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Using Hybrid System Dynamics and **Discrete Event** Simulations to Identify High Leverage **Targets for Process** Improvement in a Skill-based Organizational Structure

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Agenda

- Introduction & Motivation
- Characterizing the Work
- Managing the Data
- Building and Iterating the Models
- Preliminary Conclusions and Future Research





Motivation for this Research

- Team is responsible for the enterprise technology stack supporting ~200 hospitals of various types and dozens of ambulatory clinics and physician practices in more than 30 states across the US. The infrastructure is *big*.
- The team consists of ~150 people organized into smaller, skill-based times. The **teams are small**.
- The "DevOps in the public cloud" model is *the* trend for infrastructure management if you write your own applications...
 - We do not many key clinical systems, including core Electronic Medical Records (EMRs) currently deployed, are unsupported in the cloud.
- The key question: can an IT organization like mine **leverage DevOps** processes, techniques and technologies to meet our mandate with **on-premises infrastructure**?





The Healthcare Provider IT Organization

- Following from the nature of the systems involved, many if not most provider IT organizations are built to evaluate, purchase, deploy, integrate, support and upgrade / replace systems (primarily on premises) not to build them.
 Most development is focused on integration and analytics.
- Hospitals and health systems provide services from **extensive facilities** that require IT infrastructure throughout, with network and computing end-user equipment widely deployed.
- Teams are organized around departmental silos: technologies (such as networks or endpoint computing), applications (especially certain types of COTS applications), and functions (e.g., project management, integration, or data & analytics). Cross-functional teams are few and usually ephemeral, focused on specific projects.
- Cost pressures keep IT teams relatively small and outsourcing is common, so teams must be **responsible** for **both project** and **operational tasks** within their departmental silos.
- Common activities that require multiple skills must be coordinated across departmental silos.

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Fundamentally

Can we make this...







Characterizing the Work

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Behavior observed in case study team

Basic Organizational Structure

- Teams are built around specific technical **skill**: network engineering, email, virtualization, server administration, etc.
- Teams support **both** operational **tickets** and project **tasks** within their specialty.
 - Managing **multiple projects** within the same team is much more complex than single projects.
 - Managing **both operational and project efforts** of **varying priorities** together within a single time is even more complex.
 - Work of any type (tickets or tasks) are effectively **managed as queued, by priority**.
- Teams have engineers of varying **skill levels**.
 - There are generally fewer resources with high skill levels ("senior engineers").
 - There are usually more with medium and lower levels of skill ("engineers" and "associate engineers").
 - This varies by team actual team structures are used in the models.
- The models are intended to reflect the flow of work over relatively **short timelines**, so there is **no ability to flex staff** through hiring (or contract outsourcing).
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- C. Kang and Y. S. Hong, "Evaluation of Acceleration Effect of Dynamic Sequencing of Design Process in a Multiproject Environment," Journal of Mechanical Design, vol. 131, no. 2, Jan. 2009, doi: 10.1115/1.3066599.
- S. Fricke and A. Shenbar, "Managing multiple engineering projects in a manufacturing support environment," IEEE Transactions on Engineering Management, vol. 47, no. 2, pp. 258–268, May 2000, conference Name: IEEE Transactions on Engineering Management. doi:10.1109/17.846792.



Basic Work Characteristics

- Work items (tickets and tasks) can **require** one or more resources with a certain **level** and **type** of technical **skill**.
- Work items generally represent <5 hours of work *effort* (distinct from total *duration*)
 - Staff may not have sufficient service time available to complete a ticket or task, resulting in **re-queuing** and **switching costs**.
 - Available service time can change based on **re-prioritization** (escalating / expediting) of other work.
- Work items have a **priority** (low, medium or high) which **can change** over time.
 - Project tasks become more time-sensitive as they get close to or past their due date, especially for those on the critical path ("timeliness").
 - Operational tickets are more likely to be escalated the longer they're open ("**responsiveness**").
- Quality or the lack of it is reflected through the creation of **rework** (in the case of tasks) and **incidents** (in the case of tickets).
 - Probability for errors is initialized based on existing data but can change over time due to dynamic influences.
 - Actual quality is influenced by a difference between the required skill level and the ability of the assigned resource.

