

A Lightweight, Scalable Grid Computing Framework for Parallel Bioinformatics Applications

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Outline

- (1) Introduction: why parallel bioinformatics on computational grids
- (2) The TaskSpaces framework for grid computing
- (3) Application: finding correctly folded active RNA motifs
- (4) Conclusions and future work

(1) Introduction: why parallel bioinformatics

- bioinformatics: large data sets
 - bioinformatics: computationally expensive algorithms
- ⇒ 1 CPU is not enough, need distributed/parallel computing for bioinformatics

Introduction: why grid computing?

- where can I find computational power?
 - my desktop machine (1 CPU)
 - other computers in my lab (~ 10 CPUs)
 - linux cluster at CS department (~ 100 CPUs), SHARCNET
 - superclusters at SHARCNET (~ 1000 CPUs, at Waterloo)
 - parallel supercomputers at SHARCNET (~ at McMaster)
- ‘dream’: depending on problem size, I want to be able to run the problem on any of these machines
(or all at the same time when I have a very large problem, or when I have to beat a deadline!)

why grid computing?

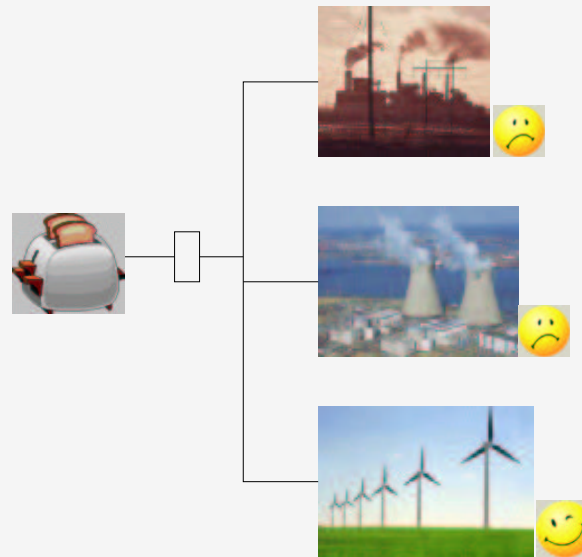
- problems with this 'dream':
 - variety of hardware
 - variety of operating systems, OS versions
 - variety of software versions
 - variety of queueing systems
 - need to install and maintain code on all those machines?
 - do I distribute work/data by hand? scripts? how are results centralized?
 - what happens to scripts when machines are added, removed?
 - ...

grid computing

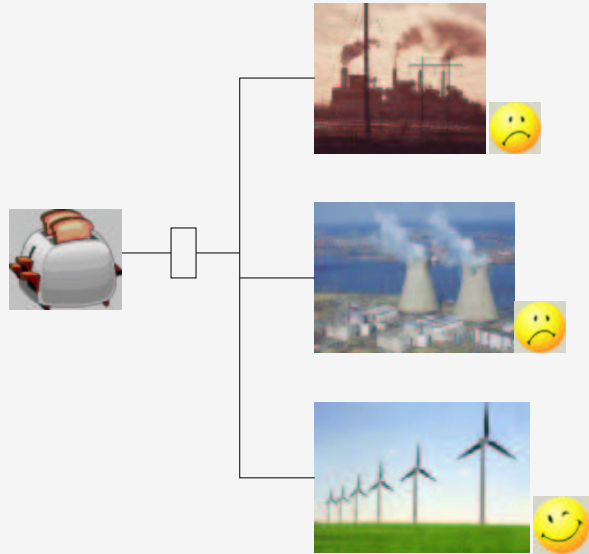
- an answer: concept of 'grid computing'
 - analogy: power grid
 - user wants electrical power
 - user doesn't care where the power is produced
- 'electrical power is interchangeable commodity'



