Qurry: A prototype quantum programming language

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Abstract—The core philosophy of Ourry is that simple language features, in aggregate, can make quantum programming significantly easier, by offering lightweight abstractions in the spirit of modern C++. This allows users to implement quantum algorithms cleanly, without glossing over necessary lower-level details of quantum computing, such as the abstract topology of a particular computer. Qurry takes a top-down approach, moving from the abstract goal of creating higher-order functions and datatypes to the level of individual language features. While the desired semantics of a true quantum programming language are not yet completely crystalized, the creation of simple language abstractions will elevate the level at which quantum programs are thought about, potentially enlightening the creation of a true quantum programming language. Many existing quantum programming languages are truly circuit languages, with few features above the level of gates, and sometimes rudimentary functions or macros, and there is currently a tendency not to stray very far from this model. Lastly, Qurry is not just a language, but a software ecosystem which is meant to catalyze the development of quantum programming languages.

Index Terms—Quantum Computing, Programming Languages

I. Introduction

NNOVATION in near-term quantum programming requires the use of lightweight abstractions. Lightweight abstractions, as defined by Bjarne Stroustrup [TODO: Cite], are abstractions that lower the cognitive load on the user, without sacrificing understanding of the underlying processes behind particular code. Layers of abstraction are a fundamental idea in all of computer science, and quantum computing is no different. Currently, quantum computing operates on the abstraction that is the gate-level, where programs are defined by gates acting sequentially on particular qubits, instead of, for instance, specific microwave pulses (or another implementationspecific low-level mechanism). The universal gate model of quantum computing generally allows a quantum programmer to ignore many details of the quantum computer they are running on: Sources of error aside, modeling the bond energy of molecular hydrogen is hypothetically the same on an iontrap quantum computer as on a superconducting quantum computer. However, some have argued for the importance of hardware, as in Google's phrase "hardware aware, not hardware agnostic". [TODO: Cite] Many aspects of hardware are particularly important, for instance, topology, which will potentially result in a programmer needing to modify a quantum algorithm for it to run on two separate computers. The goal of lightweight abstractions is to preserve these concerns, while still taking burden off of the programmer, and giving them a richer language with which to express themselves.

Quantum computing [motivations]

[TODO: General motivations, cite Rigetti, cite Haskell/LISP/Rich Hickey, cite Probabilistic programming languages] [Cite Peter Selinger's papers]

At time of writing, Rigetti pyquil contains the following quantum gates and operations:

Single qubit gates and operations:

RESET, I, X, Y, Z, H, S, T

Qubit gates taking an angle as the the parameter and qubit as the second:

• RX, RY, RZ, PHASE

Swap operators, where each takes two qubits, and PSWAP takes an additional angle as a first argument:

· SWAP, ISWAP, and PSWAP

Controlled operators:

- CZ, CNOT ['control', 'target']
- CSWAP ['control', 'target_1', 'target_2']
- CPHASE00, CPHASE01, CPHASE10, CPHASE ['angle', 'control', 'target']

And of course the hybrid measurement instruction, which takes a qubit as the first argument, and a classical register as the second:

• MEASURE

And also contains the following classical operations:

- TRUE, FALSE, NOT, NEG ['classical_reg']
- AND, OR, MOVE, EXCHANGE, IOR, XOR ['classical_reg1', 'classical_reg2']
- ADD, SUB, MUL, DIV ['classical_reg', 'right']
- EQ, GT, GE, LE, LT ['classical_reg1', 'classical_reg2', 'classical_reg3']
- LOAD ['target_reg', 'region_name', 'offset_reg']
- STORE ['region_name', 'offset_reg', 'source']
- CONVERT ['classical_reg1', 'classical_reg2']

Lastly, the matrix operator model of quantum computing actually lends itself to functional programming paradigms quite nicely, because quantum programs and quantum operators are functions in a sense. Additionally, since quantum states are fixed once measured, and are generally measured at the end of a program, in a sense memory is not truly mutated (even though it appears to be). [i.e. result = Pv, not result = 0; Pv]. A quantum program itself is simply a higher order function, which operates on an initial state vector. In turn, a particular quantum program is composed further of simpler matrix-functions, which operate on their own vectors, or are composed with other matrices. For instance, consider a bell state program. As a whole, we may call the bell state program which creates the + state B, and know that B0[TODO: Dirac notation] is the application of the program B to a two-element zero vector. However, this program will further be composed as a Hadamard operator, H, and entanglement operator, CNOT, where H will operate on one qubit, and then CNOT will operate on both qubits. [TODO: Math writeup]

In a traditional circuit language, these operators are composed by simply listing which qubits they operate on, and ordering them correctly in a circuit definition file. However, with higher-order functions, quantum operators can be composed in mariad helpful ways, as is common in function languages like Haskell or LISP (from which Qurry draws many influences).

