

How to perform correct and efficient Bayesian inference

- Previous methods focussed on either *correct* or *efficient*, not both
- *correctness*: general purpose probabilistic programming languages
- *efficiency*: programming from scratch

Blang

- Our effort: Blang, a Bayesian modelling language focusing on supporting correct and efficient combinatorial space sampling
 - Open source
 - Project page: <https://www.stat.ubc.ca/~bouchard/blang/>

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Declarative model construction

- `model`: a collection of probability distributions (law's) indexed by *parameters* (i.e. a family)
- one way to build a model is to *declare* (conditional) probabilities for each random variable (syntax inspired by WinBUG/JAGS)

```
x | lambda ~ Poisson(lambda)
```

- motivation: stay close to mathematical notation

```
model SimplePhyloExample {
  random RealVar shape ? : latentReal, rate ? : latentReal
  random SequenceAlignment observations
  random UnrootedTree tree ? : unrootedTree(observations.observedTreeNodees)
  param EvolutionaryModel evoModel ? : kimura(observations.nSites)

  laws {
    shape ~ Exponential(1.0)
    rate ~ Exponential(1.0)
    tree | shape, rate ~ NonClockTreePrior(Gamma::distribution(shape, rate))
    observations | tree, evoModel ~ UnrootedTreeLikelihood(tree, evoModel)
  }
}
```

Model composition

| Language | Language used for its standard library |
|-----------------|--|
| Win / Open BUGS | Pascal |
| RevBayes | C++ |
| JAGS | C++ |
| Stan | C++ |
| Blang | Blang |

Model composition

- `model`: a collection of distributions indexed by *parameters*
- can be specified using conditional distributions
- a conditional distribution can be built from a parameterized collection of distributions (parameter \rightsquigarrow conditional)
- \Rightarrow can build a `model` using other `model`'s

```
model SimplePhyloExample {  
  ...  
  
  laws {  
    shape ~ Exponential(1.0)  
    ...  
  }  
}
```

```
model Exponential {  
  random RealVar realization  
  param RealVar rate  
  
  laws {  
    realization | rate ~ Gamma(1.0, rate)  
  }  
}
```

```
model Gamma {  
  random RealVar realization  
  param RealVar shape  
  param RealVar rate  
  
  laws {  
    logf(shape, rate, realization) {  
      if (shape <= 0.0 || rate <= 0)  
        return NEGATIVE_INFINITY  
      if (realization <= 0.0)  
        return NEGATIVE_INFINITY  
      ...  
    }  
    ...  
  }  
}
```

Composition via parameters

- Pushing this idea a bit further, parameters themselves can be models (distributions)
- Other examples: distribution of branch lengths, Dirichlet process' base measure, etc.

```
model IntMixture {
  param Simplex proportions
  param List<IntDistribution> components
  random IntVar realization

  laws {
    logf(proportions, components, realization) {
      var sum = 0.0
      if (components.size != proportions.nEntries) throw new RuntimeException
      for (i : 0 ..< components.size) {
        val prop = proportions.get(i)
        if (prop < 0.0 || prop > 1.0) return NEGATIVE_INFINITY
        sum += prop * exp(components.get(i).logDensity(realization))
      }
      return log(sum)
    }
  }

  generate (rand) {
    val category = rand.categorical(proportions.vectorToArray)
    return components.get(category).sample(rand)
  }
}
```

```
...
  observation | proportions, lambda, rho
    ~ IntMixture(
      proportions,
      #[
        Poisson::distribution(lambda),
        YuleSimon::distribution(rho)
      ]
    )
...
```

Usage example

Example: a mixture of discrete distributions, taking a list of distributions as parameter: the mixture components

- Each Blang model is turned into an inference program (currently command line, but more interfaces under development; currently working on the r interface)
- This program takes as input observed data and outputs posterior samples for the unobserved variables
Concretely, inputs are currently command line arguments for each variable (organized hierarchically, most with sensible default values. try `--help`).
Outputs are *tidy* csv files (Wickham, 2013).

