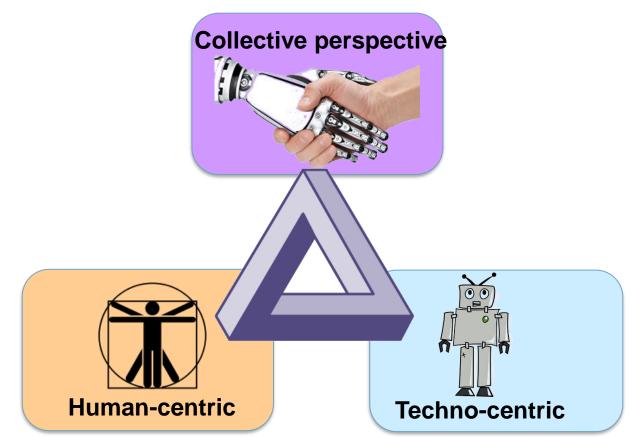


Part I

Three perspectives on Al



3 PERSPECTIVES ON ARTIFICIAL INTELLIGENCE





TECHNO-CENTRISM

SUFFICIENTLY DEVELOPED AI CAN BE APPLIED TO SOLVE ANY PROBLEM.

II AI MIGHT INTRODUCE ADDITIONAL PROBLEMS, WHICH CAN IN TURN BE SOLVED BY AI.

III THE MORE AI IS DEVELOPED, THE LESS USER INTERACTION IS NEEDED.

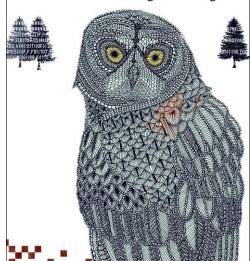
IV AI HAS VASTLY MORE L POTENTIAL THAN HUMAN INTELLIGENCE



NICK BOSTROM

SUPERINTELLIGENCE

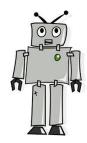
Paths, Dangers, Strategies





ALPHA (GO) ZERO



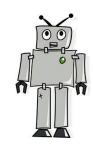




THE ROLE OF THE HUMAN







- > **Problem:** How to maintain human control over Artificial Super Intelligence.
- > Solution: Program human values into Al system



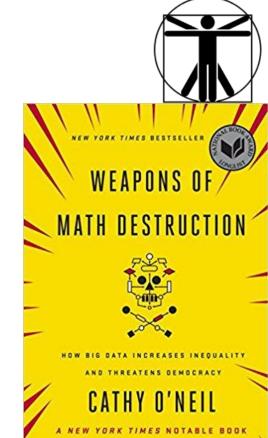
HUMAN CENTRISM

I
AI IS NO MATCH
FOR HUMAN
INTELLIGENCE.

YOU CANNOT
SOLVE PROBLEMS
CAUSED BY AI BY
APPLYING MORE
AI

III
AI WILL NEVER
ACT WITHOUT
HUMAN
INVOLVEMENT

IV
AI WILL REMAIN
RELATIVELY
LIMITED FOR THE
FORESEEABLE
FUTURE.







PROPERTIES OF A WMD

- > Non-transparant: It is unclear how AI arrives at its conclusions.
- > Scale: The decisions made by AI affect large groups of people.
- Damage: The AI brings damage to large groups of people.



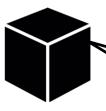
We must turn around underperforming schools in Washington D.C.

MATHEMATICA

We develop an objective and accurate model IMPACT to assess a teacher's performance

Policy Research

Along with 205 other teachers with a low IMPACT score, I got fired. Why?



It's a complex algorithm you won't understand. Furthermore, it's corporate secret.

Many of my students came from a different school where they tampered test _____ scores. They started scoring less in my tests...