- J. M. Lyneis and D. N. Ford, "System dynamics applied to project management: a survey, assessment, and directions for future research," System Dynamics Review, vol. 23, no. 2-3, pp. 157–189, 2007, _eprint: https://onlinelibrary.wiley.com/doi/pdf/10.1002/sdr.377. doi: 10.1002/sdr.377.
- J. Wiik and K.-P. Kossakowski, "Dynamics of Incident Response," Apr. 2017, url: https://www.first.org/resources/papers/2005.



Why Hybrid Modeling?



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Identifying the improvement opportunities

- Find where opportunities exist and degree of benefit:
 - Fit available data to determine the distribution of work arrival times using Palisade @Risk.
 - Use Discrete Event Simulation (DES) predict the performance of work queuing in terms of total time, with prioritization by ticket or task using Matlab Simevents (module of Simulink).
 - Use System Dynamics (SD) Simulation account for influences that affect total work time, especially the impact of managerial pressure on rework and work stoppages driven by re-prioritization using Vensim.
- Couple models into a hybrid simulation to obtain benefits of both modeling methods by iterating through each model sequentially.
 - Data sharing currently manual, could be automated.
- Why these tools? In short: **cost**, **accessibility** and **familiarity**.
 - Note: one commercial tool can handle all these functions and would ease integration (AnyLogic).
- R. Jonkers and K. E. Shahroudi, "Connecting Systems Science, Thinking and Engineering to Systems Practice for Managing Complexity," Journal of the International Society for the Systems Sciences, vol. 65, no. 1, 2021, number: 1. url: https://journals.isss.org/index.php/jisss/article/view/3875
- B. K. Choi, D. Kang, and B. K. Choi, Modeling and Simulation of Discrete Event Systems. Somerset, US: John Wiley & Sons, Incorporated, 2013, url: <u>http://ebookcentral.proquest.com/lib/csu/detail.action?docID=1402564</u>
- K. Chahal, T. Eldabi, and T. Young, "A conceptual framework for hybrid system dynamics and discrete event simulation for healthcare," Journal of Enterprise Information Management, vol. 26, no. 1/2, pp. 50–74, Jan. 2013, publisher: Emerald Group Publishing Limited. doi: 10.1108/17410391311289541



Managing the Data

Tool-based data extract and estimates

Available Work Performance Data

- Operational requests and incidents are managed in **ServiceNow**.
 - Data gathered by team, ticket type and priority for model for six months, from 1/1/23 to 6/30/23 including both arrival and closure dates.
 - Arrival data fit as unique Poisson distributions by team using Palisade @ Risk for use in the DES models as _ inter-arrival data.
 - Closure data will be used in verification and validation of the DES results.
 - **Priority** breakdowns are based on actual data by team. _
 - In the case of the team analyzed: High 0.5%, Medium 40%, Low 59.5%. •



Fitting Existing Operational Data

- **Poisson distribution** selected for fit of arrival times as this directly translates into **exponential random inter-arrival times** in DES, defined through Palisade @Risk (note: could be done in Matlab if needed repetitively).
- Example: Incident arrival fitted Poisson distribution mean of 6.6364 yields lambda of 0.150685, resulting in an entity generation equation of dt = -0.150685*log(1-rand()) in SimEvents.





Estimated Work Performance Data

- Data determined through interviews with department managers due to the lack of reportable data in existing systems:
 - Scheduled tasks also use exponential inter-arrival times but occur less frequently (and often in bunches in reality).
 - Lambda of .5 for project tasks and 1 for maintenance tasks.
 - **Error rates** are initialized on entity generation with a probability of 0.5%.
 - Service Times:
 - *Required* Service Time, est. as exponential (lambda of .15), calculated on entity generation.
 - *Actual* Service Time, est. as exponential (lambda of .3), calculated as entity is "picked up" for work by an engineer.
 - Required **skill level**, set on *entity* (work item) generation: High 5%, Medium 25%, Low 75%.
 - Actual skill level is set at each *server* (staff engineer) in the DES model.
 - Modification of error probability and required service times, based on skill level data (discussed on next slide).
- Parameters requiring further investigation / justification:
 - Effects of managerial **pressure** and **fatigue**.
 - Initial **error** and **stoppage** rates.
 - Tuned to fit observation initially.



