pour water in pan
- 3. Put pan on furnace
- 4. Put on lid
- 5. Light fire
- 6. Wait until bubbles
- 7. Add broccoli to pan
- 8. Wait 5 minutes

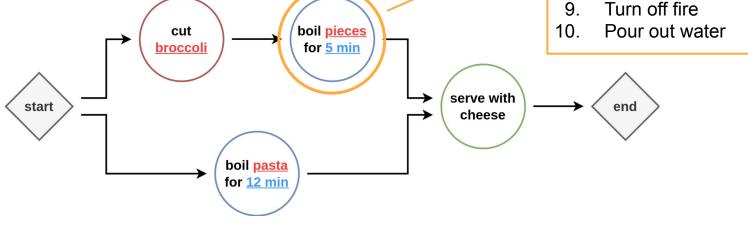


## Generalising tasks

- However, it's nice if some tasks can be parameterized \_
- This means we can **re-use tasks**! \_
  - ...even from previous workflows -

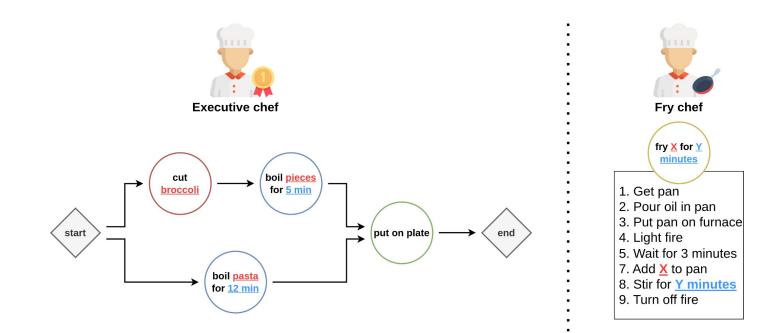
### Boiling X

- Get pan 1.
- 2. Pour water in pan
- 3. Put pan on furnace
- Put on lid 4.
- 5. Light fire
- 6. Wait until bubbles
- 7. Add X to pan
- 8. Wait <u>Y minutes</u>
- Turn off fire



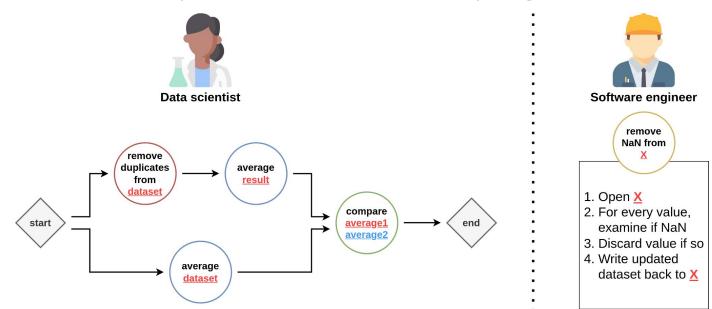
## Separation of concerns

- If we can re-use tasks, we can now write them beforehand
- Which means now **specialized people** can do **different things** in **parallel**!



## Formalising programs

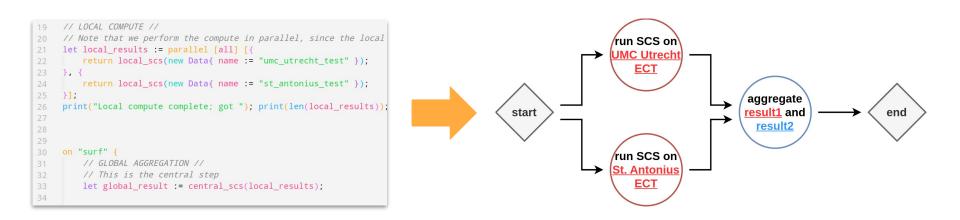
- We can formalize programs in exactly the same way
  - Also only describe high-level details
- Workflows written by scientists, tasks written by engineers



## Formalising workflows, practically



- The EPI Framework uses **BraneScript**
- Tasks are represented as functions, dependencies are derived
- Other methods supported in the future



## Formalising tasks, practically

- Tasks are represented as functions
- Functions are grouped in packages
- Packages are implemented as Docker containers
  - Inputs read from environment variables
  - Output written to stdout
- Run isolated (input and output dictated by Brane)