(1) standard interface:
electrical plug



grid computing

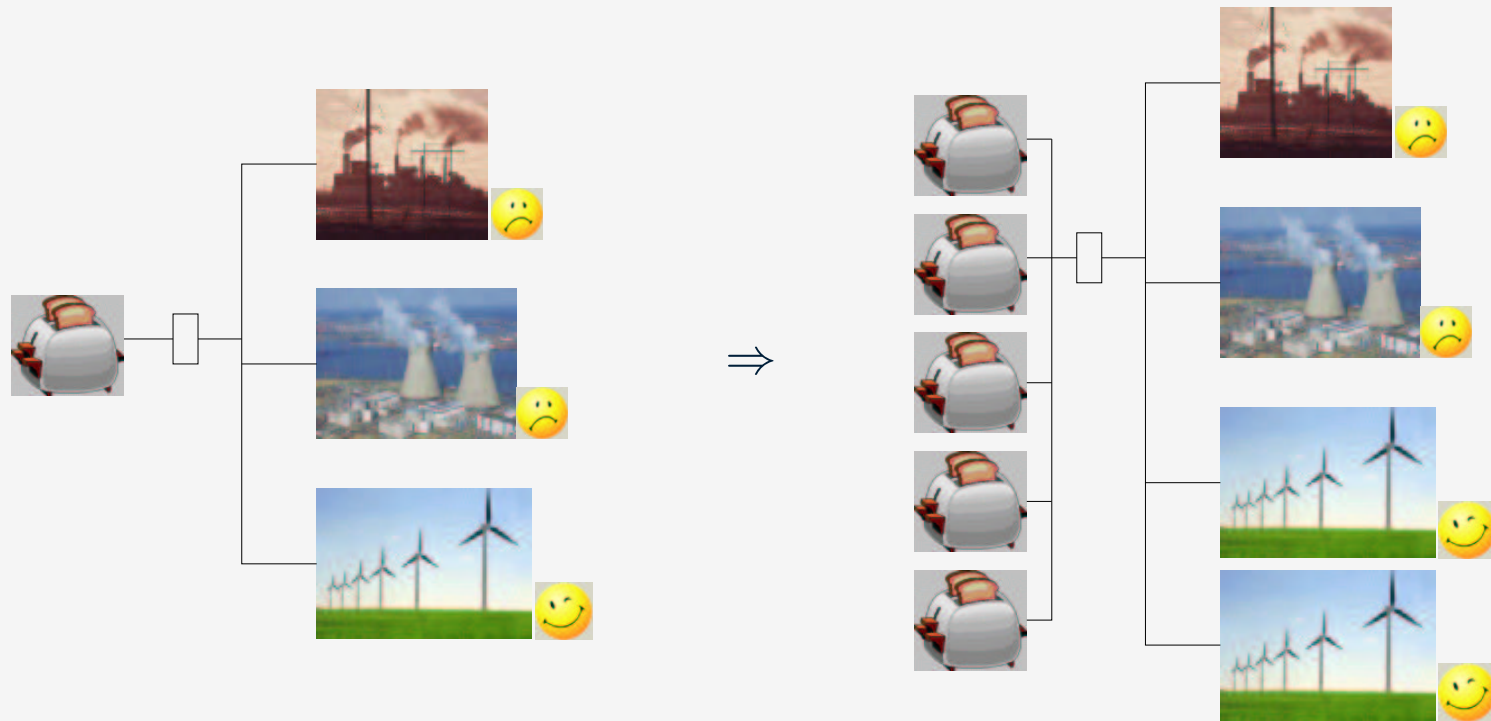


(2) scalable:

user: many appliances = many plugs

producer: when extra power is needed, switches in extra power plants

grid computing



(2) scalable:

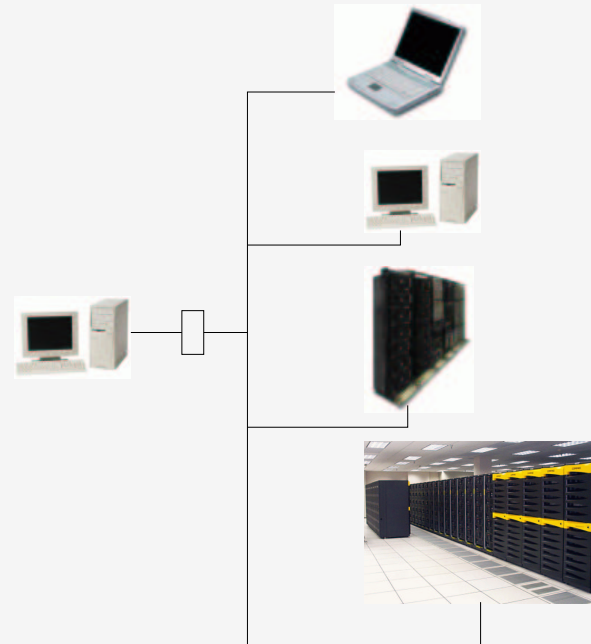
user: many appliances = many plugs

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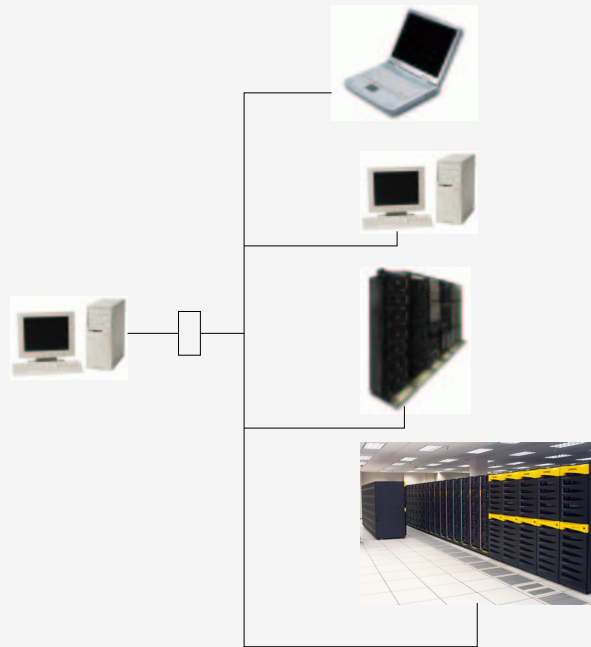
grid computing

- grid computing: analogous concept
 - user wants computing power
 - computing cycles
 - storage
 - network capacity
 - user doesn't care where the power is produced
 - 'computing power is interchangeable commodity'