Sarah Wysocki





OTHER EXAMPLES OF WMD'S



Predictive policing

BUSINESS NEWS OCTOBER 10, 2018 / 5:12 AM / 6 MONTHS AGO



Amazon scraps secret Al recruiting tool that showed bias against women

Scan CV's

Political campaigning

Assess creditworthiness

Predict chance of recidivism

Calculate insurance premium



THE ROLE OF THE HUMAN



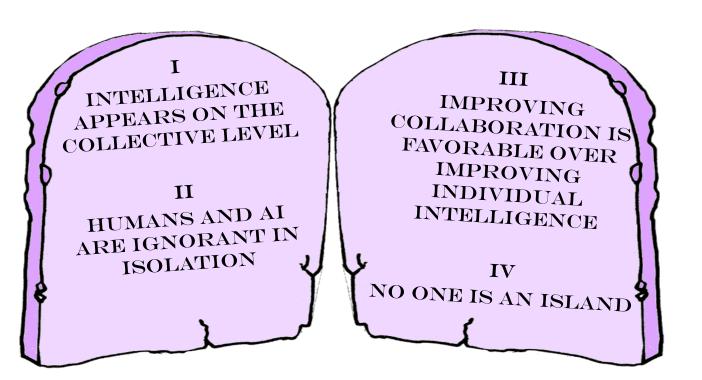


> Solution: Apply AI sparingly.





COLLECTIVE INTELLIGENCE





"A richly detailed guidebook leaders need to capture the opportunities of AI and the fourth industrial revolution."

-ILMS SOWNE
Fourth and Section Chairma World Concept Fourth

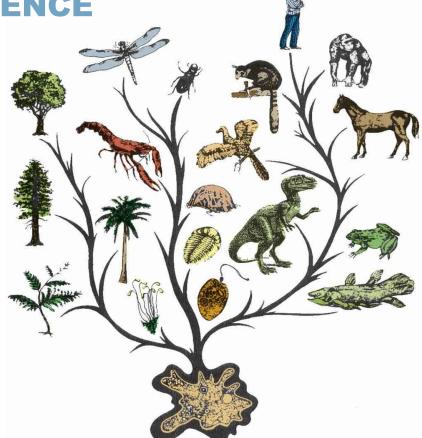
HUMAN + Reimagining Work in the Age of AI MACHINE

PAUL R. DAUGHERTY
H. JAMES WILSON

HAPVARD BUSINESS REVIEW PRESS



SOCIAL INTELLIGENCE







SOCIAL AI IS ESSENTIAL













THE ROLE OF THE HUMAN





Problem: Integrating AI into teams, organisations, and society inevitably disturbs the equilibrium between autonomy and control.

Solution: Detect and redirect undesirable developments.



EXAMPLE RADIOLOGY

- Deep learning networks achieve super human performance, on segmentation and identification of malicious tissue in X-ray images.
- Al is expected to revolutionize radiology
 - Al-replacement: Some tasks are completely taken over by Al. E.g. visual interpretation of radiology images. Results in deskilling of radiologists.
 - Al-augmentation: For some tasks, the Al system augments the human. E.g. planning a patient treatment. Requires the human to maintain expertise and acquire additional expertise on how to use the Al support system.
 - Al-maintenance: These are tasks that are added to the radiology workflow that did not exist before. Requires a whole new set of skills. Examples are: (re-)training the Al system.