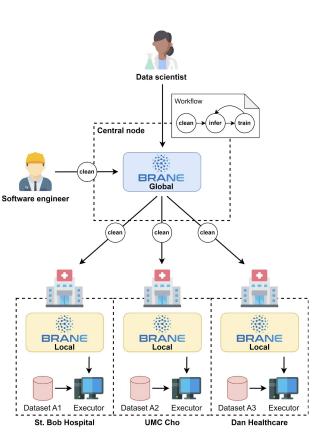
8		6
9	<i>#combine results and multiple E variables</i>	6
0	<pre>multipliedEVariablesPerStratumPerDelta &lt;- localEValueRe:</pre>	6
1	<pre>pivot_wider(names_from = mTotal, values_from = mE) %&gt;9</pre>	6
2	<pre>pivot_longer(cols = -c(stratum, delta), names_to = "m"</pre>	6
3	<pre>mutate(mTotal = as.numeric(mTotal)) %&gt;%</pre>	7
4	#when we did not gain new information at a certain til	
5	<pre>mutate(mE = replace_na(mE, 1)) %&gt;%</pre>	7
6	<pre>left_join(localEValueResult2, by = c("stratum", "delta</pre>	7
7	<pre>mutate(mE2 = replace_na(mE2, 1)) %&gt;%</pre>	7
8	<pre>mutate(mETotal = mE1 * mE2) %&gt;%</pre>	7



51	<ul> <li>stratifiedConfidenceFunctions POC.</li> </ul>
52	- POC helper functions.R
53	
54	# Define the entrypoint: i.e., whick
55	entrypoint:
56	kind: task
57	exec: run.sh
58	
59	# Define the functions in this packa
60	actions:
61	'local_scs':
62	command:
63	args:
64	- local_scs
65	<i># It takes a dataset (the local</i>
66	input:
67	- type: Data
68	name: input
69	<i># It outputs the local result</i>
70	output:
71	<ul> <li>type: IntermediateResult</li> </ul>
72	name: output
73	'central_scs':
74	command:
75	args:
76	- central_scs
77	capture: marked

## Takeaways

- Workflows formalise analyses of data scientists
  - Encoded as a series of **tasks** with **dependencies**
  - Represented as a graph
- Doing so, they are high-level programs
  - Only details relevant for the scientist are expressed
  - Others are inferred by the framework's "expertise"



## **III.** Policies

Policy experts

## Back in the kitchen...

- Policies are **constraints** on what *should* happen
  - Directly, they prohibit some things happening in the workflow
- Can be from various sources
  - Laws (GDPR), organisational policies, contracts, etc
- Can be on different levels
  - Directly prohibit actions, impose conditions, ...

### Kitchen rulez

- 1. Your workspace must be hygienic
- 2. You must wash your hands before touching food
- 3. Listen to your boss
- 4. Thou shalt not put pineapple on pizza

## The problem with policies

- Policies, however, tend to be very vague
  - Especially laws, to allow a judge to interpret
- Various sources of vagueness
  - **Abstract** terms (e.g., "What does 'hygienic' mean?")
  - Incomplete definition (e.g., "How often do I need to wash my hands?")
  - Needing context information (e.g., "Who is my boss?")

#### Kitchen rulez

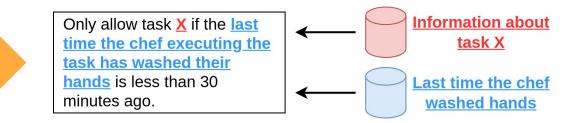
- 1. Your workspace must be <u>hygienic</u>
- 2. You must <u>wash your</u> <u>hands</u> before touching food
- 3. Listen to your <u>boss</u>
- 4. Thou shalt not put pineapple on pizza

## **Constraining recipes**

- Luckily, we can scope kitchen policies to recipes (workflows)
  - These are already very concrete and formal (executable/computable)
- Thus we can express policies as rules over recipes
  - Plus extra context
- Doing so forces us to get concrete!

#### Kitchen rulez

- 1. Your workspace must be hygienic
- 2. You must wash your hands before touching food
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- 4. Thou shalt not put pineapple on pizza



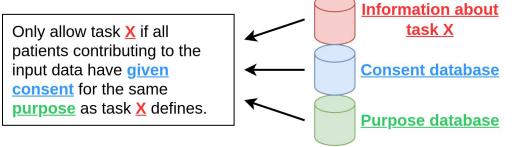
## Constraining workflows

- The same therefore applies to EPI Framework workflows!
- Can be defined by hospitals individually
  - I.e., hospitals are in charge of their own behaviour
- Written by policy experts



#### **GDPR** (approximately)

- 1. You can only use data for the purpose patients have given consent for.
- Patients must be able to revoke their consent
   ....



