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AI and Enterprise Business Processes + NL-Driven Enterprise

Alexei Lapouchnian, PhD, Manager, Digital Accelerator, Natural Resources Canada

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AI and the Enterprise

- Based on research conducted at the [University of Toronto](#) in partnership with a large technology company
- AI and the Enterprise
 - How do we introduce AI into the IT landscape of an enterprise?
 - *Specifically*: How do we inject AI into [enterprise business processes](#) (BPs)?
 - *More specifically*: How do we [augment decisions](#) made as part of BPs with AI?



AI's Adoption in Enterprises: Some Issues

- Adoption and operation requires specialized skills (e.g., in ML), user training, etc.
- Where will AI create **value** in organizations and where should it not be deployed?
- How do we integrate AI-driven systems into **existing processes**?
- How do we ensure that **black-box** AI systems comply with various **regulations**? **Audit trails** may be needed!
- How do we guard against **unintended consequences** of using AI systems that may have **financial/reputational damage**?
- How do we make sure that AI systems are **trusted** and **accepted** within the organization, by both management and individual users?



Improving Adoption of AI in Enterprises

- Understand and manage **complexity**
- Build and maintain **trust** in AI
 - **Transparency** (when needed/desired)
 - Ability to provide **explanations/justifications** for AI's **recommendation** to users
 - Clear **evolution paths** for AI-driven systems and user engagement with them
- **Customized/personalized, context-aware** interactions



AI-Enhanced Business Processes

- Integration of AI into **business processes**
 - BPs – typical way organizations define work to be done
- BPs contain **decisions** (and activities)
 - E.g., whether to approve a loan or an insurance claim
- Currently – most decisions are done by humans
- With the help of AI-based systems
 - Introduce AI Advisors to **support human decision making**
 - Many options for such systems: from gathering/filtering relevant information for humans to supporting fully automated decisions



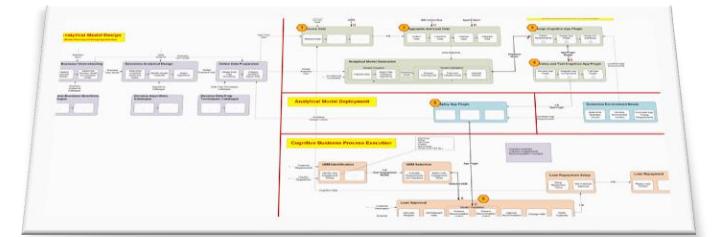
AI and Business Processes – Complexity

- Introduction of AI affects:
 - The process containing the decisions aided by AI
 - Other processes are introduced or are impacted as well, e.g.:
 - **Design** of the AI-based system, data collection, model training, etc.
 - **Monitoring of decision/recommendation outcomes** and their effects on *business goals*
 - **Context monitoring** for situation awareness
 - **User engagement** with AI-driven systems
 - Etc.
- AI and organizational BPs both **evolve**
 - BPs: due to changes in **requirements**, **business context**, etc.
 - Systems: increased sophistication, changes in recommendation quality

