Abstract—As computer systems grow in size and complexity, their integration, while a necessity, becomes more difficult. Service oriented architectures and middleware systems in general deal with the issue by decoupling the components. However, these architectures are still designed and enacted from a centralised perspective: a single process invokes remote services, unaware of being part of a larger, more complex workflow. We claim that the orchestration-based approach does not scale well with increasing complexity and heterogeneity of the components, such as those required in the enactment of medical guidelines and workflows.

Medical guidelines encode different aspects of a medical procedures, and they specify different level of abstraction in the procedure. They have to be adapted to the realities of different clinics and hospitals, to advances in knowledge, and to the changing of available resources within an institution. We address the problem of representing and enacting medical guidelines using a fully distributed approach. The framework provided by OpenKnowledge, based on sharing choreographies among actors, allows the representation and enactment of the coordination aspect of guidelines and the discovery of the medical knowledge provided by the distributed actors.

Index terms—service oriented architecture, semantic service composition, service choreography, health informatics

I. INTRODUCTION

Software systems are getting more and more complex. An important source of the complexity is the need to integrate different, heterogeneous subsystems. Middleware provides the facilities for connecting software components, reducing their interdependencies and simplifying their reusability in different systems.

Service oriented architectures are part of a wide family of middleware systems: every component exposes services accessible through the network. Complex systems can be pulled together, invoking services belonging to different and possibly external systems using workflow languages [4] like the Business Process Execution Language (BPEL) [2] or Yet Another Workflow Language (YAWL) [33]. These frameworks are based on a centralised, imperative paradigm: a central process controls everything, the other services are passive, unaware of being part of a larger, more complex workflow. In this approach, called orchestration, services are usually wired together into an application at design time, leaving little margin if, for example, one of the services is unavailable at run-time.

We claim that this approach does not scale well with the growing complexity of systems. This is the case, as we will see in the next sections, for medical procedures, where complex workflows, developed by committees, involve the interaction of many different actors such as desktop applications, web services, databases, diagnostic devices and monitoring tools and have to be adapted to the contingent and varying realities of hospitals and clinics. We advocate a different paradigm, based on sharing choreographies among actors. We believe that both design and execution of complex systems can benefit from this approach. At design-time, the choreography paradigm forces the developers to think in terms of the actors, their roles and their interactions, making them explicit and abstracting away from the details of their specific activities. While the general approach is to share description of services, and pull them together at design-time into a specific application, our approach takes the opposite direction and shares the choreographies. At execution-time, shared choreographies are the contracts that actors can search, verify and agree to follow: the binding with actors takes place at run-time, not at design-time.

The OpenKnowledge1 project has allowed us to develop a fully distributed, open, peer-to-peer framework, focussed on shared interaction models that are executed by peers. In this paper we focus on the application of this framework to the coordination of medical guidelines, using as a case study the assessment procedure (called triple assessment) followed by a patient suspected of breast cancer.

In Section II we introduce the problem requirements of formalising and enacting medical guidelines, exemplified by the breast cancer assessment case study in Section III. Then in Section IV we discuss the advantages of a choreography-based approach to the problem, against an orchestration-based one. In Section V-B we cover OpenKnowledge, as a flexible, choreography-based framework that can address some of the requirements of medical guidelines. The evaluation of the framework is presented in Subsection V-C. Finally other approaches to choreography are briefly reviewed in Section VI.

II. MEDICAL GUIDELINES

Gaps between medical theory and clinical practice are consistently found in health service research. Care procedures can differ significantly between different health centres, with varying outcomes for patients, and medical errors could be avoided if standard procedures were followed consistently. One of the causes of discrepancies in care is the difficulty in distributing and sharing efficiently the large volume of

1http://www.openk.org
information continuously produced by medical research. These issues have pushed the development of clinical practice guidelines: several studies [19] have shown that published guidelines can improve the quality of care.

Guidelines are usually defined by a committee of experts and are provided as booklets, often hundreds of pages long, covering relatively narrow fields or specific pathologies. Hundreds of guidelines are available, and generalist doctors are expected to be aware of, and follow, the guidelines relevant to each patient. The result is that guidelines are rarely followed, and inconsistencies in medical treatments are not reduced as much as hoped.