(1) standard interface



grid computing

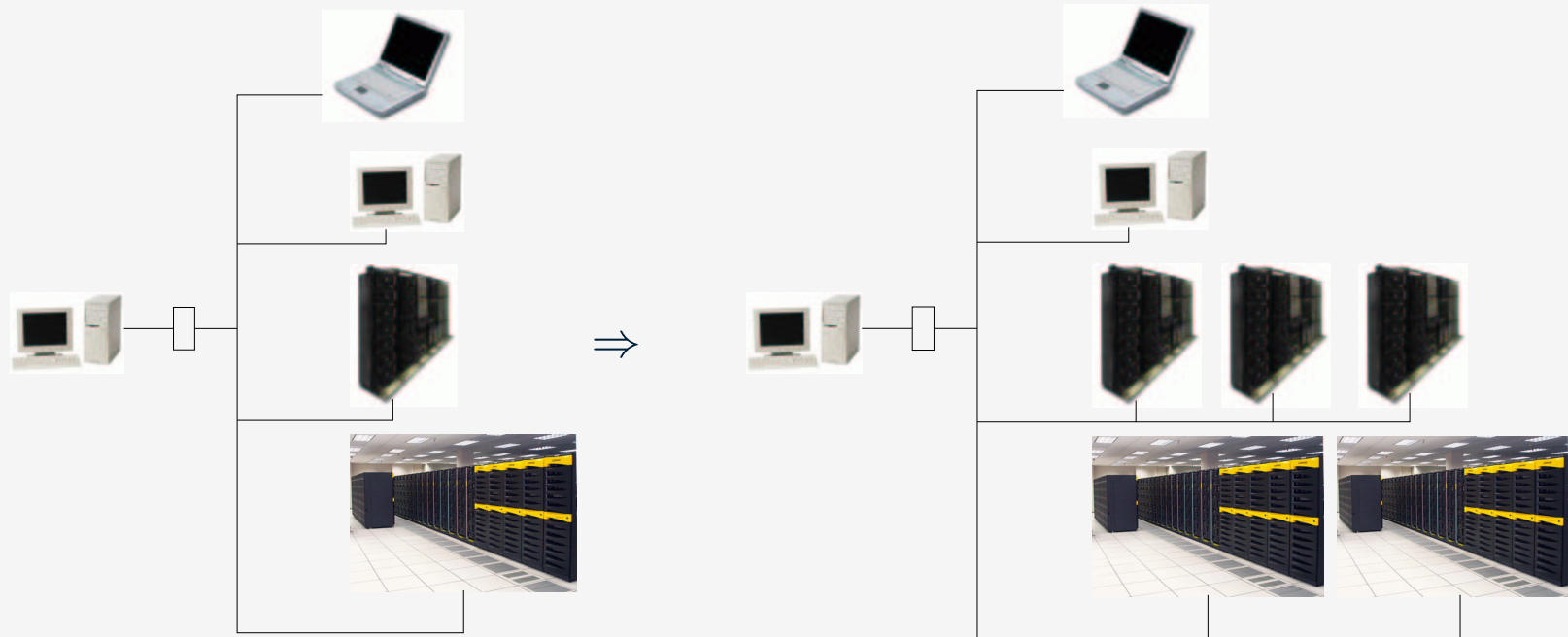


(2) scalable:

user: many applications at the same time

producer: when extra power is needed, switches in extra
'computing plants' (compute engines)

grid computing



(2) scalable:

user: many applications at the same time

producer: when extra power is needed, switches in extra
'computing plants' (compute engines)

grid computing

- the analogy is not perfect: computing power \neq electrical power
- some additional requirements for computational grids:
 - (3) secure resource sharing
 - information is not anonymous (unlike electrical power)
 - (4) 'transactions', fault-tolerance
 - information is not replaceable (unlike electrical power)
 - (5) resource allocation, scheduling (queues)
 - large computers work with queueing systems
 - (6) install and compile application code and data on compute engines
 - variety of hardware, OS, OS and software versions, . . .
 - (7) communication between processors for parallel computing

(2) The TaskSpaces framework for grid computing

two major design choices:

(A) use the Java language

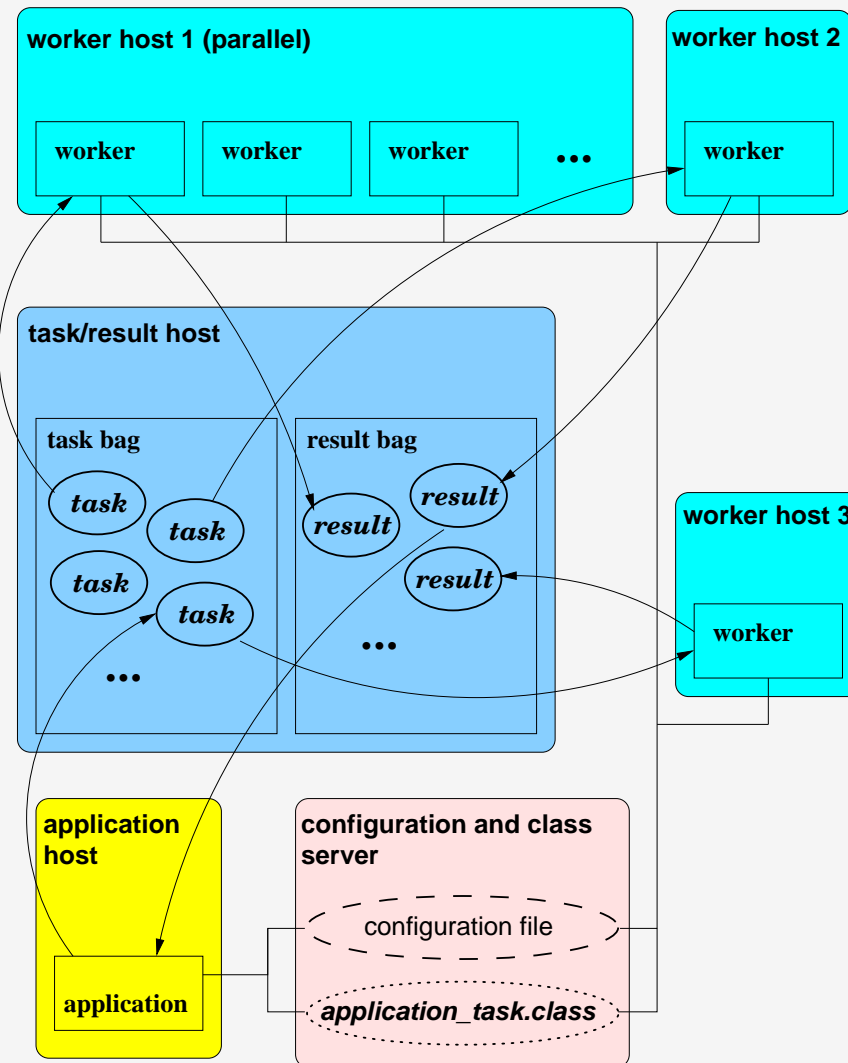
- concept of ‘virtual machine’:
 - behaves the same on all OS, hardware
 - ‘executable byte-code’ fully interchangeable
- has built-in security support
- object = data + code
- has built-in networking