Classical probabilistic programming languages are a recent innovation from the MIT cognitive science community. Essentially, they create a way for non-expert programmers to access the power of Bayesian inference. Users can create simple probabilistic models in standard code, and then run them through an expert-created inference backend. Famously, this has resulted in dramatically reduced code complexity, with a famous case where a 50-line probabilistic program could compete with traditional approaches to face recognition [TODO: Cite].

In addition to being a prototype quantum programming language, Qurry defines a software stack surrounding the language, which is intended to make development more pleasant. For instance, this software stack makes it exceptionally easy to add new language features and libraries to Qurry. This allows one to rapidly test new ideas in quantum programming and let the language evolve on its own as opposed to architecting a top-down "perfect" language.

Since quantum computers are simply special probabilistic computers, Qurry also attempts to create a classical statistical library for high-level modeling. This is particularly useful in the same way that a classical probabilistic programming language is, namely for modeling anything statistical, and especially for bayesian machine learning. For instance, the R. Tucci and H. Dekant's group have shown uses for this through their software, Bayesforge [TODO: Cite]. Qurry includes simple statistical packages for creating states, but no inference engine. [However, Qurry might allow one to interface with Bayesforge]

II. FEATURES

Qurry as a language is simply a circuit language with an overlay of higher-order functions. [Explain circuit languages and the functionality Qurry includes here, through pyquil]

A. Higher Order Functions

The simplest overlay is quite trivial. It is the *map* function, which allows an arbitrary quantum operator to be applied to several qubits. For instance, the operation (*mapHmyqubits*) does a hadamard state preparation, which is common in the beginning of some quantum algorithms. While simple, this example is important because *map* is a higher-order function. Namely, *map* takes two arguments: first, a linear operator on qubits, and second, a block of qubits where the linear operator is applied to each qubit in the block. Because this first argument is itself a linear operator (a function of sorts [TODO: Verify]), *map* is a higher-order function. Higher-order functions are already used implicitly when composing linear

operators to create a quantum program, so it makes sense to expose them. [Discuss the map construct].

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Additionally, many constructs are defined in terms of unitary gates. Any of these is also a higher-order function. A still simple but more useful example is a controlled unitary. First, consider a unitary that is controlled from one bit:

- cascade
- clear
- cond
- do
- macro

B. Higher Order Datatypes

[discuss define [TODO: memory model]]

Equally important is the ability to create higher-order datatypes. Instead of operating on the level of qubits, are able to arrange raw datatypes into more complex structures. The simplest, which is common in any quantum circuit language, is the ability to create arrays of qubits. [Discuss the *block* construct].

However, for full completeness, one must be able to arrange qubits into special structures, in the sense of C++ structs. Further, in object oriented programming, classes are essentially equivalent to structs, and then collections of functions that operate on a struct. For simplicity, elegance, and robustness, Qurry does not implement encapsulation or inheritance, but instead uses public access by default (in the spirit of Python, since after all, Qurry has a Python interface), and relies on composition instead of inheritance. [Discuss the *datatype* construct]. With the inclusion of composable data structures, users are able to create arbitrarily nested structures of qubits with semantically named fields.

Recursively composed higher-order datatypes, in combination with recursively composed higher-order functions are the foundation for creating a more abstract programming language.

[TODO: Elaborate on everything currently in lists]

III. STANDARD LIBRARY

Qurry contains mechanisms which enable easy inclusion of qurry code in the form of libraries. As an example, Qurry's standard library is implemented in this fashion.

Explain how the statistics library can be easily implemented. At time of writing, Qurry contains the following constructs:

- gaussian
- bernoulli
- multinomial
- uniform

Similarly to in a classical probabilistic programming language, these enable the creation of classical probabilistic states, which can then be used in quantum programs. For instance, it is possible to create a multi-dimensional gaussian distribution, and then entangle an auxillary qubit with the state of the gaussian distribution.

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IV. EXTENSION

Making additions to Qurry is particularly easy: For instance, the map feature is defined using the following python code:

```
from ..compiler.utils import named_uuid
def map(operator, blockname, kernel=None):
    Apply a single-qubit operator to every
    (map H blocka)
    try:
    except KeyError:
        raise ValueError ('The block {} is not 0.3 left in 23d', 'Or 34' mano (in lamble))
```

V. COMPARISONS

Draw examples from Nielsen and Chuang, and the general literature.

Simply change (if (condition) (branch) (branch)) into " Condition Measure [] Jump label branch branch etc "

Curry also supports variable naming, blocks of qubits, classical callbacks, imports, ... Curry can be called as a library and operated from python

There are some easy targets for providing abstraction: common things like functions, conditionals, loops, integer data types, and so on. However, let's jump into the quantum/probabilistic side of things.