Information technology can improve the situation. Many clinical guidelines provide informal descriptions of workflows and rules that can be translated into formal, machine-executable, representations. Research has suggested that computerised clinical supports can improve practitioner compliance with guidelines [17].

Guidelines encode at least two separate types of knowledge: they specify the overall coordination between actors, both clinical and non-clinical (for example, the results of all the exams are sent to a multidisciplinary team for discussion and to the Electronic Health Record of the patient) and specific medical knowledge required in taking decisions (for example, dosing drugs or classifying a melanoma). Guidelines are developed for general adoption: the coordination specifications need to be adapted to the contingent realities of different institutions and clinics, and the medical knowledge need to be kept updated.

Finally, different guidelines specify medical procedures at different level of abstraction, from the general framework defining the milestones of screening, intervention and follow-up within the national health system, to the detail of the dosing of a specific drug. Depending on the condition and on the abstraction level, there can be varying degrees of freedom in the selection of the guideline, or in the selection of participating clinicians. For example, a woman can choose among several paths of care for her pregnancy, while the (more critical) procedure for a heart attack is more stringent. In order to enact the full pathway a patient has to go through the various specifications that should be integrated, selecting at each step the best one, and adapting them to the contingencies.

In the work described in this paper we use OpenKnowledge, a technology developed for distributed peer-to-peer systems, to represent, integrate and enact the coordination aspect of guidelines offering a level of flexibility that improves portability between different institutions.

Before the details of OpenKnowledge, we first introduce out case study and discuss the differences between the centralised and the distributed approaches.

III. CASE STUDY: ASSESSMENT FOR BREAST CANCER

We present as our case study the guideline for assessing the presence of breast cancer, because, as we will describe in more detail in the next section, a computer-based, centralised workflow has already been developed, making the comparison easier. This guideline is part of a wider workflow, that includes periodical screening, surgical intervention and follow-up.

Breast cancer is the most commonly diagnosed cancer in women, accounting for about 30% of all such cancers. One in nine women will develop breast cancer at some point in their lives. In the UK, women with symptoms that raise suspicion of breast cancer are referred by their GP to designated breast clinics. To increase the accuracy of diagnosis, a combination of clinical examination, imaging and biopsy - known together as triple assessment - is recommended.

The first element of triple assessment consists of gathering the patient details and clinical examinations, and it is completed by a breast surgeon. If the clinical examination reveals an abnormality then the patient is referred to a radiologist for imaging (either ultrasound, mammography or both). If either the examination or imaging findings warrant it, then a biopsy is performed, either by a radiologist or a surgeon, and the tissue is sent to a pathologist for examination. The collective results from all three tests influence the final management of the patient. Depending on the resources available, different imaging and pathology laboratories might be selected every time the guideline is executed, possibly from different institutions.

A small number of ‘worried well’ patients may not qualify for either imaging or biopsy and could be discharged straight away. As the entire clinical process is distributed among three different disciplines and involves a number of different clinicians, a very close co-ordination and good communication between those involved is essential for the smooth running of the clinic.

IV. CENTRALISED AND DISTRIBUTED MODELS

The triple assessment model presented in [24] is designed according to a centralised principle, and the abstract workflow is shown in Figure 1. The centralised model has been implemented\(^2\) in PROforma [29], a process modelling language developed within Cancer Research UK. Together with the possibility of representing plans, sub-plans and actions, its strength lays in its decision support system, based on argumentation.

However, in PROforma it is not possible to explicitly represent different roles in a guideline: roles are enforced by requiring different permissions to access the activities. Activities are queued in the todo list of a user, who finds it after log in. When the user finishes the activity, it is marked as completed and the workflow proceeds. While PROforma provides a strong support for representing and reasoning about medical knowledge, it is currently inadequate for representing and executing coordinated activities between distributed actors.

The distributed nature of the procedure cannot be reconstructed from its representation as a centralised workflow. Moreover, medical knowledge specific to different participants, that is, the arguments and rules necessary for decisions in various phases, is centralised in a single procedure, making it impossible to reuse the same knowledge in different guidelines. If the model is implemented in another workflow language such as BPEL, the knowledge could be split amongst a group of services, possibly each based on PROforma, but

\(^2\)A demo is available at: http://tinyurl.com/tripleassessment
Figure 1. Centralised representation of the triple assessment.