⇒ provides natural and adequate answers to

- (1) standard interface
- (3) secure resource sharing
- (6) install and compile application code on compute engines:
not needed! → download task objects = code + data
- (7) communication between processors: send objects

TaskSpaces: design choices

(B) use tuple space concept, bag-of-tasks paradigm



- space and time decoupling

- worker executable < 2 kb

⇒ provides natural and adequate answers to

(2) scalable:

user: concurrent applications in one or several bags

producer: switches in extra 'worker farms'

(5) resource allocation,

scheduling: task bag is simple decoupled 'superqueue'

high throughput grid experiment

Blue Horizon, SDSC, San Diego, CA (4 workers/processor)	64	128	240
P4 Linux, CU Boulder, CO (2 workers/processor)	4	4	4
Itanium Linux, CU Boulder, CO (2 workers/processor)	4	4	4
forseti1, NCSA, Urbana, IL (1 worker/processor)	16	16	16
hermod, NCSA, Urbana, IL (1 worker/processor)	16	16	16
Total number of workers	104	168	280
Total execution time	105s	103s	101s

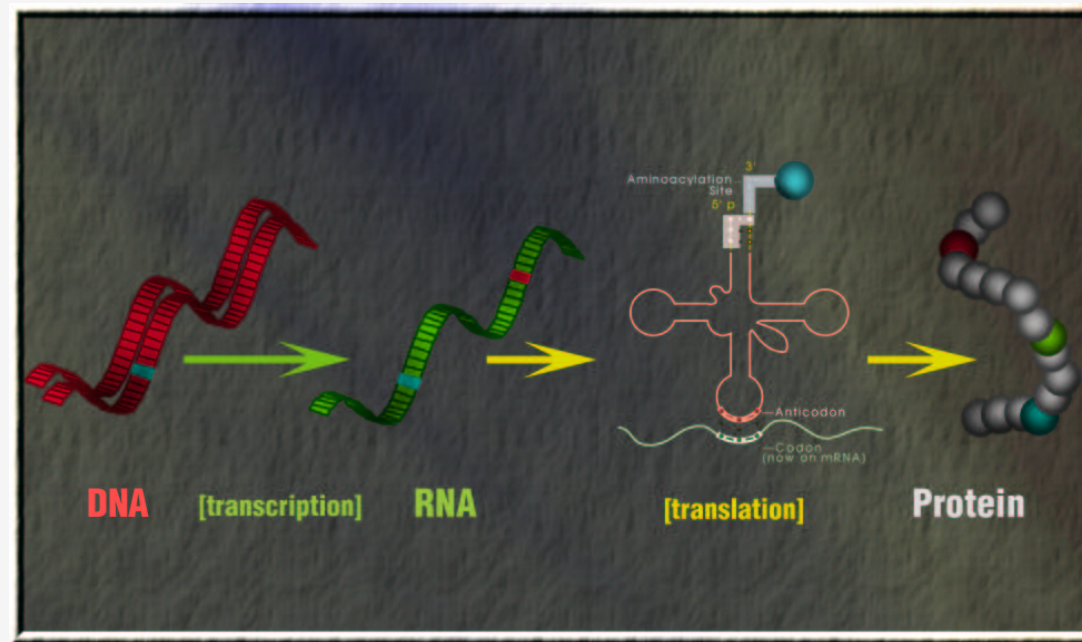
High throughput grid experiment (50 Jacobi iterations, 500x500 grid points per worker).
Number of worker processes and total execution times are shown. Problem size is constant per worker process.