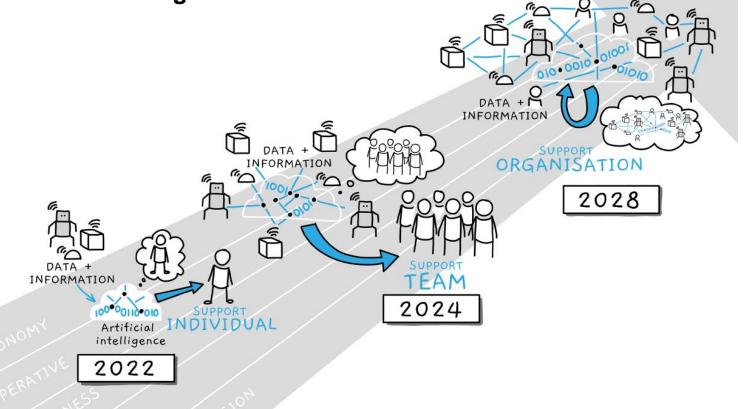


Part II Team Design Patterns

From smart personal assistant to Al-worker in an organisation



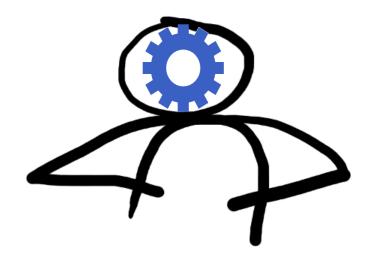
THO innovation for life

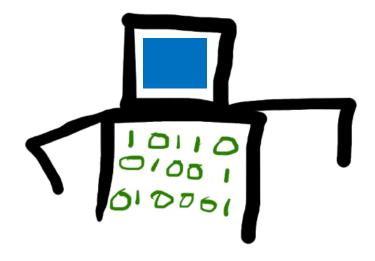


SELF AR PREDICTION



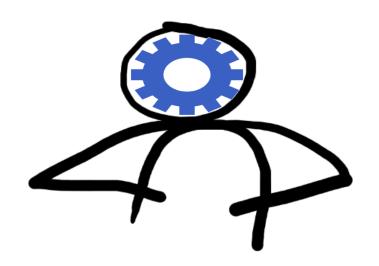
COMPUTER AS A TOOL

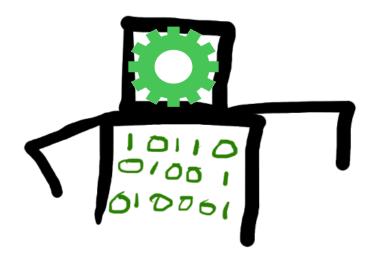






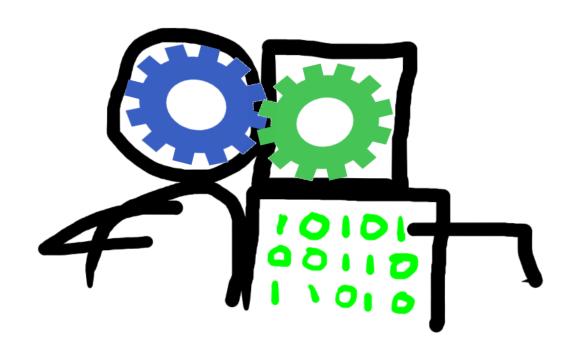
COMPUTER AS AN ISOLATED AGENT





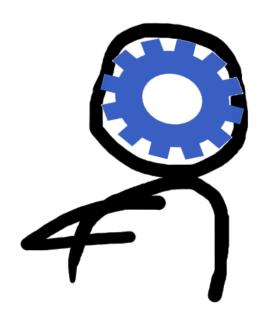


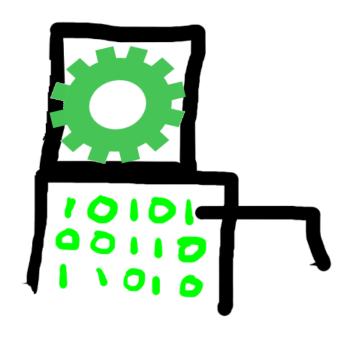
COMPUTER AS A TEAMMATE





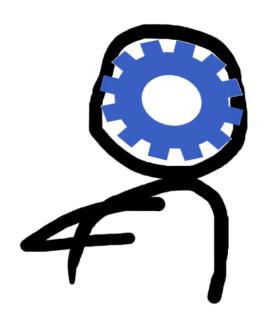
DYNAMIC TEAMING

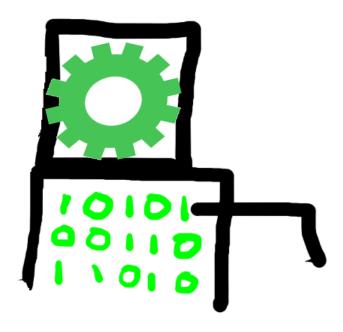






DYNAMIC TEAMING

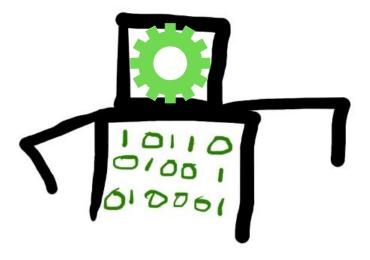


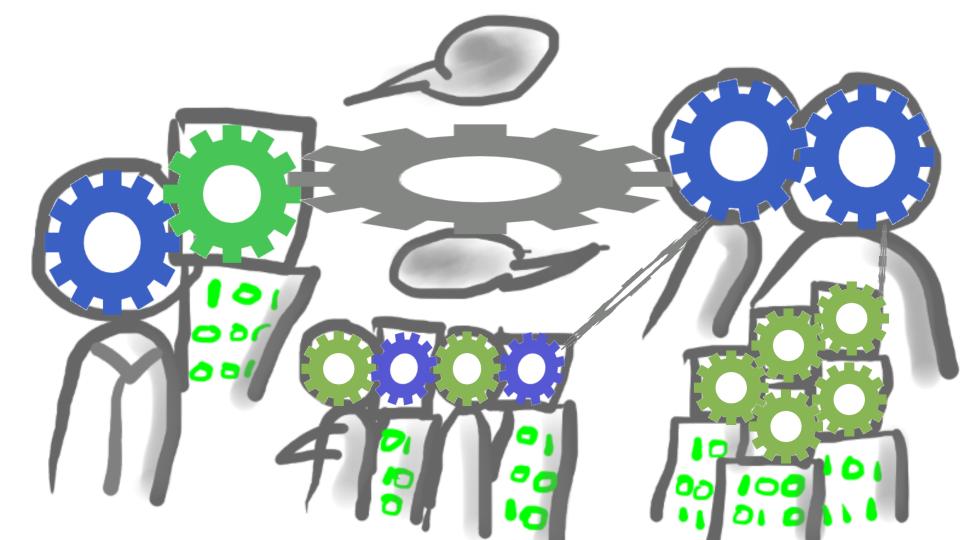


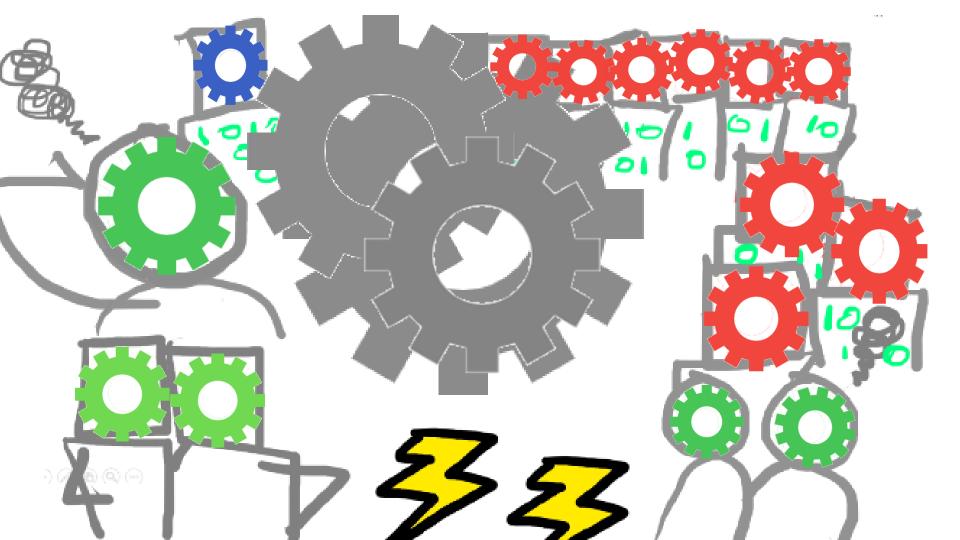


RUNAWAY AI











TEAM DESIGN PATTERNS

- How to design coherent human agent teams in a way that is
 - Simple and intuitive to allow communication among stakeholders
 - **General** enough to represent a broad range of teamwork
 - Descriptive enough to allow comparison of different solutions and situations
 - **Structured** enough to have a pathway from the simple intuitive description to the more formal specification.
-) Fosus on:
 - Nesting
 - Time

Team Design Patterns

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ABSTRACT

This paper proposes an intuitive graphical language for describing the design choices that influence how intelligent systems (e.g. artificial intelligence, robotics, etc.) collaborate with humans. We build on the notion of design patterns and characterize important dimensions within human-agent teamwork. These dimensions are represented using a simple, intuitive graphical iconic language. The simplicity of the language allows easier expression, sharing and comparison of human-asent teaming concepts. Having such a language has the potential to improve the collaborative interaction among a variety of stakeholders such as end users, project managers, policy makers and programmers that may not be human-agent teamwork experts themselves. We also introduce an ontology and specification formalization that will allow translation of the simple iconic language into more precise definitions. By expressing the essential elements of teaming patterns in precisely defined abstract team design patterns, we work towards a library of reusable, proven solutions for humanagent teamwork.

CCS CONCEPTS

- Computing methodologies → Artificial Intelligence; Intelligent Agents - Human-Centered Computing → Interaction Design; Interaction design theory, concepts and paradigms

KEYWORDS

human-agent teaming; design patterns; joint activity; joint cognitive systems; long term teaming;

1. Introduction

Teaming is something people do every day. Children learn it at an early age and can quickly and easily adapt their teaming skills to novel situations with different people. Given people's intuitive ability to team in varying circumstances, it would seem that coding such common sense in a machine would be straightforward, but codifying common sense has been an elusive goal in more areas than teamwork. Currently, most machines lack even the most basic teaming skills [12].