Models will fundamentally be composed of, generally, wave functions: Superpositions over all possible states. First, consider modeling a classical distribution. We can successfully produce sampleable classical distributions on a quantum computer. For instance, consider the following model from the Church programming language tutorial. This code is specifying a probabilistic grammar for simple sentences about cooking.

"scheme (define (transition nonterminal) (case nonterminal (('D) (multinomial(list (list (terminal 'the)) (list (terminal 'a))) (list (/ 1 2) (/ 1 2)))) (('N) (multinomial (list (list (terminal 'chef)) (list (terminal 'soup)) (list (terminal 'omelet))) (list (/ 1 3) (/ 1 3) (/ 1 3)))) (('V) (multinomial (list (list (terminal 'cooks)) (list (terminal 'works))) (list (/ 1 2) (/ 1 2)))) (('A) (multinomial (list (list (terminal 'diligently))) (list (/ 1 1)))) (('AP) (multinomial (list (list 'A)) (list (/ 1 1)))) (('NP) (multinomial (list (list 'D 'N)) (list (/ 1 1)))) (('VP) (multinomial (list (list 'V 'AP) (list 'V 'NP)) (list (/ 1 2) (/ 1 2)))) (('S) (multinomial (list (list 'NP 'VP)) (list (/ 1 1)))) (else 'error)))

More succinctly, this is specifying the following (toy) language model: "scheme D(eterminer): (uniform 'the' 'a') N(oun): (uniform 'chef' 'omelet' 'soup') V(erb): (uniform 'cooks' 'works') A(dverb): (uniform 'diligently') AP(Adverb Phrase): (uniform A) NP(Noun Phrase): (D, N) VP(Verb Phrase): (uniform (V AP) (V NP)) S(entence): (NP, VP) ""

To make things even simpler, let's first just consider modeling a randomly sampled Noun-Phrase (which is the first part in sampling a full toy sentence). The noun-phrase is a concatenation of a determiner and a noun. In our toy example, we have two determiners and three nouns, both uniformly sampled, which makes for a total of six options with equal probability. So, we'll need three qubits to model this. Curry has builtins for these distributions. "scheme (def determinerquibit 0) i (defination qubits 12) (bernoulli 0.5 determiner-qubit) (multinomial 0.33 0.33 0.34 noun-qubits) ""

The output is the following (using a local simulator): "" grid curry: ./compile examples/test.lisp

block = kernel.definitions[blockname] [['def', 'determiner-qubit', '0'], ['def', 'noun-qubits', '1', '2'], ['bernoulli', '0.5', 'determiner-qubit'], ['multinomial',

return '\n'.join('{} {}'.format(operator, '000': 0.17, '001': 0.16, '010': 0.17, '011': 0.17, '100': for i in range (block.start, bloc@.16ndO4':10,)6

277.4035930633545 ms simulated runtime ""

In our output, the rightmost bit is representing the determiner, and the other two bits are representing the noun. So the output is: "'python3 'the chef': 1/6, 'a chef': 1/6, 'the omelet': 1/6, 'a omelet': 1/6, 'the soup': 1/6, 'a soup': 1/6 " Now, let's consider the rest of the model. When we sample a Verb Phrase, it contains recursive elements. So, it will branch (with equal probabilities) to either (V AP) or (V NP). Before diving in, let's look at branching in quantum computers.

Consider preparing a bell state: " (h 0) (cnot 0 1) " And distinguish this from the following, which will produce the same classical measurements, but no entanglement (because the state of the first qubit is known before producing the state in the second qubit). In this case, the state 01 is possible, because the first qubit may be measured in the 1 state, and the second qubit is unprepared, and in the zero state. "" (bernoulli $0.5\ 0$) (measure 0 0) (if 0 (x 1) (nop)) ""

So, when creating a probabilistic model which branches, we distinguish between these two types of branching, because only one truly creates an entangled state. However, this makes representing information slightly more difficult, because we will not know which bits correspond to which states (unless we encode this, which we will).

VI. CONCLUSION

In the creation of a Qurry and its corresponding framework, it is hoped that this will aid the development of quantum algorithms, as algorithm designers will have a new, richer, more abstract vocabulary with which to express themselves. To recap, this goal is approached in the following N ways. By introduction of lightweight abstractions from the C++ school of thought, efficient and transparent programming interfaces are created. Through specialized libraries, Qurry can claim to be a generalized library, while still offering powerful subframeworks for specific tasks. With functional programming paradigms, Qurry can move towards higher levels of abstraction as the semantics of quantum programming become better understood. Lastly, by creating a rapid prototyping framework, new language features can be developed in a bottom-up style, which will allow Qurry to be created naturally, instead of artificially.

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APPENDIX A APPENDIX ONE

Appendix content

ACKNOWLEDGMENT

The author would like to thank Dr. Will Zeng of Rigetti computing, an organizer of the Unitary Fund, Dr. Ajay Bansal of Arizona State University, PLoS and other donators to the unitary fund, and ASU's FURI program.

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