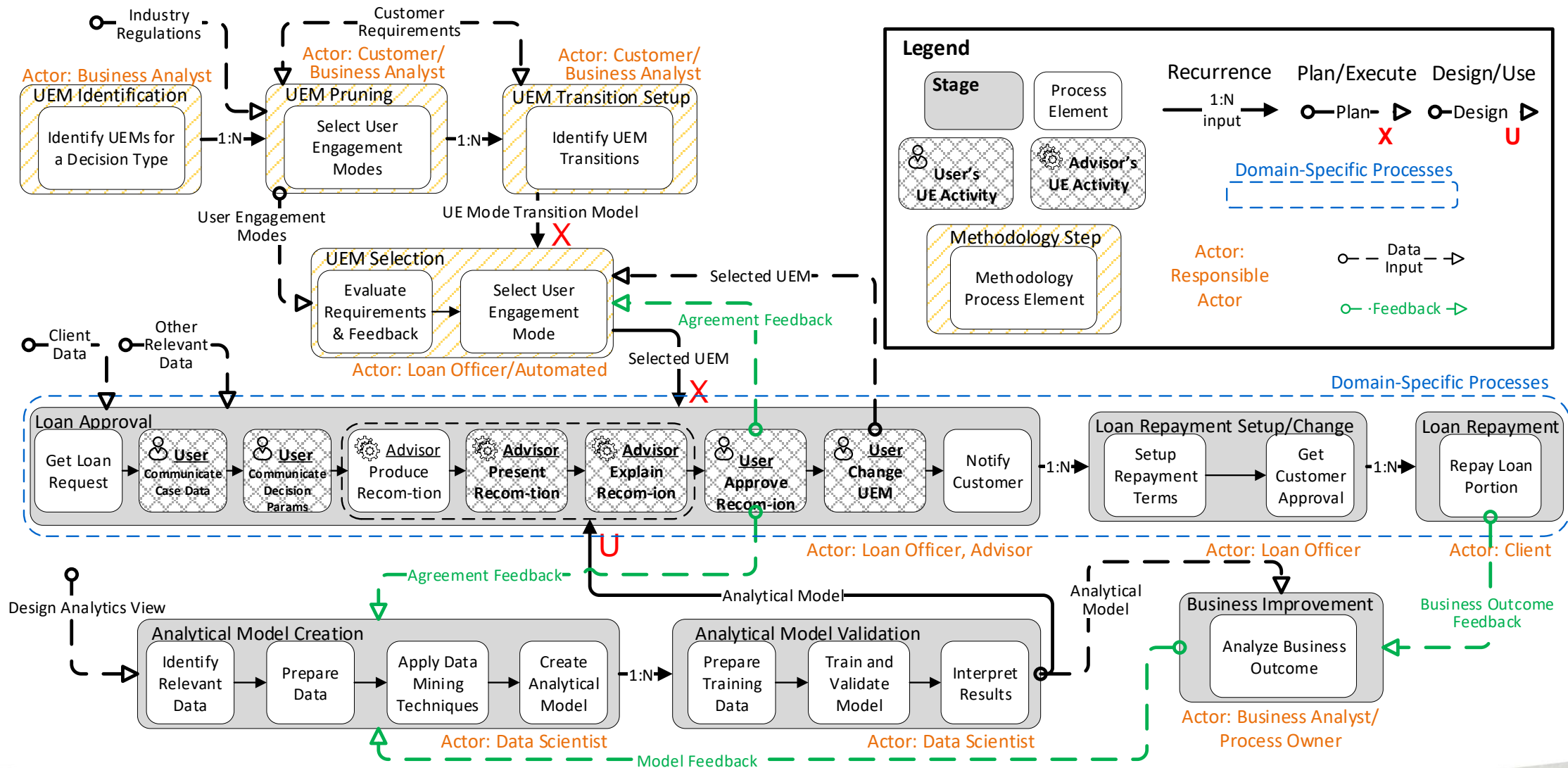


Business Process Architectures

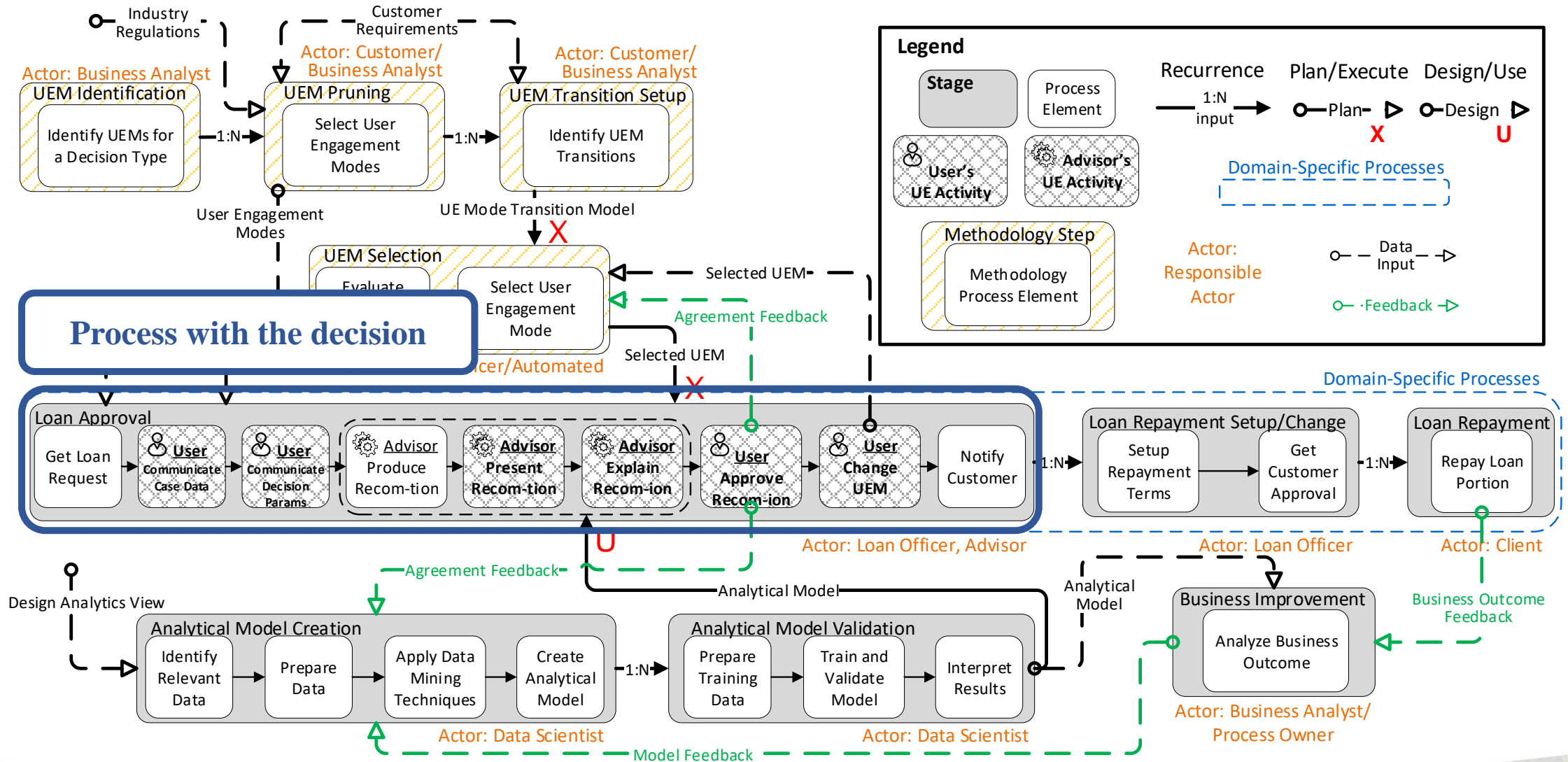
- Need to model and analyze multiple related BPs – i.e., need to focus on **BP Architectures** (BPAs)
- Business Process Architecture - model showing **BPs and their relationships**
- We use a previously proposed notation called *hiBPM*