Try at home!
Needed: Oracle
Java 8, git,
POSIX

```
> git clone https://github.com/UBC-Stat-ML/blangExample.git
[cloning]

> ./gradlew installDist
[downloading dependencies and compiling]

> ./build/install/example/bin/example \
  --model.observations.file data/primates.fasta \
  --model.observations.encoding DNA \
  --engine SCM \
  --engine.nThreads Max \
  --excludeFromOutput observations

Preprocessing started
4 samplers constructed with following prototypes:
RealScalar sampled via: [RealSliceSampler]
UnrootedTree sampled via: [SingleNNI, SingleBranchScaling]
Sampling started
[sampling progress report]
Normalization constant estimate: -1216.1211229417504
Final rejuvenation started
Preprocessing time: 141.4 ms
Sampling time: 2.304 min
executionMilliseconds : 138405
```

Open type system

- Blang allows object-oriented development of custom random variable datatypes (combinatorial objects)

```
/**
 * An unrooted phylogenetic tree.
 * @author Alexandre Bouchard (alexandre.bouchard@gmail.com)
 */
@Samplers(SingleNNI, SingleBranchScaling)
class UnrootedTree {
    val Map<UnorderedPair<TreeNode, TreeNode>, Double> branchLengths
    val UndirectedGraph<TreeNode, UnorderedPair<TreeNode, TreeNode>> topology
    ...
}
```

```
class SingleBranchScaling extends MHSampler {
    @SampledVariable UnrootedTree variable

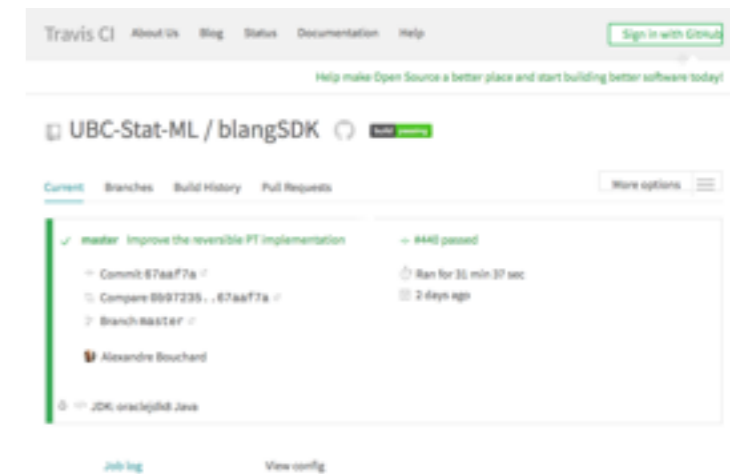
    override propose(Random rand, Callback callback) {
        val allEdges = newArrayList(variable.topology.edgeSet)
        val edge = sample(allEdges, rand)
        val oldValue = variable.getBranchLength(edge)
        val m = exp(2.0 * log(2.0) * (rand.nextDouble - 0.5))
        callback.setProposalLogRatio(log(m))
        variable.updateBranchLength(edge, m * oldValue)
        if (!callback.sampleAcceptance)
            variable.updateBranchLength(edge, oldValue)
    }
}
```

- For scalability, user may need to write a sampler, but..
- user can write using the same syntax
- Bar is lower in terms of efficiency : thanks to advanced posterior simulation methods (beyond MCMC)
- Blang helps you checking *correctness* of implementation

Correctness

- How to check MCMC code is invariant and/or irreducible?
- Seems hard; surprisingly very good tests can be constructed
- Example: for discrete `model`'s, Blang's `DiscreteMCTest` utility will:
 - enumerate all the execution traces for each associated samplers
 - compute the posterior π by enumeration
 - build an explicit transition matrix M for each sample
 - check that $\pi = M\pi$, as well as irreducibility
- Many more tests, existing and novel, including tests for continuous models (e.g., Geweke, 2004)