## Expected workflow policy types

- Policies fundamentally control data
- Tasks access data as part of a workflow
- As such, policies will allow/deny tasks to process their data
- Examples:
  - Laws (GDPR)

"Tasks using my data are only allowed if no patients sourcing that data has retracted consent"

- Organisational policies

"The result of tasks processing this dataset may only be seen by people with this role"

- Agreements/Contracts

"This dataset is allowed to be transferred to St. Bob if used for this task and this workflow"

## Policies as reasoners

- The complexity of policies shows policies need reasoning
  - Specifically: policies need to be **logical rules**
- Logic programming languages already exist
  - Datalog<sup>1</sup>
  - eFLINT<sup>2</sup>
  - SEASO<sup>3</sup>
  - ...
- However, can be any language
  - All that matters is that an **allow/deny** is produced
- <sup>1</sup> <u>https://www2.cs.sfu.ca/CourseCentral/721/jim/DatalogPaper.pdf</u>
- <sup>2</sup> <u>https://gitlab.com/eflint</u>
- <sup>3</sup> <u>https://github.com/sirkibsirkib/seaso</u>

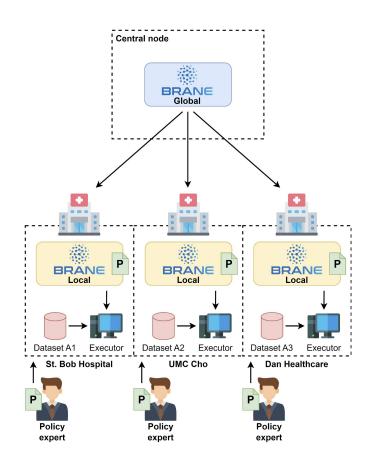


8	Fact consent Identified by patient * purpose.
9	Fact contributed Identified by patient * data.
10	
11	Fact input Identified by data * task.
12	
13	Fact data-purpose Identified by data * purpose
14	Holds when (Exists patient : contributed(patient, d
15	Fact task-purpose Identified by task * purpose.
16	
17	Fact allow Identified by task
18	<pre>Holds when Not(Exists data : input(data, task))   </pre>
19	
20	
21	+purpose(ColdResearch).
22	+data("A").
23	<pre>+patient("Amy").</pre>
24	<pre>+contributed(patient(Amy), data(A)).</pre>
25	<pre>+consent(patient(Amy), purpose(ColdResearch)).</pre>
26	<pre>+patient("Bob").</pre>
27	+contributed(nation(Rob)) data(A))



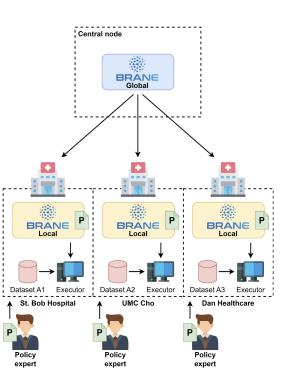
### Policies as brain

- Policies determine a hospital's actions
- Completely in **hospital control** (i.e., hospitals have autonomy)
- This also affords dynamic updates
  - Based on situations, new laws, ...



## Takeaways

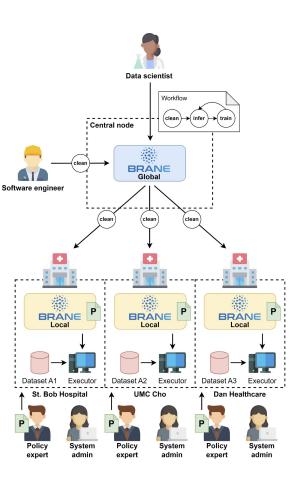
- Policies are **formalisations** of various kinds of rules
  - Formalising laws, organisations policies, agreements, ...
  - They are **concretised** versions of the original rule
- Policies constrain which tasks are allowed
  - And therefore workflows
- Expressed as logic rules (or whatever is needed)



## **IV.** Conclusion

## Takeaways

- The EPI Framework is a data-sharing platform
  - Designed for research context
- Built to understand the work at hand (workflow) and test it to policy
  - Ensures compliance of the scientist's actions
- Separation of concerns to harness complexity



## What next?

- If you're curious, check the wiki!
  - https://wiki.enablingpersonalizedinterventions.nl
- System requirements
  - Local: Windows, macOS or Linux machine with Docker<sup>1</sup>
  - Complete: Local machine + Linux server running Docker<sup>1</sup>





### Come see the demo in the second hour! :)





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https://enablingpersonalizedinterventions.nl

https://github.com/epi-project/brane

https://wiki.enablingpersonalizedinterventions.nl



The icons (not logos) in this presentation are from: Freepik, Ultimatearm, Vector Valley

## V. Bonus slides

## The dream of policies

Ö

- "hospitals have autonomy" (slide 27) is problematic!
- It's crucial that hospitals are autonomous...
- ...but therefore, we can't force them to (not) do things!
  - They can always leak the data we share somehow
- As such, writing a policy does not (necessarily!) enforce it
  - Specifically: a hospital is guaranteed control until shared
- Policies thus need to **consider trust** in receiving parties