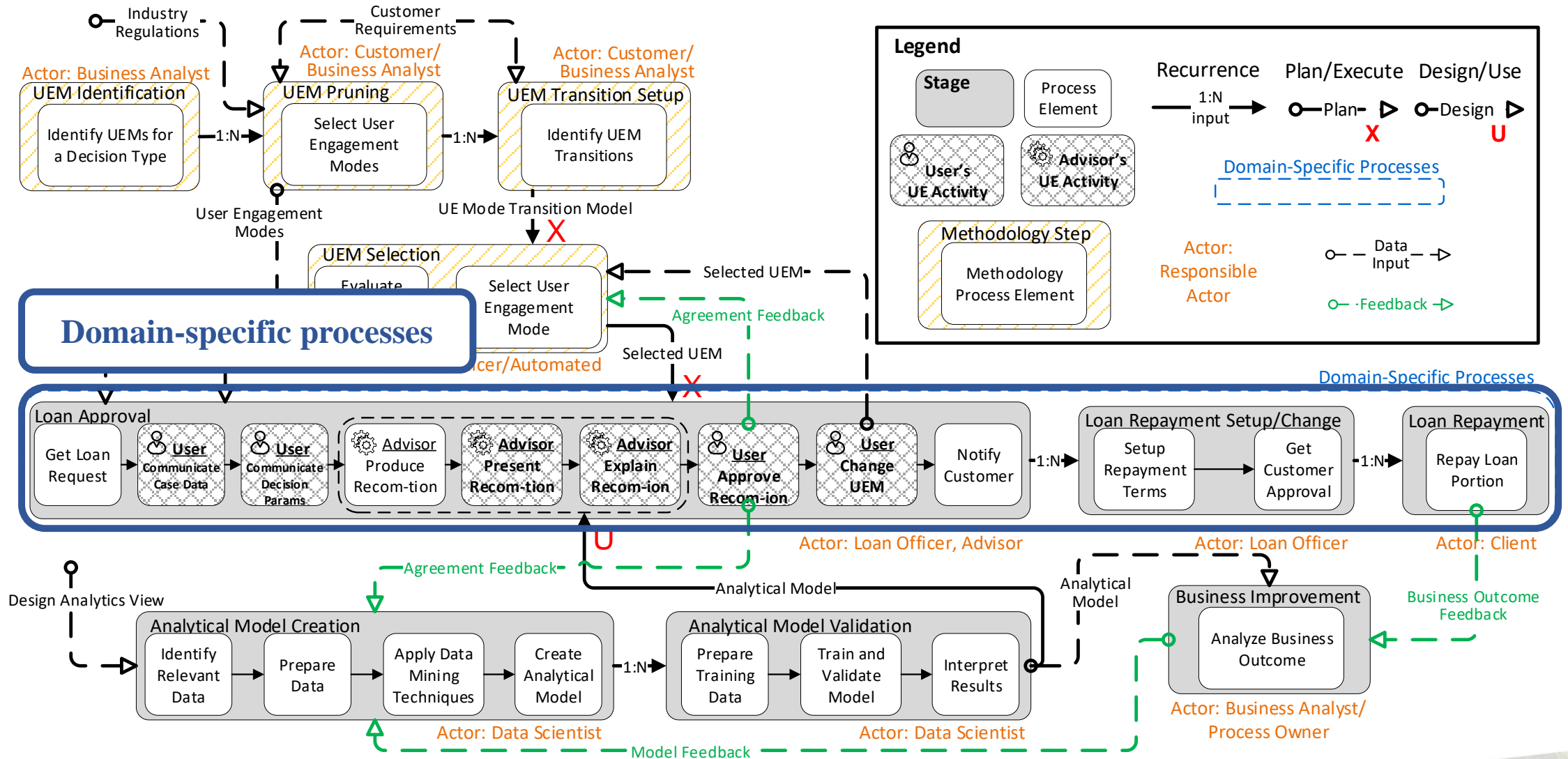
Business Process Architectures



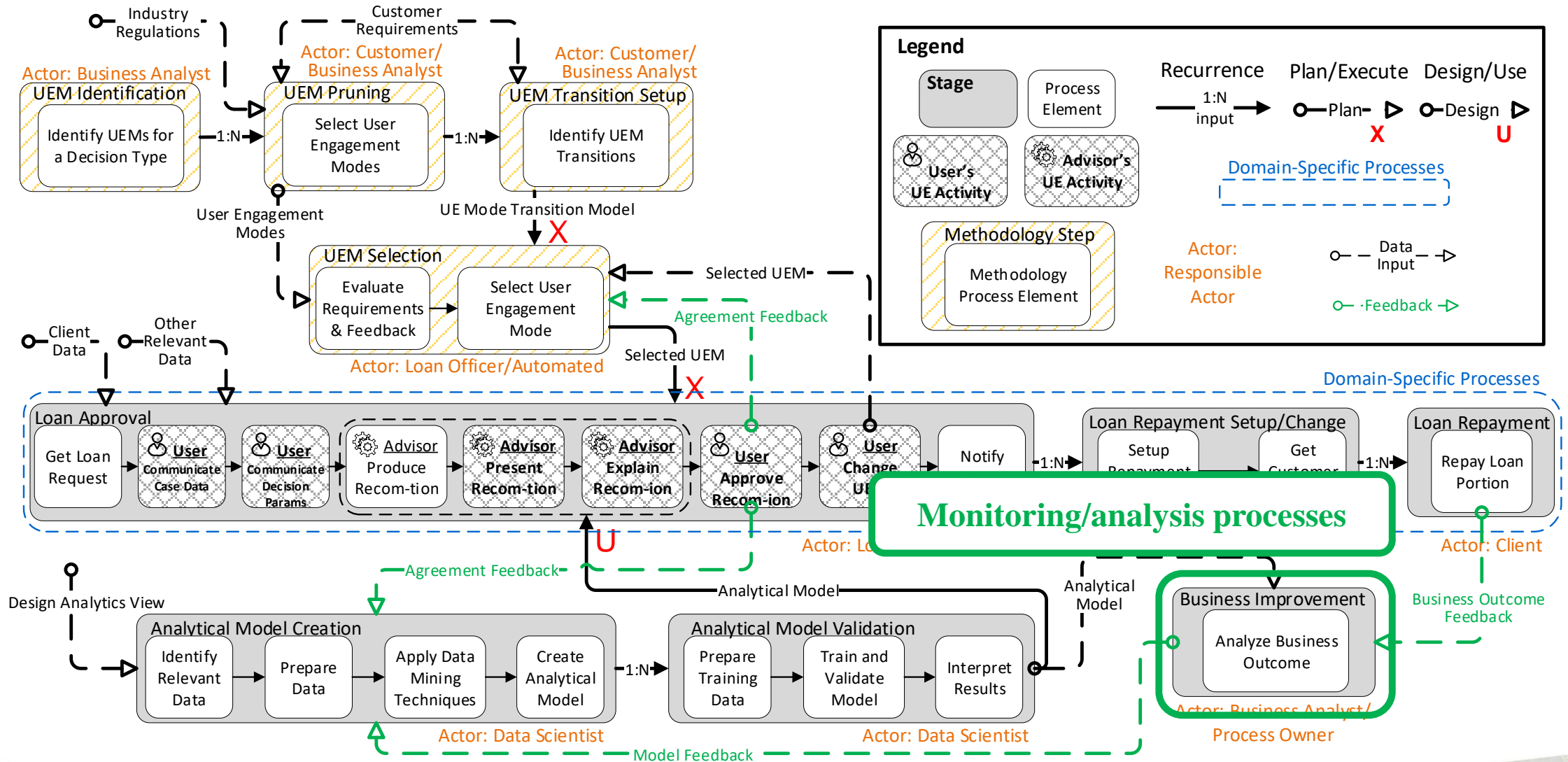
Business Process Architectures



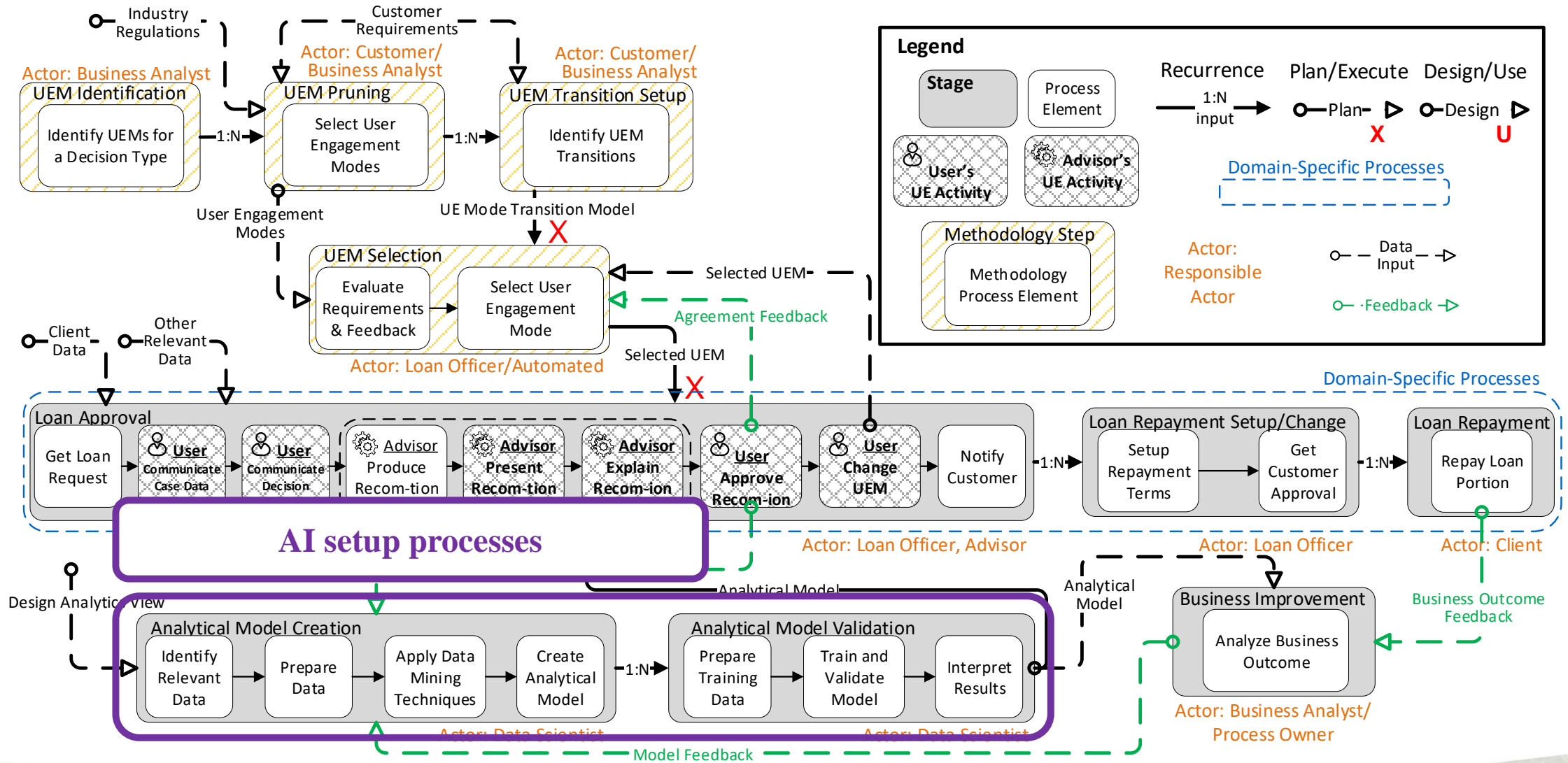
Business Process Architectures



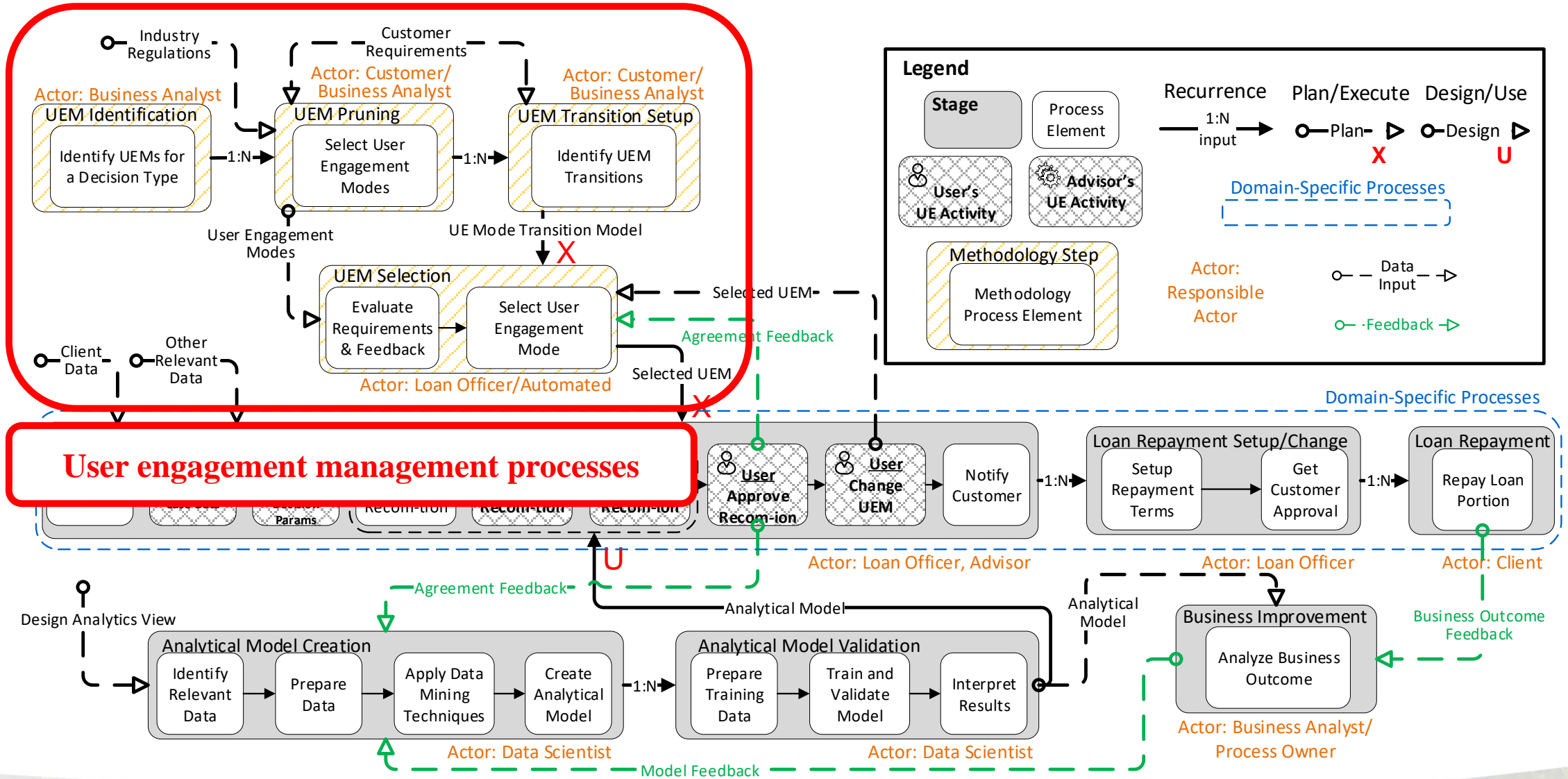
Business Process Architectures



Business Process Architectures



Business Process Architectures



User Engagement

- Our view of user engagement:
 - How users are interacting with AI and how AI is interacting with users **in the context of decision making**
- Objectives for designing user engagement with AI:
 - Minimize **disruptions** to users and to existing BPs
 - **Comply with** organizations' **internal requirements** and **constraints** as well as those from their respective business domains
 - Change based on different situations – **contexts**
 - Support desired level of **transparency** of systems' operations
 - Build and maintain the desired level of **trust** in AI-driven systems