Dynamic Parameters

- Work **stoppage** rates:
 - An entity with remaining "Required Service Time" is returned to the queue.
- **Error** generation rates:
 - Initial .5% chance of generating an error (either rework or new incident), set on entity generation.
 - Modified by *difference between required* skill level and *assigned* skill level by rule. Examples:
 - If the "skill difference" is 2 (a high skill level resource with a configured skill level of "3" working on a low skill task with required skill of "1") the rework probability drops to .3%.
 - With a "skill difference" of -2 (a low skill resource working a high skill task) the rework probability increases to 1%.
- Required service time also increases by rule based on **skill level differences**. Examples:
 - A high skill resource working on a low skill task reduces the required service time (set on entity creation) by 20%.
 - A low skill resource working a high skill task increase the required service time by 20%.



Building and Iterating the Models

Modeling the process and feeding in the data

Single-Team DES Model

- Five work **generators** with separate inter-arrival functions:
 - Incidents
 - Requests
 - Project tasks
 - Maintenance tasks
 - Incidents from errors / rework
- A single team queue with individual queues per resource ("server").
- **Restart** and **error generation** functions.
- NOTE: model does not yet enable managerial escalation of priority while an entity is queued. This may require creation of a custom queue object.





Generators

Parameterizing the DES Entity Generators



Routing Entities through Attributes



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Error Generation



1)

Probability of **Rework** initialized to .5% on **Generation** (later



Single-Team SD Model



- Modeled as two parallel flows of work:
 - Enable future differences between tasks and tickets in modeling of *timeliness* vs.
 responsiveness.
 - Enable differences in handling of rework vs. incidents resulting from operational changes.
- Influencing factors affecting both workflows shown in the center of the model:
 - Gaps in quality, productivity and timeliness.
 - Managerial pressure and quality focus as well as effects of fatigue.
- Note: internal model parameters not shown in this top-level diagram.



Initializing the SD Model from DES Data





Parameterizing SD Behavior





Iteration 2: Feed SD data back to DES Model

Two data elements fed from SD back to DES model during iteration 1. Rework Probability

- 2. Service Time

Captures increase in error generation due to management pressure

🚹 Block Parameters: IncGen

Entity Generator Generate entities using intergeneration times from dialog or upon arrival of events. Optionally, specify entity types as anonymous, structured, or bus.



🚹 Block Parameters: SrEng1

Entity Server

Serve multiple entities independently for a period of time and then attempt to output each entity through the output port. If the output port is blocked, the pending entity stays in this block until the port becomes unblocked. You can specify the service time, which is the duration of service, via a parameter, attribute, or signal.

When the block permits preemption, an entity in the server can depart early through a second port.

Main Event actions	Preemption Statistics
vent actions	Entry action:
intry* Service complete*	Called after entity has entered this block. To access attribute use: entity Prioriby
ixit Blocked Preempt	<pre>1 %entity.ServiceTime =3*log(1-rand()) %Base run 2 entity.ServiceTime =201*log(1-rand()) %Iteration 2 3 4 entity.ServiceStartTime = getStartTime()-entity.QueueTime; 5 5</pre>
Entity structure v entity Priority ReqSkillLevel ReqSkillType ReqServiceTime LeadTime Urgency RemWork ServiceTime AsignSkillLevel Rework WorkType StartTime QueueTime ServiceStartTime	<pre>6 entity.AsignSkillLevel = 1 7 8 skilvldiff = entity.AsignSkillLevel - entity.ReqSkillLevel 9 if skilvldiff == 2 10 entity.ServiceTime = entity.ServiceTime*0.80 11 entity.Rework = entity.Rework-0.002 12 elseif skilvldiff == 1 13 entity.ServiceTime = entity.ServiceTime*0.90 14 entity.Rework = entity.Rework-0.001 15 elseif skilvldiff == -1 16 entity.ServiceTime = entity.ServiceTime*1.1 17 entity.Rework = entity.Rework+0.0025 18 elseif skilvldiff == -2 19 entity.ServiceTime = entity.ServiceTime*1.2 20 entity.Rework = entity.Rework+0.005 21 end</pre>
id	
priority	Insert pattern
2	OK Cancel Help Apply

Captures decrease in available time to complete a work item due to management pre-emption



Preliminary Conclusions and Future Research

Expanding the simulations

Results after two iterations

- Key parameters shared between models:
 - Inter-arrival rates.
 - Work pickup and stoppage rates.
 - Error generation rates.
- After a full iteration of the DES and SD model, injecting SD results back into DES results in:
 - Significant increase in incidents from errors overall, as queuing delays led to increased managerial pressure.
 - Minimal to no impact to completion times in **high priority** work.
 - Significant increase in completion times of medium and low priority work.