Figure 2. Activity diagram for the triple assessment

Figure 3. Fragment of the sequence diagram of a run of the triple assessment
the process itself would still be represented from a single perspective.

A more realistic representation of the flow of the procedure is provided by the UML Activity diagram of Figure 2, in which the participants and their interactions are made explicit. There are five main actors: the breast surgery service (BSS), in charge of the first three activities in the workflow in Figure 1, the breast imaging service (BIS), responsible for the imaging decision and for the two alternative possible examinations, the breast pathology service (BPS), in charge of the biopsy, the multi-disciplinary team (MDT), responsible for the final decision together with the surgery service, and the patient.

The aim of this work is to allow the distribution of knowledge and information to the actors in charge, the reuse of knowledge in different guidelines, and the possibility of easily adapting guidelines to the realities of actual institutions and hospitals. We obtain this by separating the two different aspects of guidelines into two abstraction layers. The higher level consists in the specific knowledge and services, local and shared, and the participants agree to follow it. The lower level consists in the choreography that specifies the expected behaviour interface of the participants. The choreography is shared, and the participants agree to follow it. The lower level consists in the specific knowledge and services, local to participants, that are mapped to the choreography. The choreography provides a specific context and the semantics for the interaction, giving the boundaries of the task of integrating the different participants.

V. THE OPENKNOWLEDGE APPROACH

A. Sharing choreographies

A choreography-based architecture can help developers to think about distributed applications from a different perspective, one which scales better with an increasing number of interacting actors. A distributed application becomes a set of interactions between actors that assume different roles, their internal behaviour is kept separated from their external behaviour.

In most distributed approaches, the workflow engineer selects the participants: the list of participants is a basic construct both in choreography languages such as WS-CDL [21], BPEL4Chor [11] and in orchestration languages such as BPEL [2]. Some workflow architectures, such as Taverna [20], let the developer specify a list of alternative services to call if the first fails, but the binding is still at design time. Service descriptions are stored in shared repositories and the designer looks up the specific service for a task.

This approach makes portability more difficult. As we have seen, guidelines are usually written by a committee of experts with a view to being generally applicable: when the plan has to be implemented by adopting institutions, it may require heavy customisation to adapt it to their requirements (for example, the list of services to invoke is likely to be different, and therefore part of the specification has to be changed).

In our approach, the choreographies are published in a shared repository, similarly to the publication of a guideline by a committee. Participants search the appropriate choreographies when they need to perform an activity that requires the coordinated activity of other actors. The choreographies are matched against the participants’ capabilities: the participants select the choreography that best fit them and advertise their intention to perform a role in them. If all the roles are filled, the interaction can start. In our triple assessment example, it means that roles such as the breast imaging service or the breast pathology service are bound at run time, when the procedure needs to be performed, based on their availability.

We believe that this allows increased reuse both of choreographies and of participants’ components, and allows the deployment of distributed applications with different degrees of freedom, maintaining the same architecture. Without altering the architecture presented in this paper, it is possible to implement a closed system where for a task there is only one possible choreography and a fixed set of other participants that are bound to it, and an open system where for a task a participant has the freedom of choosing different choreographies, and has a choice between different participants who are free to accept.

In the choreography model, issues of heterogeneity and brokering need to be addressed. Participants are likely to be different and they need to understand one another. The same services may be available from many peers, and the search and discovery process can be complex, especially if it needs to be performed at real time. The OpenKnowledge framework deals with these issues.

B. The OpenKnowledge Framework

The OpenKnowledge kernel [28] provides the middleware that assorted services can use to interact using a choreography-based architecture able to deal both with the semantic heterogeneity of the actors and with their discovery. It has been designed with the goals of lightweightness and compactness, allowing a short development cycle for distributed applications.

The framework allows a direct translation of a choreography oriented design into an executable application. The core concept is the shared interaction models (IM), performed by different applications and service providers. These actors are the participants, called peers, of the interactions, and they play roles in them. In an interaction all the roles have equal weight; the behaviour of all the peers and in particular their exchange of messages are specified.

IMs are written in the Lightweight Coordination Calculus (LCC) [26], [27], a compact, executable choreography language based on process calculus. Its syntax is shown in Figure 4. The IMs are published by the authors on the distributed discovery service (DDS) with a keyword-based description [22].