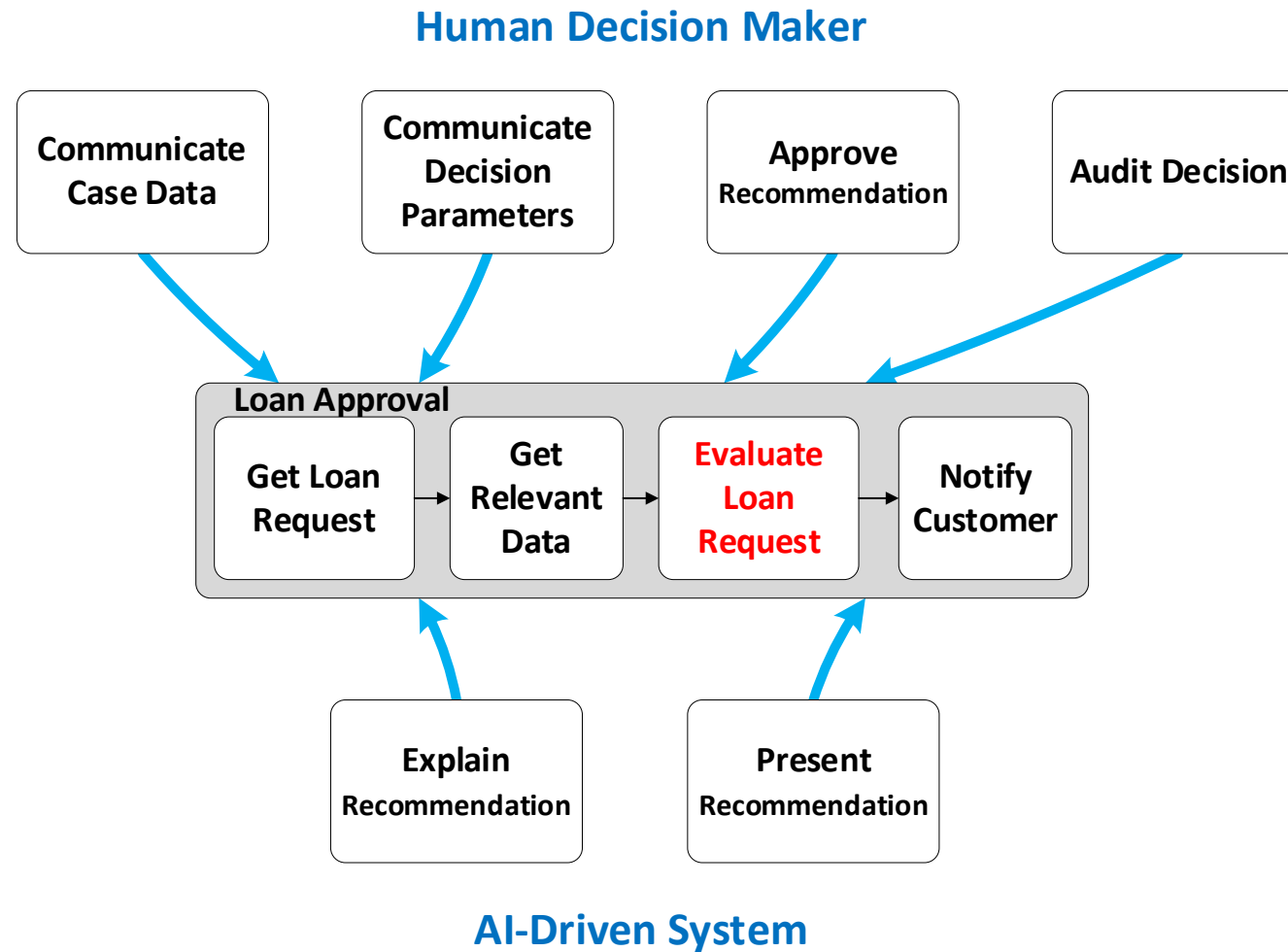


Building User Engagement

- What are the building blocks of UE?
- **User Engagement Actions (UEAs)**
 - Elementary activities/interactions that comprise UE with AI
 - Different combinations of UEAs define particular user engagement configurations
 - **NOTE:** The set of UEAs depends on the nature of the task – e.g., on the type of decision to be made
 - Can also be influenced by the business context, regulations, etc.



User Engagement Actions



User Engagement Actions (UEAs) [1]

- Sample UEAs for decision making
 - Actors: human decision maker (H) and AI assistant (AIA)
- **Communicate Case Data**
 - H communicates the details of the particular decision instance to AIA. Alternatively, AIA can obtain the relevant data itself.
- **Communicate Decision Parameters**
 - H communicates decision parameters to AIA, including the criteria for making it, the desired confidence levels, etc.
- **Present Recommendation**
 - AIA presents the recommended decision alternative to H. For decisions with more than two potential alternatives (those that are non-boolean), a ranked list of alternatives may be given.



User Engagement Actions (UEAs) [2]

- **Approve Recommendation**
 - H either **approves or rejects** the previously presented recommendation. Variants for decisions with more than two options may include the ability for the user to pick an alternative option.
- **Explain Recommendation**
 - **Explanation and justification** of the recommendation are presented by AIA to H.
- **Present Decision**
 - AIA **presents previously made decision** to H (or to a specially designated auditor). Variants include the presentation of batches of decisions for subsequent audits.
- **Audit Decision**
 - H (or a **designated auditor**) **audits the decision** previously made by AIA. Variants include the audit of batches of decisions.



User Engagement Options

- Many ways for users to engage with AI systems
 - Different ways to **communicate**
 - Different **granularity/precision** of communication – e.g., per decision instance, per set of instances, for all instances, per user role, etc.
 - Results in **different positions** of UEAs in the BP architecture
 - The **type of decision/problem** – e.g., approval vs. resource allocation



Representing and Analyzing User Engagement Options

- Need a **systematic** way of analyzing user engagement
- Need to support the analysis of:
 - Organizational requirements, industry and other regulations
 - Individual decision makers' objectives
- Need to be able to represent:
 - The **space of options** for UE
 - The **criteria** for selecting the best options and **triggers** for change
 - **Trade-offs**



Possible Solution

- A pattern-based UE representation framework
 - Identify a limited number UE patterns for particular decision types
 - **UE pattern** – a typical, tried-and-tested collection UEs arranged in a particular fashion in a BP architecture, i.e., it is a **particular way to engage with an AI-based system**
- Customized for particular decisions and organizations
 - Reduces the space of options for UE design
 - Provides **transparency** and **predictability**



UE Patterns for Decisions

- **P1: Supervised learning** – AI (AIA) monitors the work of human decision maker (H)
 - Decisions made by a human expert. AIA is using case data and context + the decision outcome as the input to a supervised learning algorithm.
- **P2: AIA as an Advisor** – Decisions made by a human expert. AIA's recommendations are presented as advice.
- **P3: AIA-made decisions approved by a human** – Human approves/rejects CA's recommendations (per instance).
- **P4: Human informed of AIA-made decisions** – AIA makes decisions, while a human is informed (per instance).
- **P5: AIA's decisions with (batch) human audits** – Humans audit automatically produced decisions once per N number of decision instances, once in a time interval, etc.
- **P6: AIA's decisions with human audits on request** – Humans have the opportunity to audit AIA's recommendations whenever they wish.
- **P7: AIA's decisions with automated self-audits** – No humans are involved by default. Humans can review the self-audits.