Given the difficulty of codification, one alternative is the use of teaming theory and guidelines such as [14]. These principles identify important considerations for designers. However, they are often abstract, requiring significant interpretation to translate into a specific domain and are challenging to instantiate without human-machine teaming expertise. The use of good examples of teaming behavior is another approach (e.g. [13]), but reuse of examples depends on application details making specific examples into to seneralize.

We propose borrowing the concept of design patterns to assist in the understanding and designing of human-machine systems. Design patterns are reusable solutions to recurring problems. The patterns try to capture the common invariant properties of the problem and the essential relationships needed to solve the problem. Design patterns are not solutions to particular problems, they are not rules to be followed, nor are they templates to be instantiated. They are abstract solutions that allow a designer to reuse ideas that worked in the past for commonly faced problems. These patterns can be extended to meet varying teaming needs across a variety of teaming contexts.

Team pattern design solutions should be (1) simple enough to provide an innuitive way to facilitate discussions about human-machine teamwork solutions among a wide range of stakeholders including non-experts, (2) general enough to represent a trong range of teamwork capabilities, (3) descriptive enough to provide clarity and discernment between different solutions and situations, and (4) structured enough to have a pathway from the simple intuitive description to the more formal specification. This paper processes an approach that meets all of these requirements.

Additionally, our approach captures two critical aspect of teaming that are missing in current approaches and often overtooked in design: nesting and time. Nesting refers to the recursive and compositional nature of activity. When a human collaborates with a machine, the work is embedded in larger organizational and procedural structures [20] and can often be decomposed into simpler structures. Counceting these levels of design from individual AI systems to whole human-AI societies can be regarded as one of the great research challenges for the coming decades [17]. Additionally, joint activity is a process, extended in space and time [3]. One of the main advantages of teams is their flexibility to adapt, which means they will change patterns over time. Our team design pattern language provides a means to capture both nesting and time.

The paper is organized as follows. First, we discuss the background of design patterns, and its relation to team patterns. In Section 3, we discuss the basic building blocks of team design

¹ Both authors contributed equally to this paper



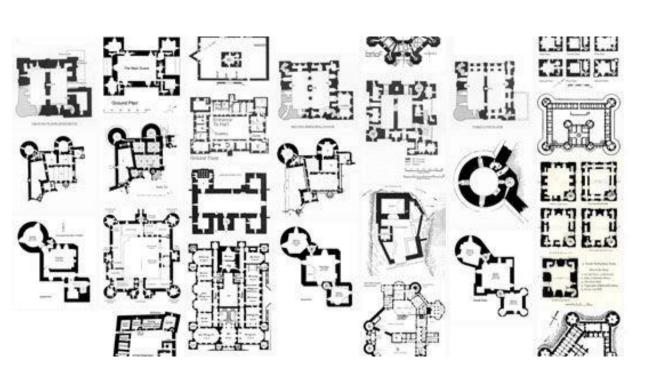
CHRISTOPHER ALEXANDER







A PATTERN LANGUAGE



A Pattern Language Towns · Buildings · Construction Christopher Alexander

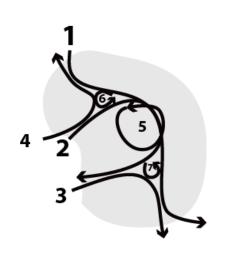
Sara Ishikawa - Murray Silverstein

Max Jacobson · Ingrid Fiksdahl-King Shlomo Angel



Christopher Alexander's Default Design Approach







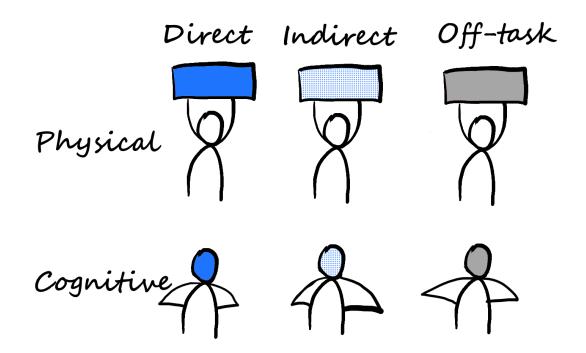
a. Start with a whole...

b. Differentiate it...

c. Into parts...