An IM in LCC is a contract that defines the expected, externally observable, behavioural interfaces of the roles that participants can take in the interaction: an IM is a set of role clauses. Participants in an interaction take their entry-role and follow the unfolding of the clause specified using combinations of the sequence operator (’then’) or choice operator (’or’) to connect messages and changes of role. Messages are either outgoing to (’⇒’) or incoming from (’⇐’) another participant in a given role. A participant can take,
Model ::= \{Clause, …\}
Clause ::= Role :: Def
Role ::= a (Type, Id)
Def ::= Role | Message | Def then Def | Def or Def
Message ::= M ⇒ Role | M ⇒ Role ← C | M ⇒ Role | C ← M ⇒ Role
C ::= Constant | P(Term, …) | ¬ C | C ∧ C | C ∨ C
Type ::= Term
Id ::= Constant | Variable
M ::= Term
Term ::= Constant | Variable | P(Term, …)
Constant ::= lower case character sequence or number
Variable ::= upper case character sequence or number

Figure 4. LCC syntax

a(bis,BSS)::
  registration(P) ← a(patient,P) then
  null ← triage(TR,P) and examine(TR,ER,P) then
    null ← isImagingDecision(ER) then
    doImaging(P,TI) ⇒ a(bis,BIS)
    ← typeofImaging(ER,TI)
  or
  (null ← isFreeHandBiopsy(ER)
   then
   null ← doFreeHandBiopsy(P,FHBR))
  or
  null ← true
  then
  null ← isBiopsy(ER) then
  doSampleAnalysis(P) ⇒ a(bps,BPS) then
  doMDM(P,TR,ER,FHBR) ⇒ a(mdt,MDT)
then
a(bis,BIS)::
doImaging(P,TI) ← a(bss,BSS) then
null ← acquire_image(P,TI,IMR) then
image_report(P,IMR) ⇒ a(bss_wait_reports,BSS)
then
  (null ← isBiopsyDecision(IMR) then
   null ← doBiopsy(P,IMR,S) then
   doSampleAnalysis(P) ⇒ a(bps,BPS) then
   doMDM(P,TR,ER,FHBR) ⇒ a(mdt,MDT)
  or
  null ← true

Figure 6. LCC clause for the breast imaging service (BIS) role
during an interaction, more roles and can recursively take
the same role (for example when processing a list). Message
input/output or change of role is controlled by constraints. In
its definition, LCC makes no commitment to the method used
to solve constraints - so different participants might operate
different constraint solvers.

Figures 5 and 6 show the LCC clause for the breast surgery service (BSS) and the breast imaging service (BIS) roles in the
triple assessment procedure described in the activity diagram
of Figure 2. In the surgery service clause, the participant who
takes the bss role starts waiting for the registration message
from a patient. When the request message arrives, it executes
the triage and the examination on the patient. Based on the
result of the examination, the surgery service either asks the
imaging service to acquire an image, or performs a free hand
biopsy, asks the pathologist to analyse the sample, and sends
a request for a multi-disciplinary meeting. Then the process
changes role, and waits for the reports from the contacted
specialists. When all the reports have arrived, the process
returns to the main role and makes the decision about the
patient.

In the imaging service clause, the actor taking the role
starts by waiting for a request doImaging(P,TI) from the
surgery service. Then it acquires and analyses an image using
the method specified in TI (either an ultrasound scan, or a
mammography), and sends the report to the surgery service.
If required it performs a biopsy, asks a pathologist to analyse
the sample and sends a report to the multi-disciplinary team
for discussion.

Most of the constraints, such as triage(P,TR) or
doFreeHandBiopsy(P,FHBR), correspond to external
activities, that a doctor or a radiologist perform. They may be
completed by filling in computerised forms or by receiving
data from an external device such as an ultrasound scan
machine. Some constraints, such as acquire(P,TI,IMR)
may launch further interactions (in this case, a different one
depending on the method required, and involving a radiolo-
gist).