build passing



```
Worker information
Build system information

Blang_TestSDKDistributions > determineTest STANDARD_OUT
2472 Running DetermineTest on model: RecBinomial [IntSliceSampler]
2473 Running DetermineTest on model: Hyperbolic [IntSliceSampler]
2474 Running DetermineTest on model: Geopertz [RealSliceSampler]
2475 Running DetermineTest on model: SumOf [RealSliceSampler]
2476 Running DetermineTest on model: Weibull [RealSliceSampler]
2477 Running DetermineTest on model: F [RealSliceSampler]
2478 Running DetermineTest on model: Logistic [RealSliceSampler]
2479 Running DetermineTest on model: Logistic [RealSliceSampler]
2480 Running DetermineTest on model: Geometric [IntSliceSampler]
2481 Running DetermineTest on model: Laplace [RealSliceSampler]
2482 Running DetermineTest on model: Uniform [IntSliceSampler]
2483 Running DetermineTest on model: Normal [RealSliceSampler]
2484 Running DetermineTest on model: StudentT [RealSliceSampler]
2485 Running DetermineTest on model: StudentT [RealSliceSampler]
2486 Running DetermineTest on model: StudentT [RealSliceSampler]
2487 Running DetermineTest on model: HalfStudentT [RealSliceSampler]
2488 Running DetermineTest on model: ChiSquared [RealSliceSampler]
2489 Running DetermineTest on model: Bernoulli [CategoricalSampler]
2490 Running DetermineTest on model: Beta [RealSliceSampler]
2491 Running DetermineTest on model: NegativeBinomial [IntSliceSampler]
2492 Running DetermineTest on model: NegativeBinomial [IntSliceSampler]
2493 Running DetermineTest on model: NegativeBinomialMeanParam [IntSliceSampler]
2494 Running DetermineTest on model: Beta [RealSliceSampler]
2495 Running DetermineTest on model: Binomial [IntSliceSampler]
2496 Running DetermineTest on model: Categorical [CategoricalSampler]
2497 Running DetermineTest on model: ContinuousUniform [RealSliceSampler]
2498 Running DetermineTest on model: DiscreteUniform [IntSliceSampler]
2499 Running DetermineTest on model: Dirichlet [SimplexSampler]
2500 Running DetermineTest on model: Dirichlet [SimplexSampler]
2501 Running DetermineTest on model: SymmetricDirichlet [SimplexSampler]
2502 Running DetermineTest on model: LaplaceUniform [SimplexSampler]
2503 Running DetermineTest on model: Exponential [RealSliceSampler]
2504 Running DetermineTest on model: Gamma [RealSliceSampler]
2505 Running DetermineTest on model: GammaMeanParam [RealSliceSampler]
2506 Running DetermineTest on model: MultivariateNormal [RealSliceSampler]
2507 Running DetermineTest on model: Poisson [IntSliceSampler]
2508 Running DetermineTest on model: MarkovChain [CategoricalSampler]
2509 Running DetermineTest on model: DynamicNormalMixture [SimplexSampler]
[RealSliceSampler] [CategoricalSampler]
```

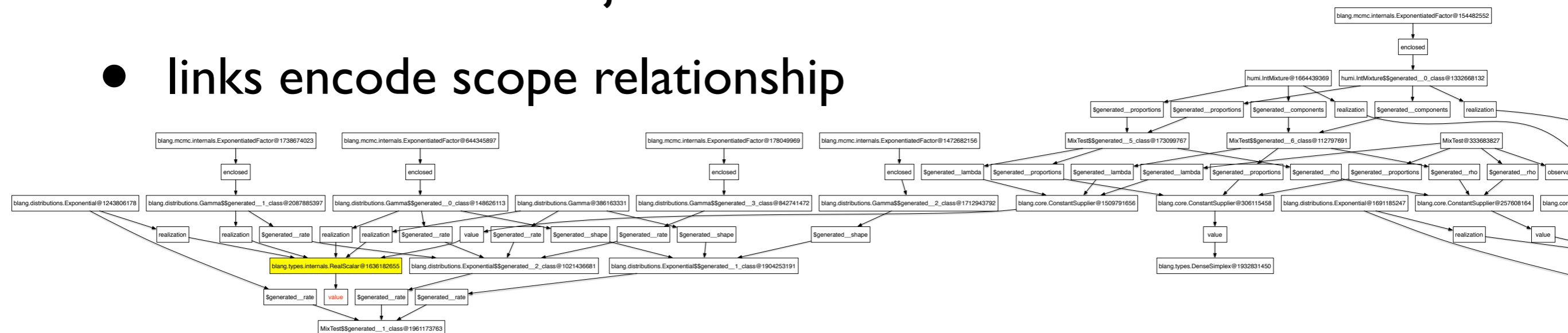
JUnit integration

Posterior inference

- Blang uses the declarative syntax to automatically build a *sequence* of distributions instead of just the posterior distribution
- Key to the advanced *inference engines* available
 - *Non-reversible Parallel Tempering*
 - *Sequential Change of Measure*