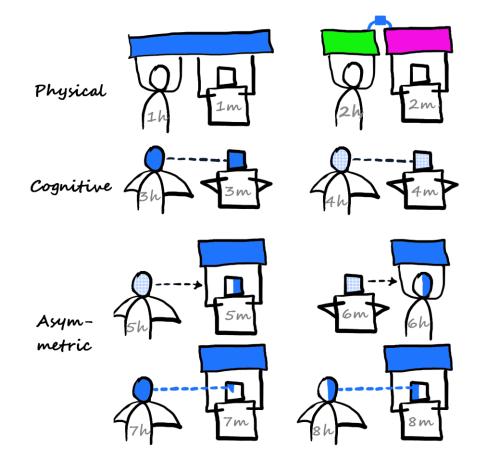


BASIC TYPES OF WORK



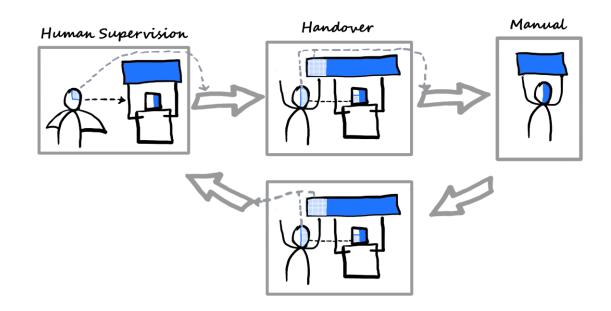


JOINT WORK



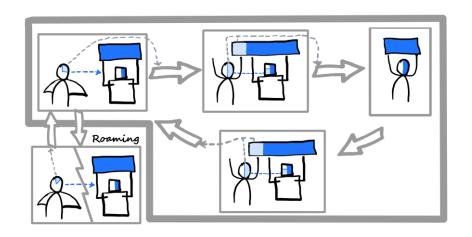


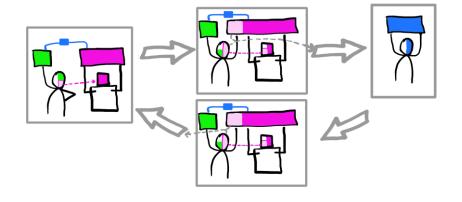
SUPERVISORY CONTROL





VARIANTS OF SUPERVISORY CONTROL

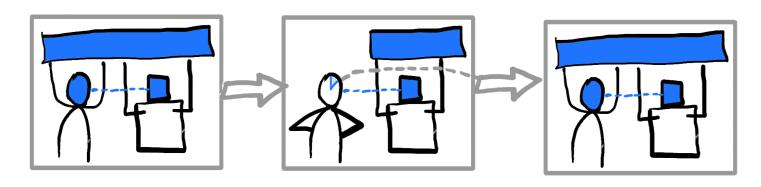






HIGHLY AUTONOMOUS PATTERNS







FORMAL SPECIFICATION

Teleoperation: Team Design Pattern

Name: "Tele operation"

Image: Img1
Use when: "machine has lim

Use when: "machine has limited autonomous capability, and human skilled operators are available..."

Positive effect: "Clear single point of control at human operator"

Negative effect: "Imposes heavy taskload on the human"
Example: "Teleoperation of a UAV..."

Involves actors : [7h,7m]

7h : Human

Name: "7h"

Performs < level of engagement = 1.0 > Teleoperating

Teleoperating: Direct Cognitive Work

Name: "teleoperating" Relevance = "on-task

Monitors < modality = auditive> [7h,7m]

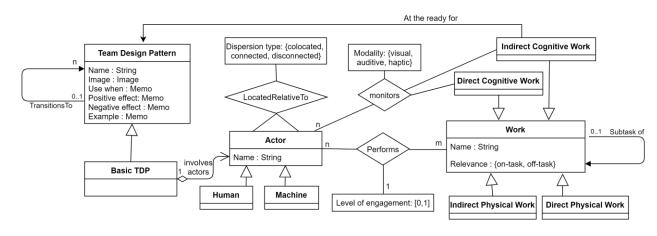
7m: Machine

Name: "7m"

Performs <level of engagement = 0.1> Teleoperating Performs <level of engagement = 1.0> PerformInstructions

PerformInstructions: Direct physical Work

Name: "PerformInstructions" Relevance = "on-task



GOAL: develop a pattern library for meaningful human control.



SUMANTA BARUAH.