LCC prescribes only the ordering of the exchanged mes-
sages and their pre and post-conditions in the form of con-
straints peers need to solve. However, in OpenKnowledge
it is possible to annotate every element in the IM in order
to enrich the description of the interaction. Annotations are
mainly used to define the ontological type of the variables
in messages and constraints, but time-out for messages and constraints could be expressed using the same mechanism. Figure 7 describes the syntax used in annotations. An example of annotation for the variables in role bss is shown in Figure 8. This specific example shows how variables can be structured terms: the variable P is, in the choreography, the structure patient(name,surname,date_of_birth, address(street,post_code)). The structure is a short-hand for an XML-like schema, where the functions and the parameters are nested tags in a tree. Annotations are always relative to a specific role-clause in the interaction: the scope of a variable is always limited to a clause.

Annotations are conceptually separated from the IM itself. It is possible to attach different annotations to the same IM to adapt it to different requirements and contexts: for example, the same abstract IM could be used in a different community, that specifies patients by their insurance numbers, and therefore annotates variable P differently.

A peer that wants to perform some task, such as providing an imaging service for breast cancer screening, searches for published IMs for the task by sending a keyword-based query to the DDS. The DDS collects the published IMs matching the description (the keywords are extended adding synonyms to improve recall) and sends the list back to the peer, that needs to choose the one to subscribe.

In open systems, IMs and peers may be designed by different entities, and therefore the constraints and the peers’ knowledge bases are unlikely to correspond perfectly. The heterogeneity problem is dealt with in three phases. We have already seen the first phase: the DDS matches the interaction descriptions using a simple query expansion mechanism. Then the peers compare the constraints in the received IMs with their own capabilities [18], and finally the peers need to map the terms appearing in constraints and introduced by other peers [9]. The scope of the matching problem is limited to the specific IM in the second phase, and to the specific interaction run in the third phase.

The peer capabilities are provided by plug-in components, called OKCs (OpenKnowledge Components) that can be published by the developers on the DDS and downloaded by other peers. An OKC exposes a set of Java methods that are compared to the constraints in the IMs. The arguments of methods can be annotated similarly to variables in the IM. Arguments, like variables in constraints, can be structured terms. Figure 8 shows an annotated method with structured arguments. As annotations are short-hands for XML-like trees, values in an argument are accessed by their path, similarly to a very simplified XPath: for example, to obtain the street of a patient, the method doTriage will call the method getValue("/street") of the argument P.

The peer matches the annotated signatures of the constraints and of the methods, transforming them into trees and verifying their distance [15], [18]. The comparison process creates adaptors, that bridge the constraints to the methods, as shown in Figure 9. An adaptor has a confidence level, reflecting the distance between the constraints and the matching method, that gives a measure of how well the peer can execute an interaction, and it used to select the most fitting IM. Once the peer has selected an IM, it subscribes to its role in the discovery service. Figure 10 shows a screenshot of network status when roles in an interaction are subscribed by at least one peer.

When all the roles are filled, the discovery service chooses randomly a peer in the network as coordinator for the interaction, and hands over the IM together with the list of involved peers in order to execute it.

The coordinator first asks each peer to select the peers they want to interact with, forming a mutually compatible group of peers out of the replies. The selection process is subjective to the peers. All the participants receive the list of peers subscribed to all the roles, and they can check the subscriptions, selecting the preferred ones. A peer can also select none of the participants, excluding itself from a particular run of the interaction, for instance due to overload.

The framework provides only an interface for the selection

```java
annotation := @annotation/about, innerAnn @
about := @role(Role)@message(M)
@constraint(Term)
@variable(Variable)
innerAnn := annotation|tree

Figure 7. Annotations syntax

Constraint annotations
@annotation(@role(bss),
@annotation(@variable(TR),
risk_level)
)
@annotation(@role(bss),
@annotation(@variable(P),
patient(name,surname,date_of_birth,
adress(street,post_code)) )
)

Java method annotation
public boolean doTriage(Argument P, Argument TR)
{
...
}

Figure 8. Annotations for the constraint triage(P,TR) and for a corresponding method

```
method, as its implementation is delegated to the application developer. However, different strategies have been tested, as we will see in the evaluation section.

While different implementations are possible, in the current version of the OpenKnowledge kernel the coordinator executes the interaction, instantiating a local proxy for each peer. The remote peers are contacted only to solve constraints in the role they have subscribed.