Exploiting sparsity

- Samplers can be much faster if they avoid recomputing all model components
- these savings are formalized by sparse *factor graphs* (Clifford, 1990)
- Idea: building factor graphs using *scoping information*
- Details: get factor graph by running linear time graph algorithms on an *accessibility graph*:
 - in which vertices objects
 - links encode scope relationship



Language details

- Built using Xtext, (framework for designing programming languages)
- Allowed us to “quickly” (3yrs) build an extensive multi-paradigm language
 - in addition to declarative capability, supports functional, generic and object programming, static typing, just-in-time compilation, garbage collection, etc
- advanced IDE support
- Runs on JVM, interoperates with Java
 - fast (just-in-time compiled)
 - can exploit arbitrarily many cores

```
grammar ca.ubc.stat.blang.BlangDsl with
org.eclipse.xtext.xbase.annotations.XbaseWithAnnotations

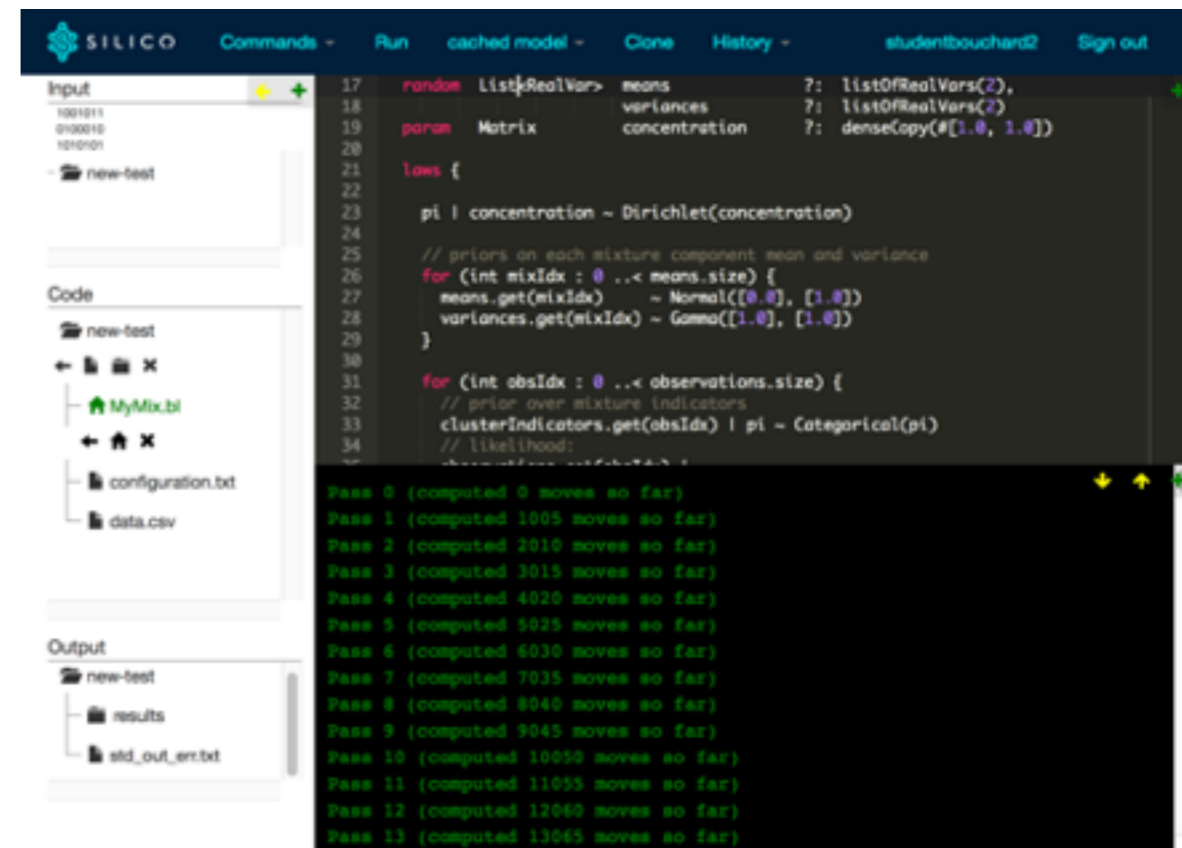
generate blangDsl "http://www.ubc.ca/stat/blang/BlangDsl"

import "http://www.eclipse.org/xtext/common/JavaVMTypes" as jvmTypes
import "http://www.eclipse.org/Xtext/Xbase/XAnnotations" as xAnnotations

BlangModel:
    {BlangModel}
    ('package' package=QualifiedName)?
    importSection=XImportSection?
    (annotations += XAnnotation)*
    'model' name=ID '{'
    (variableDeclarations += VariableDeclaration)*
    'laws' '{' (lawNodes += LawNode)* '}'
    ('generate' '(' generationRandom = ValidID ') generationAlgorithm =
XBlockExpression )?
    (variableDeclarations += VariableDeclaration)*
    '}'
;
...
```