⇒ **TaskSpaces framework: parallel scientific computing with intertask communication is scalable on the internet!**

TaskSpaces grid computing framework

- (1) standard interface: Java task object
- (2) scalable: through bag-of-tasks
 - user: many applications at the same time
 - producer: switches in extra 'computing plants'
- (3) secure resource sharing: digital certificates, to be implemented
- (4) 'transactions', fault-tolerance: rudimentary, needs more thinking
- (5) resource allocation, scheduling: task bag = decoupled superqueue, rudimentary
- (6) get application code on compute engines: download task object
- (7) interprocessor communication for parallel computing: send objects
- (8) user gets charged according to use: to be implemented
- (9) Quality of Service
- (10) resource discovery

(3) Application: finding correctly folded active RNA motifs



- modern cells mostly use RNA molecules as *passive message*, but they also use RNA as an *active catalyst*
 - example: Hammerhead Ribozyme catalyzes cleavage of RNA
- RNA molecules that bind to specific molecules can experimentally be selected (SELEX)
 - example: Isoleucine Aptamer binds to Isoleucine aminoacid

chemically active motifs

- chemical activity characterized by specific *motif*
- *motif* has several *modules* with specified paired bases
- the *modules* are separated by arbitrary *spacer*

Isoleucine Aptamers:

```
gggagaggauacuacacgug-----AUCGUCGCGUUAUUGGGGUUCCACUAUACCAGGUUCU [...]  
gggagaggauacuacacgug-----GAAUCUCGCGUUAUUGGGGUGUCCUUAUCACUUUUGCU [...]  
gggagaggauacuacacgugUGGCUGUUCUCCGCGUUAUUGGGGUUACCUCUCAACUCUGCU [...]  
gggagaggauacuacacgug-----UACACGUUAUUGGGGAUCUACUACUCGCUACAGU [...]  
gggagaggauacuacacgug-UAGUGACAUCGCUUUCUAUUGGGGAUCGUACUUACGACGUU-U [...]
```

1 2

5' ncu ac nn
3' NGG U NN
G U U A U
2

Hammerhead Ribozymes:

```
[137] CCGCGAACCA CUGAUGA GUCGACG-----CGUCGAC GAAA GUAGAAAUUC [...]  
[ 65] GGAUUGGCCA CUGAUGA UUCUCACCCU [57] UGACUUAGAA GAAA GUAGUAAGAG [...]  
[ 71] ACGGGGUCCA CUGAUGA AUUCUGCGGG [45] ACCGCAGAAA GAAA GUAGUGUAAA [...]  
[139] GUUAGGCCA CUGAUGA G-----U-----UGAAA CGUCACACCU [...]  
[146] GGACUGUCCA CUGAUGA CGUCGAUACA----CCAUCGAUA GAAA GUAGUGAAGA [...]  
[ 87] UUUGGGUUCA CUGAUGA AAGGUGAAU [15] AGUGCACACG GAAA GUAGUGUUAG [...]  
[ 63] AGGCUUUACU CUGAUGA GCCAAC-----C-----CAUGGC GAAA CUAGGUUCUA [...]
```