C. Framework evaluation

The preceding sections have demonstrated how clinical guidelines (typically designed to operate in centralised systems) can be reconstructed in a de-centralised style. Importantly, we have done this without requiring a special-purpose representation language or enactment system - we have used the generic OpenKnowledge kernel system. This means that the mechanisms developed and evaluated for the generic kernel apply, as do the evaluations already performed on those mechanisms. In this section we summarise what we know to date about the scalability and performance of these mechanisms.

Beyond the basic mechanisms of enactment, the key ingredients necessary to build large scale systems of peer to peer choreography using the approach described above are:

- Discovery: It has to be possible to discover interaction models efficiently in the peer network.
- Ontology matching: It has to be possible to adapt to differences in naming the objects described in interactions.
- Reputation: Peers, which may not have interacted with others previously, must have ways of judging with which of the many other peers it might be useful to interact.

We summarise current progress in evaluation on each of these three dimensions. Further details of the evaluative experiments, plus the OpenKnowledge kernel system, can be obtained at www.openk.org.

Discovery: OpenKnowledge relies on keyword matching (via distributed hash tables) for the discovery of interaction models so we were concerned to evaluate whether this sort of mechanism is likely to be effective when interactions are choreographing large populations of Web services. Large collections of interaction models do not yet exist (since this is new technology) but Web service technology is sufficiently mature that there do exist large repositories of Web service interface specifications (harvested from the Web). Since a choreographed collection of services is analogous, itself, to a service then these repositories give the closest approximation currently available to a naturally occurring population. Researchers in Amsterdam performed a large scale evaluation of the OpenKnowledge discovery service operating real-time on a large number of WSDL files [3]. The WSDL corpus (believed to be the largest such corpus in existence at the time) contains 54,245 unique WSDL documents that were obtained during a focused crawl performed in March 2008. 400 instances of the system were run simultaneously, each publishing numerous descriptors so as to give a combined system publishing throughput of one description per millisecond. To bring an element of peer unavailability into the experiments, a proportion of randomly chosen peers were made to fail, without notifying other peers. Each live node made a series of queries and collected the results, so as to give a combined system query throughput of one query per millisecond. Our approach showed significant performance gain, compared to two reference approaches: it either maintained very high recall with four times less messages or used the same low number of messages for a 29% gain in recall.

Ontology matching: Although many choreographies may require no ontology matching (because we want to use interaction models analogously to traditional workflow languages) some interactions can be made more accessible to a broader range of users by allowing ontology matching. The OpenKnowledge kernel therefore allows extensions to perform ontology mapping, allowing different mapping algorithms to be accessed via the kernel as either OpenKnowledge components (if the mapping implementations are available as software components that can be shared) or as services (if the mapping implementations are hosted remotely). We then exploit OpenKnowledge’s use of interaction models to allow different forms of dynamic ontology matching at runtime. These different forms are are: semantic matching of structures within an interaction model; prediction of word use based on statistical analysis of interactions; and ontology mapping via alignment between interaction models. To evaluate these, the OpenKnowledge matching component was applied to two selected test sets: real-world ontologies (we used different versions of the Standard Upper Merged Ontology, SUMO, and the Advanced Knowledge Technologies, AKT, ontologies); and a large-scale synthesized set obtained from real world GIS web services [32]. The results confirm the robustness of the OK matching component (in terms of standard relevance metrics: precision, recall and f-measure) against variation of internal parameters (i.e. thresholds) and probability of alterations. The OpenKnowledge matching plug-in (integrated in the OpenKnowledge kernel) is capable of reproducing results similar to state-of-the-art baseline matchers for syntactic variation while outperforming it when semantic variations are applied.

Reputation: The simplest measure of reputation in service...
choreography is (arguably) the level of successes or failures of peers when involved in interacting with others. This is used by the OpenKnowledge kernel to give basic statistical measures of the reputation of a peer (with respect to its performance in earlier interactions), provided at no cost to the users of the system and giving a baseline mechanism for choosing one peer over another (analogous to the use of page ranks in choosing one page over another in the Web). Other levels of sophistication in trust and reputation can be built on top of this, allowing for diversity in approaches. Similarly to the ontology mapping approach (with which this interacts) specific algorithms for assessment of good enough answers and trust assessment can then be plugged in to extend the basic mechanism. We have evaluated this mechanism in two contrasting domains. In bioinformatics we applied it to matching protein spectra returned from three of the most common proteomics data services. Our reputation mechanism allowed us to rank automatically these services appropriately in terms of the quality of data provided [25]. In emergency response, researchers at Trento applied the mechanism to the trust over peers in the evacuation phase of the simulated emergency, using GIS data from the Trentino region of Italy. Simulations using a centralised plan for response were compared to the same system in which decision making was decentralised (hence more robust to failure of individual components) [31]. It was possible, using our peer ranking system to detect failing peers, to maintain levels of response comparable to a centralised system in the presence of significant levels of disruption in the peer network (simulated by randomly failing a proportion of peers).