Workflows

- Blang IDE:
 - leverages static types (eg smart links, call/type hierarchies, refactoring)
 - debugging, profiling, code coverage analysis, etc
- Web IDE
- Command line



Reproducibility & dissemination

- Code fully deterministic
- even in multithread mode
- checked via test units
- Create versioned packages containing models/samplers that others can import
- built on Maven
- smart dependency resolution

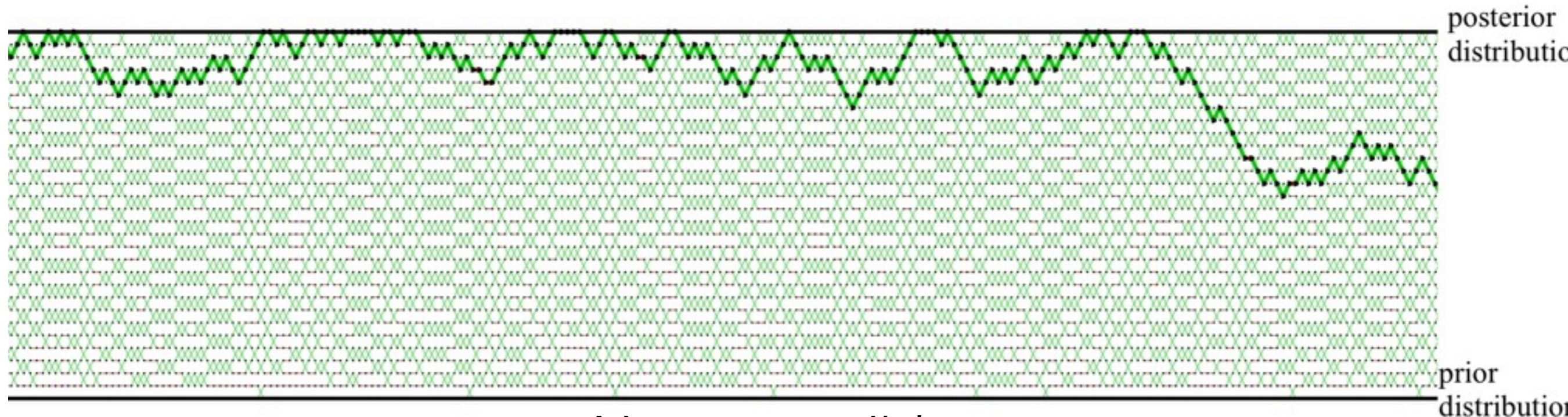
dependencies.txt

```
ca.ubc.stat.nowellpack:1.0.5  
ca.ubc.stat.conifer:2.0.4
```