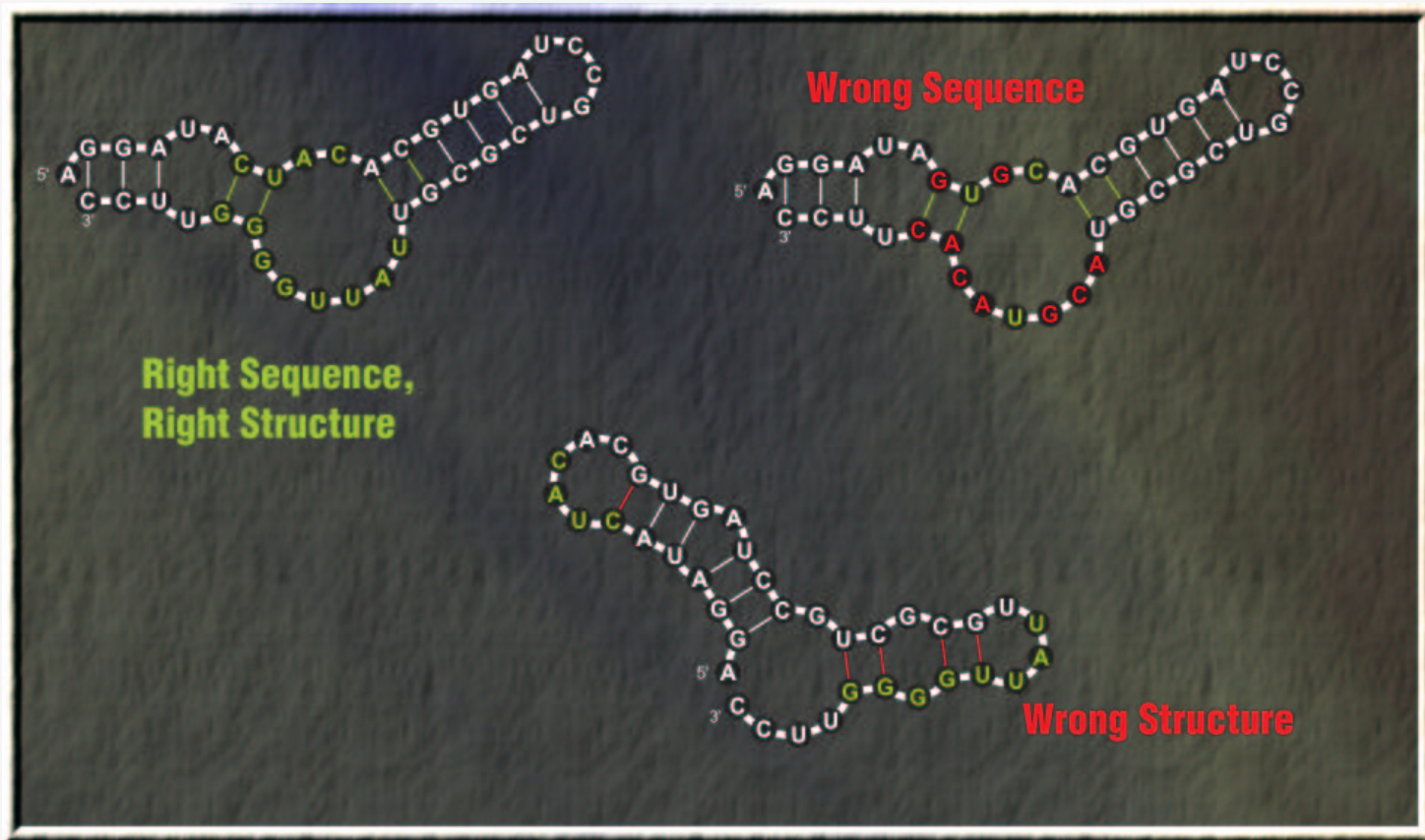
2 3

5' nnnnn u x nnnn
3' NNNNN NNNN
A A C U G A
3 A G A N N 2
C N N N N N N

spacer module spacer module spacer

sequence and structure

- right sequence and right structure are both important



(Isoleucine aptamer)

finding correctly folded active RNA motifs

QUESTION:

given a random RNA molecule of a certain length (e.g. 50 nucleotides), what is the probability that this RNA molecule

- contains the (Isoleucine aptamer) motif in the sequence

and

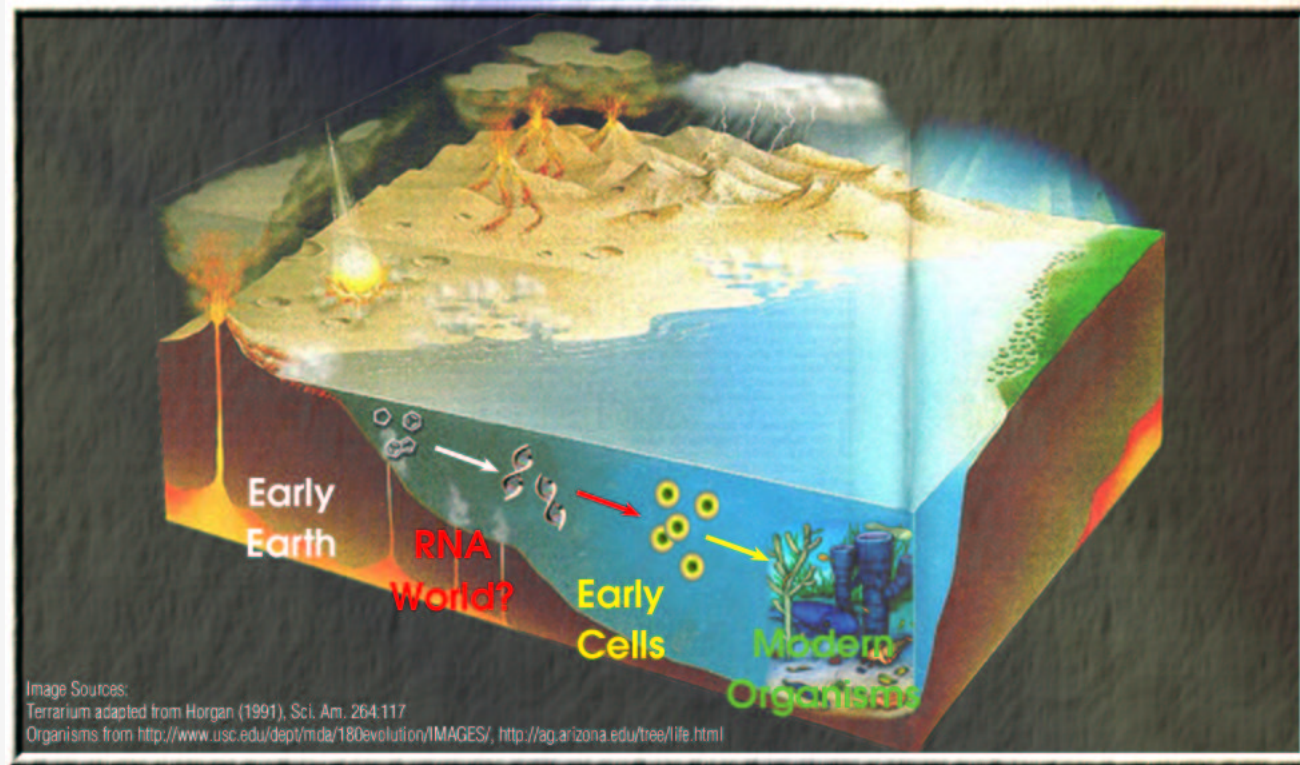
- folds in the correct (Isoleucine aptamer) structure

- applications:

- (1) improve efficiency of SELEX experiments by biasing composition of random pool

- (2) how likely is origin of an RNA world from a small number of random RNA molecules?

application: RNA world



if a small number of random RNA molecules has a reasonable probability of containing molecules with various chemical functions, then a primitive metabolism may have originated from random RNA molecules