The evaluations summarised above do not, of course, predict how effective our system will be in some specific, new application. Evaluation of that requires specific testing with the demands of that application in mind. It does, however, show that the central elements of this style of open choreography have desirable scaling properties that skilled engineers might harness to adapt the general framework to specific applications - for example the healthcare domain of this paper.

VI. RELATED WORK

The majority of workflow research has focused on designing languages for and implementing service orchestrations, from the view of a single participant. Examples can be seen in the Business Process Modelling community through BPEL [30] and life sciences community through Taverna [23]. However, there are relatively few languages targeted specifically at service choreography, the most widely known are:

- **WS-CDL**: The Web Services Choreography Description Language (WS-CDL) is the proposed standard for service choreography, currently at the W3C Candidate Recommendation stage. However, WS-CDL has met criticism [7], [13] through the Web services community. It is not within the scope of this paper to provide a detailed analysis of the constructs of WS-CDL. This research has already been presented [16]. However, it is useful to point out the key criticisms with the language: WS-CDL choreographies are tightly bound to specific WSDL interfaces, WS-CDL has no multi-party support, no agreed formal foundation, no explicit graphical support and few or incomplete implementations.

- **Let’s Dance** [34]: is a language that supports service interaction modelling both from a global and local viewpoint. In a global (or choreography) model, interactions are described from the viewpoint of an ideal observer who oversees all interactions between a set of services. Local models, on the other hand focus on the perspective of a particular service, capturing only those interactions that directly involve it. Using Let’s Dance, a choreography consists of a set of interrelated service interactions which correspond to message exchanges. Communication is performed by an actor playing a role. Interaction is specified using one of three Let’s Dance constructs: precedes – the source interaction can only occur after the target interaction has occurred; inhibits – denotes that after the source interaction has occurred, the target interaction can no longer occur, and finally weak precedes – denotes that the target interaction can only occur after the source interaction has reached a final status, e.g. completed or skipped. A complete overview of the Let’s Dance language is presented in [34], including solutions to the Service Interaction Patterns [8].

- **BPEL4Chor** [12]: is a proposal for adding an additional layer to BPEL to shift its emphasis from an orchestration language to a complete choreography language. BPEL4Chor is a simple, collection of three artifact types: participant behavior descriptions define the control flow dependencies between activities, in particular between communication activities, at a given participant. A participant topology describes the structural aspects of a choreography by specifying participant types, participant references and message links; this serves as the glue between the participant behavior descriptions. Finally participant groundings define the technical configuration details, the choreography becomes Web service specific, concrete links to WSDL definitions and XSD types are established. BPEL4Chor is an effective proposal and importantly conforms to standards [4] by enhancing the industrially supported BPEL specification. BPEL4Chor encourages reuse by only providing a specific Web service mapping in the participant grounding. Furthermore, unknown numbers of participants can be modelled, not possible with WS-CDL.

There are several proposals for extending the Business Process Modelling Notation [1]; the de-facto standard for business process modelling. Although the BPMN allows an engineer to define choreographies through a swimlane concept and a distinction between control flow and message flow, it only provides direct support for a limited set of the Service Interaction Patterns and not some of the more advanced choreography scenarios. [10] introduces a set of extensions for BPMN which facilitate an interaction modelling approach as opposed to modelling interconnected interface behaviour models. Authors claim that choreography designers can understand models more effectively, introduce less errors and build models more efficiently. Evaluation concludes that the majority of the Service Interaction Patterns can be expressed with the additional extensions. [14] discusses the deficiencies of the BPMN for choreography modelling and proposes a number of direct extensions for the BPMN which overcome these limitations.
Hybrid orchestration/choreography approaches exist, in particular the Circulate architecture [6] maintains the robustness and simplicity of centralised orchestration, but facilitates choreography by allowing services to exchange data directly with one another. Performance analysis [5] concludes that a substantial reduction in communication overhead results in a 2–4 fold performance benefit across all workflow patterns.