NOTE: Management would intervene differently well before extremes seen at the end of the iteration





Observations from Simulation

- The team as modeled is **not in control** in either baseline or second iteration:
 - Medium and high priority work items are handled reasonably quickly initially, but low priority items experience significant and growing delays.
 - With the impact of managerial pressure and fatigue added to a second iteration, performance on medium priority work degrades and low priority work delays grow increasingly long.
- 1/3 more work items were stopped and re-queued due to management pressure, and there was a very large increase in Incidents from Rework (errors).
- This resulted in a 25% increase in work completed in response to incidents (because there were so many more of them due to errors) as well as an **increase in all completion times (**18% for P1s and a 125% increase in P2s).
- For all intents and purposes, by the end of the simulation many **P3s simply remained queued** with 75% less completed during the run.
- These characteristics are **directionally consistent with observations** from the case study team in the real world.



Preliminary Conclusions

- Potential improvement areas that can be influenced by automation:
 - Reducing completion times by **reducing queue times**, either from stop and restart or from transfers.
 - Improving operational responsiveness through triggering automation on arrival (e.g., incident or request arrival).
 - Improving actual work quality / reducing the rate of rework and the generation of new incidents by reducing or eliminating human error and variability.
 - Increasing pickup and completion rates.
 - NOTE: focus on relatively *high-volume* and *repeatable* work for automation to be worth the effort.
- Additional options for improvement beyond automation are suggested by the models and many have been studied previously – but are not the focus of this research:
 - Close coordination between project and functional leaders on work priorities.
 - Limit the number of projects in process through governance and prioritization.
 - Separate operational and project responsibilities between different staff.
 - Hire and (and train) multi-skilled resources.



Limitations of the Models

- Each model iteration currently run for very long periods to indicate accurate **direction**, not **magnitude**. •
- To closer approximate reality, feedback between the two models should be *much more rapid weekly if not daily*. •
- From a practical standpoint this requires *automation of a test harness* encompassing both simulation tools to • initialize and transfer data outputs from on model / iteration to inputs in the next model / iteration – or the use of a modeling tool that supports both simulation types.





Expanding Simulation to Two Teams

- Adding a second team to the SD and DES models introduces additional rules for managing work passing between them:
 - Each team supports mixed work types e.g., both unscheduled and scheduled that arrive at differing rates.
 - Each team supports multiple concurrent projects / products at different stages of planning and execution, as well as multiple concurrent incidents and requests.
 - Active work in progress can be returned to the queue for a new reason: it can be "reassigned" to another team's queue due to a lack of certain technical skill in the originating team.
 - Note that this can happen multiple times with a given work item.
 - Reassignments between teams occur at a certain rate (determined by the degree of coupling between the teams) and require coordination to ensure efficient completion and are an indicator of a higher level of overall work item complexity.
 - The models will only depict interactions between two skill-based teams interactions become much more complex as additional teams become part of the work process.
- **New dynamics are expected** that could introduce new criteria for identification and prioritization of automation targets.



Ongoing Data Analysis

- The comparison of baseline simulation results to actual team performance data for validation remains in process.
- Research to better justify the internal parameters in the SD model regarding the effects of management pressure and fatigue is needed.
 - To date the literature indicates the *direction* of these effects, but not the *magnitude*.
- Sensitivity analysis of overall completion times to restart and error rates, and upstream influences on those in the SD model, is needed to better inform the prioritization of managerial interventions to improve the process.



Summary

- Hybrid modeling including SD and DES predicts results that neither method would independently, particularly for a non-equilibrium system.
- Higher fidelity results require an automated test harness or a single tool that supports integration of both methods.
- Target process automation at high-volume, repeatable activities that lead to increased queue time, error rates and switching costs.



Link to the Models



https://www.engr.colostate.edu/~drherber/publication?key=Enos2024a



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Questions?

Thank you