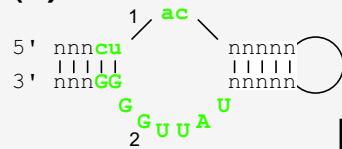
calculation of probability

(a) Sequences

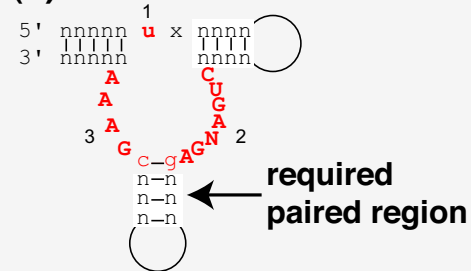


identify sites

(b) Isoleucine



(c) Hammerhead

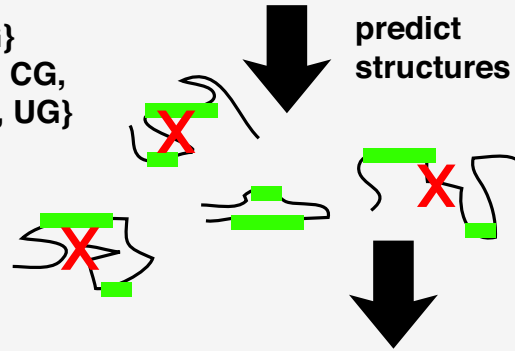


make random sequences that contain sites:

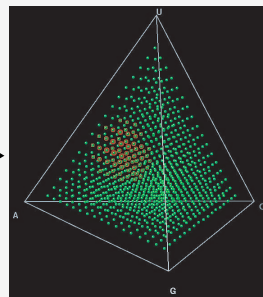
unpaired $\in \{U, C, A, G\}$
 paired $\in \{UA, AU, CG, GC, GU, UG\}$



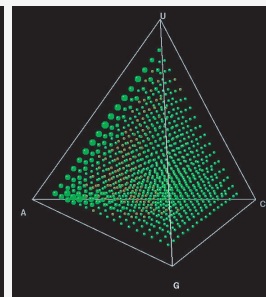
predict structures



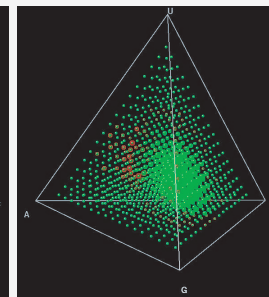
(d) Pr(seq)



(e) Pr(struct|seq)



(f) Pr(struct,seq)