VII. CONCLUSION AND FUTURE WORK

In this paper we have argued that a distributed, choreography-based paradigm has a number of practical benefits when applied to design and implementation of complex systems, such as distributed clinical workflows. The choreography paradigm provides a clean approach to issues of portability of workflow specification (allowing a high-level guideline to be implemented at different sites which are likely to have different local procedures for carrying out tasks within the guideline), re-use of process specifications (allowing site-specific implementation detail to be re-used within different high-level process specifications), and abstraction away from the specific entities providing services, which might change both between sites and between runs of the process.

OpenKnowledge provides an operational framework for quickly setting up such systems. In OpenKnowledge systems are composed around interaction models that coordinate the peers’ behaviours by specifying the roles they can take, the exchange of messages between the roles and the constraints of the messages. Peers participate in the interaction taking one (or more) roles: in order to participate they need to compare the constraint in the roles with their available services and subscribe to the interaction on a distributed discovery service, that initiates interactions when their roles are filled.

In the current version of OpenKnowledge, constraints in the choreography are semantically matched with the capabilities of the peers. However, while feasible using the annotations, there is still no support for specifying requirements on how the constraints should be solved, or on what requirements the participants should have (for example, the peer taking the doctor role should be certified by the competent institution). Improving the specifications can help the peers both in selecting the proper interactions for their goals and in matching their capabilities with those required by the choreography.

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**Paolo Besana** is a post doctoral researcher at the University of Edinburgh. He is currently working on Safe and Sound project funded by EPSRC and in collaboration with Cancer Research UK, aiming at creating a distributed approach to medical decision support systems. Previously he worked in the EU-funded OpenKnowledge project, whose target was the dynamic integration of the components and services in open, peer-to-peer systems. He holds a PhD in Informatics from the University of Edinburgh and a degree equivalent to an BSc+MSc in Telecommunication Engineering from the Politecnico of Milano. Before the PhD, he worked as software engineer for 5 years.

**Vivek Patkar** is a clinical research fellow at department of academic oncology at Royal Free Hospital. His research interests include evidence-based medical decision making and clinical decision support systems. He received his master's degree in surgery from Bombay University.

**Adam Barker** is a postdoctoral Research Assistant in the Department of Engineering Science, University of Oxford. Prior to joining Oxford, he was employed as at the National e-Science Centre (NeSC), in Edinburgh. Adam holds a PhD in Informatics from the University of Edinburgh, MSc in Distributed Systems and BSc in Software Engineering from Newcastle University. Complementing his academic experience, Adam has completed internships at Hewlett-Packard and BAe Systems. For further information and list of publications please refer to http://www.adambarker.org

**David Glasspool** is a senior research fellow in the School of Informatics at the University of Edinburgh. His research interests include executive control, routine behaviour, and planning in the face of uncertainty, in both computational systems and human cognitive psychology. He has applied this theoretical work in the development of clinical computing systems. He received his PhD in computational modelling from the psychology department of University College London.

**David Robertson** is the Director of the Centre for Intelligent Systems and their Applications, part of the School of Informatics at the University of Edinburgh. His current research is on formal methods for coordination and knowledge sharing in distributed, open systems - the long term goal being to develop theories, languages and tools that out-perform conventional software engineering approaches in these arenas. He was coordinator of the OpenKnowledge project (www.openk.org) and was a principal investigator on the Advanced Knowledge Technologies research consortium (www.aktors.org), which are major EU and UK projects in this area. His earlier work was primarily on program synthesis and on the high level specification of programs, where he built some of the earliest systems for automating the construction of large programs from domain-specific requirements. He has contributed to the methodology of the field by developing the use of "lightweight" formal methods - traditional formal methods made much simpler to use in an engineering context by tailoring them to a specific type of task. As an undergraduate he trained as a biologist and continues to prefer biology-related applications of his research, although methods from his group have been applied to other areas such as astronomy, simulation of consumer behaviour and emergency response.