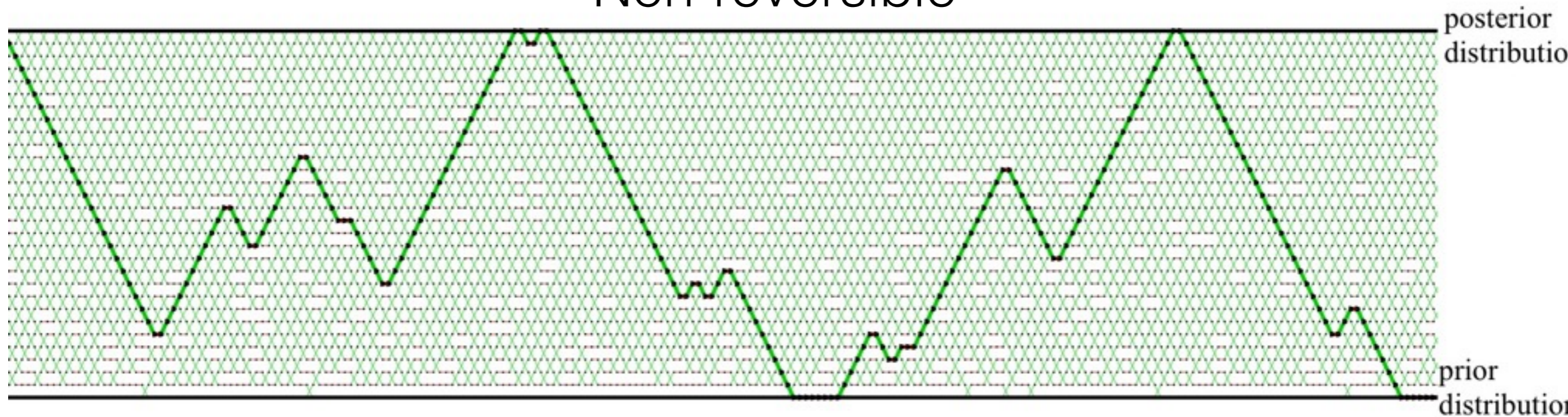
```
ca.ubc.stat.blanc:ca.ubc.stat.blanc:3.25.2  
org.eclipse.jdt:org.eclipse.jdt.core:3.10.0  
org.eclipse.core:org.eclipse.core.resources:3.7.100  
org.eclipse.core:org.eclipse.core.expressions:3.4.300  
org.eclipse.core:org.eclipse.core.runtime:3.7.0  
org.eclipse.osgi:org.eclipse.osgi:3.7.1  
org.eclipse.equinox:org.eclipse.equinox.common:3.6.0  
org.eclipse.core:org.eclipse.core.jobs:3.5.100  
org.eclipse.equinox:org.eclipse.equinox.registry:3.5.101  
org.eclipse.equinox:org.eclipse.equinox.common:3.6.0  
org.eclipse.equinox:org.eclipse.equinox.preferences:3.4.1  
org.eclipse.equinox:org.eclipse.equinox.common:3.6.0  
org.eclipse.core:org.eclipse.core.contenttype:3.4.100  
org.eclipse.equinox:org.eclipse.equinox.preferences:3.4.1 (*)  
org.eclipse.equinox:org.eclipse.equinox.registry:3.5.101 (*)  
org.eclipse.equinox:org.eclipse.equinox.common:3.6.0  
org.eclipse.equinox:org.eclipse.equinox.app:1.3.100  
org.eclipse.core:org.eclipse.core.filesystem:1.3.100  
org.eclipse.equinox:org.eclipse.equinox.common:3.6.0  
org.eclipse.equinox:org.eclipse.equinox.registry:3.5.101 (*)  
org.eclipse.osgi:org.eclipse.osgi:3.7.1  
org.eclipse.core:org.eclipse.core.runtime:3.7.0 (*)  
org.eclipse.core:org.eclipse.core.runtime:3.7.0 (*)  
org.eclipse.core:org.eclipse.core.filesystem:1.3.100 (*)  
org.eclipse.text:org.eclipse.text:3.5.101  
org.eclipse.core:org.eclipse.core.commands:3.6.0  
org.eclipse.equinox:org.eclipse.equinox.common:3.6.0  
org.eclipse.xtext:org.eclipse.xtext:2.12.0  
org.eclipse.xtext:org.eclipse.xtext.util:2.12.0  
org.eclipse.xtext:org.eclipse.xtext.lib:2.12.0  
org.eclipse.xtext:org.eclipse.xtext.xbase.lib:2.12.0  
com.google.guava:guava:14.0.19 -> 19.0  
org.eclipse.xtext:org.eclipse.xtext.lib:2.12.0  
org.eclipse.xtext:org.eclipse.xtext.xbase.lib:2.12.0 (*)  
log4j:log4j:1.2.16  
org.eclipse.emf:org.eclipse.emf.common:[2.10.1,3) -> 2.15.0  
org.eclipse.emf:org.eclipse.emf.ecore:[2.10.2,3) -> 2.15.0  
org.eclipse.emf:org.eclipse.emf.common:[2.15.0,3.0.0) -> 2.15.0  
com.google.inject:guice:3.0  
javax.inject:javax.inject:1  
aopalliance:aopalliance:1.0
```


Visualization: paths

Reversible



Non-reversible



Visualization: adaptation