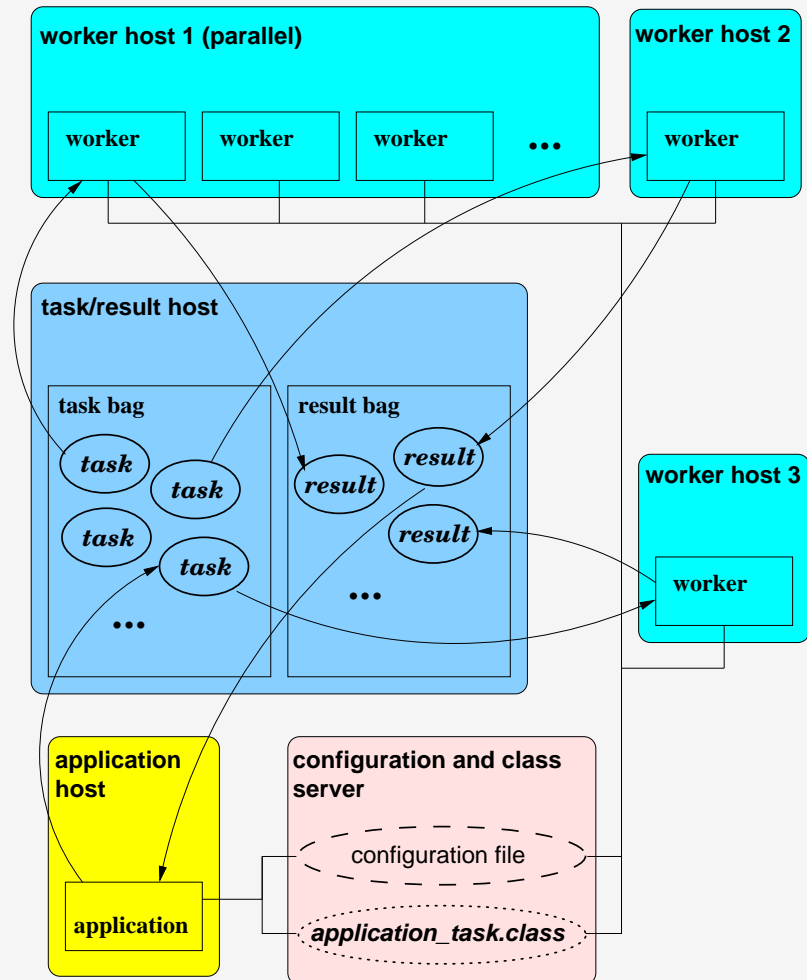


use TaskSpaces

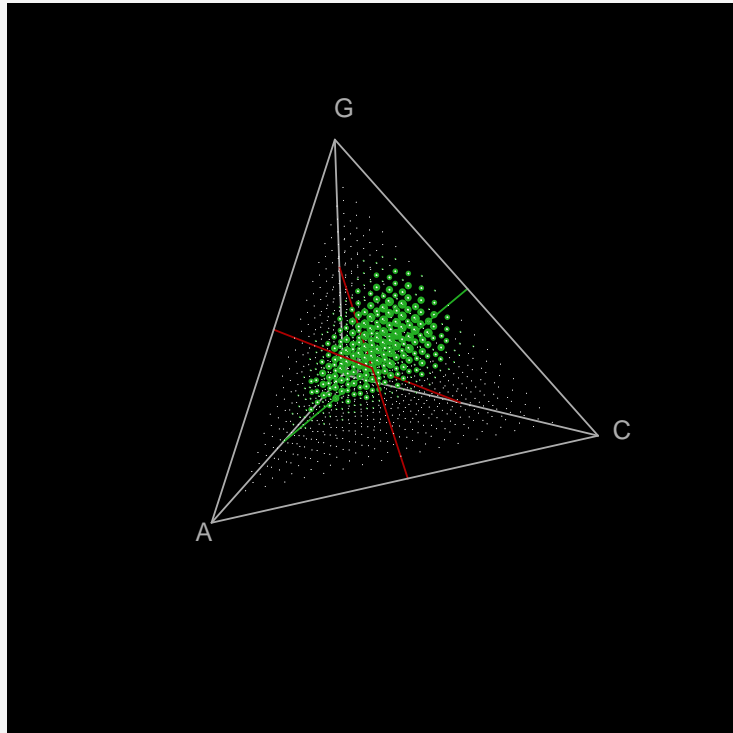
- example of one task = fold 10,000 randomized RNAs of length 50 that contain Isoleucine aptamer motif sequence, and count how many fold in the right structure
- use Vienna folding program
 - written in C
 - called from Java TaskSpaces framework on every processor
 - C needs to be compiled on every architecture
 - C executable can be downloaded with Java executable
- no communication between subtasks, so use TaskSpaces in taskfarming mode

TaskSpaces: example run

- 2% composition steps
(=18,424 different compositions, tasks)
50 nucleotides length
2000 random sequences per composition
→ 73,696,000 foldings
- use Linux Cluster Platinum at NCSA (Illinois)
~ 1000 processors
- use 260 processors, not all concurrently
600 CPU hours
(25 days on one CPU)



$$P(\text{sequence and structure}) = P(\text{sequence}) \cdot P(\text{structure}|\text{sequence})$$



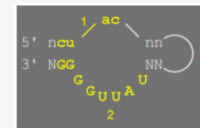
results for length 100

- Ile best at (15%, 25%, 35%, 25%) ACGU composition, factor 3.3 better than uniform composition
- HH best at (35%, 10%, 25%, 30%) ACGU composition, factor 1.4 better than uniform composition

- RNA world: $4.5 \cdot 10^9$ random 100-mers needed for Ile
- RNA world: $2.1 \cdot 10^{10}$ random 100-mers needed for HH

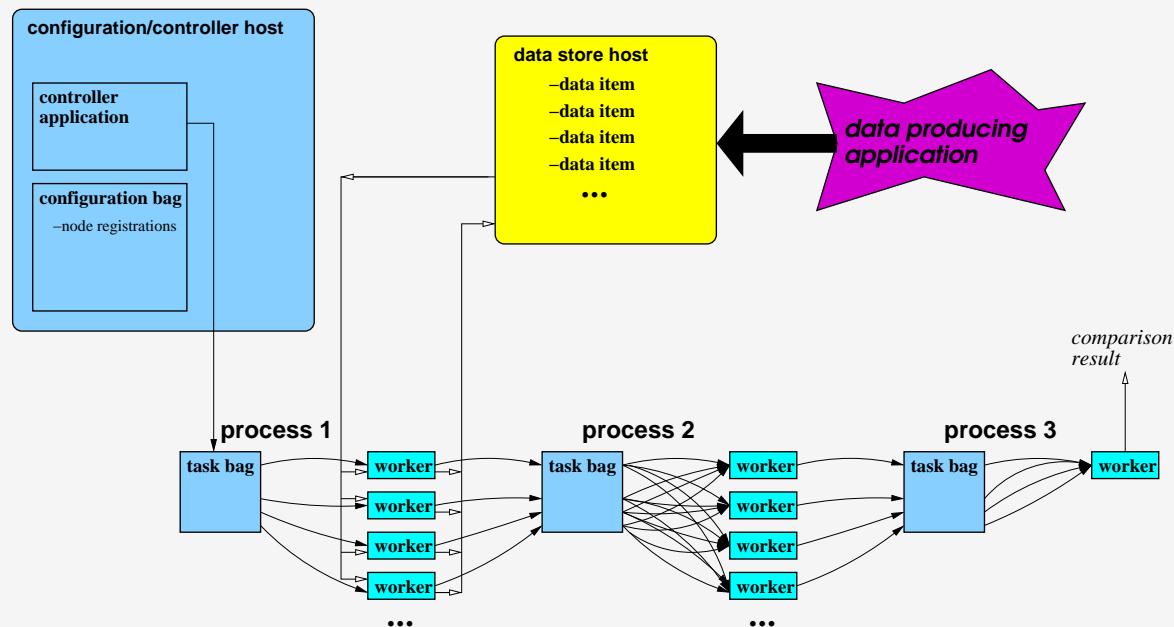
(4) Conclusions and future work

- conclusions:
 - TaskSpaces framework has been developed for parallel scientific computing on computational grids
 - TaskSpaces has been applied to parallel bioinformatics problem: finding correctly folded active RNA motifs



Conclusions and future work

- future work:
 - other parallel bioinformatics applications for TaskSpaces
 - extend TaskSpaces
 - parallel workflows for data processing
 - possible application: proteomics processing



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