Pattern-Based Evolution of User Engagement

- Define **upfront** the conditions (**triggers**) that would cause changes in user engagement (transitions between UE Patterns)
 - Predefined **user engagement evolution paths**
 - Changes – not only towards more automation!
 - Identify situations when to **change to a more automated UE Pattern** – e.g., when the decision quality is at an acceptable level, when user trust in the AIA is adequate, etc.
 - But also specify under which conditions to **revert back to a more manual configuration** – e.g., when decision quality goes below a certain threshold and/or when the business context changes significantly
 - May need to bring humans back into the loop



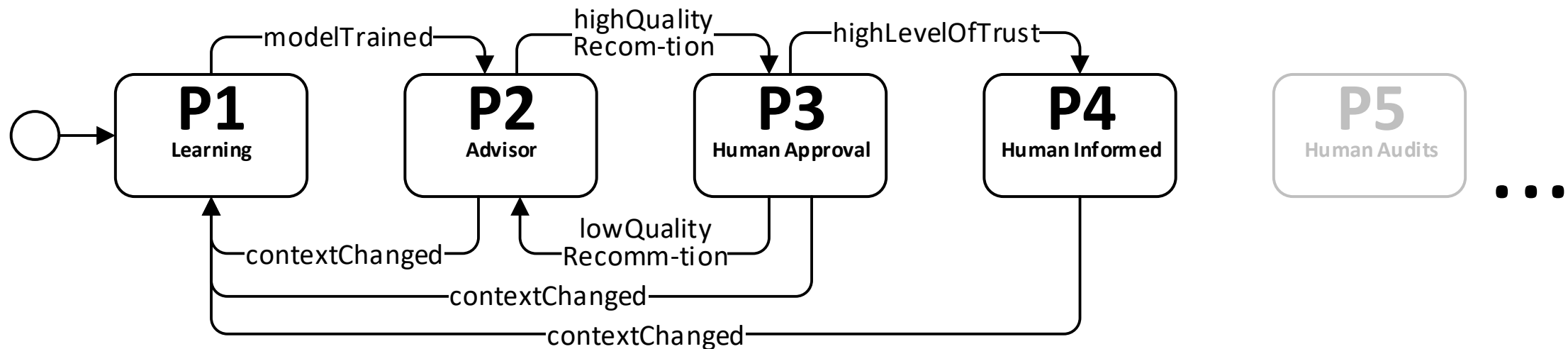
Triggers

- What triggers a change from one UE Pattern to another?
 - **Quality of decisions** produced by an AIA
 - Can use known metrics for algorithm-produced results (confidence levels, AUC ROC, etc.)
 - **Level of trust** – whole organization and/or particular user – in an AIA
 - Difficult to measure automatically
 - May need to have a **manual switch** that would allow users to change UE patterns based on their level of trust
 - **Capability/ability/skills** – not triggers per se, but necessary conditions
 - The ability of AIAs and users to perform certain activities would be the **necessary conditions for some transitions**
 - E.g., ability to audit decisions, having enough training data, etc.
 - **Context** – detect changes that make the model obsolete (model drift)
 - Revert to less automated UE patterns – likely all the way to **P1: Supervised Learning**



Specifying Transitions

- State transition models seem to be a good notation to represent changes in user engagement



Findings

- Automated decision making helps go from **human scale** to **machine scale** decision making
 - Machine scale – decisions are much faster than humans can process them
 - If humans are involved (e.g., reviewing each decision) – still human scale
- Care is needed when introducing automation
 - Many additional processes needed, existing BPs are affected (e.g., responsibility of human users)
 - Human employees will be affected
 - Many new requirements/complexities
 - Context is very important
 - Trust is needed for the solution acceptance
 - **Complex, dynamic user engagement** with AI systems



Natural Language-Driven Enterprise

- Chatbots (and other NL-driven interfaces) are now primarily used to:
 - Get domain-specific or general information using a human-like interface
 - Execute some simple actions (e.g., set up alarms/reminders, play music/video)
- Enterprises are interested in:
 - Creating enterprise-level assistants that
 - Meet enterprise requirements – e.g., know how to handle security, privacy, etc. issues
 - For instance, prevent users from sending certain information outside of the organization
 - Know the context
 - Business domain-level context
 - Company-level context
 - Department-level context
 - Team-level context
 - Personal context of the user
 - Domain vocabulary



Natural Language-Driven Enterprise

- Help employees **instantiate** and even **create processes/workflows** using NL communication
- Identify user intent in all sorts of NL communication (emails, instant messaging, etc.) and map it into executable actions/processes in any of the organizations' systems
 - E.g., producing sales reports from Salesforce, setting up a recurrent meeting with the team on Wednesdays at 1pm
- Create long-running (possibly **custom**) workflows/processes based on the intents in NL communications
 - E.g., initiate a procurement process for a new laptop or a new hiring process
 - These will need to be executed (involving potentially many people), tracked/monitored in the background, etc.
 - Successful custom processes can be adopted by the organization



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