Using the BEAM Notation

This document shall provide quick-start information for using the BEAM notation. This document is structured as follows:

- 1. Setup
- 2. Modeling
- 3. Modeling Risks and Risk Controls
- 4. Example
- 5. Dos and Don'ts Best Practices
- 6. References

1. Setup

The boxology modeler is based on draw.io, a popular diagramming tool. Here are the necessary steps to prepare for modeling your solution:

1. Install or access draw.io

To use draw.io, you have two options:

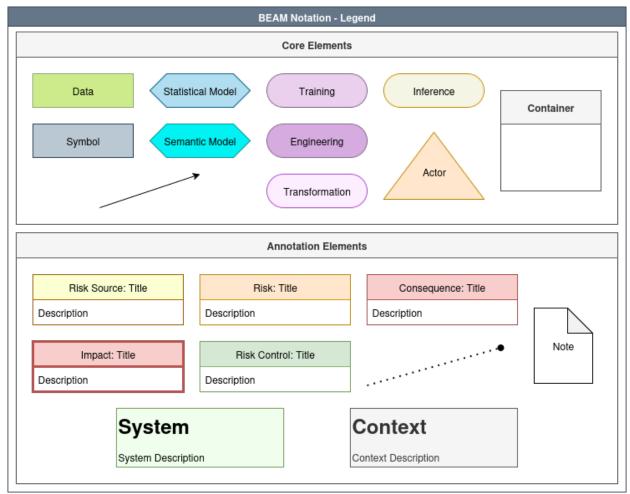
- a) Use the browser version at https://draw.io
- b) Download and install draw.io Desktop from https://get.draw.io
- 2. **Import the BEAM notation library** that contains all the graphical elements
 - a) Download the BEAM notation library from Github [1] in the latest version (file name starts with "beam_lib_v" and has the file ending .xml)
 - b) Import the library (both desktop and online version): File > Open Library from > Device > [location of your downloaded XML file]

3. Start modeling

- a) Access the library on the top-left of your draw.io application in the shapes sidebar
- b) Start by adding the "Legend" component to your drawing by clicking it
- You can now draw from scratch by copy-pasting elements from the legend or by clicking the needed components in the shapes sidebar (top-left in the draw.io application)

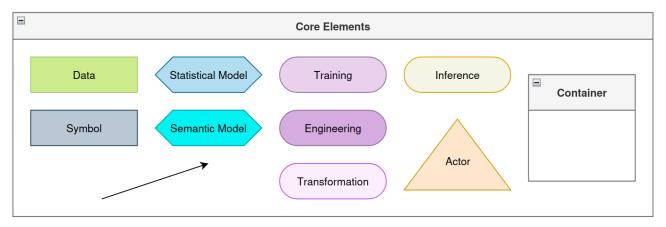
2. Modeling

In the following, we will (briefly) explain the elements provided in the library (the grouping of the components is shown in the "Legend" box). It consists of

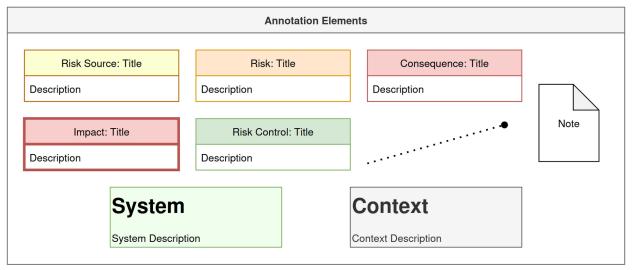


- Core drawing elements used to represent the building blocks of an AI system, including
- a) Input/Output **Data** (green box), e.g. for text, tabular data, images, ...
- b) **Symbol**ic Input/Output Resources (grey box), e.g. for knowledge graphs, ...
- c) Statistical Model (blue hexagon), e.g. for neural networks
- d) **Semantic Model** (turquoise hexagon), e.g. for an ontology
- e) **Inference Process** (light orange rounded rectangle), e.g. for using an ML model for classification or running an ontology reasoner
- f) **Training Process** (purple rounded rectangle), e.g. to represent the training process of a classification model on a given dataset
- g) **Engineering Process** (dark purple rounded rectangle), e.g. to represent the knowledge engineering process in which a knowledge engineer constructs an ontology

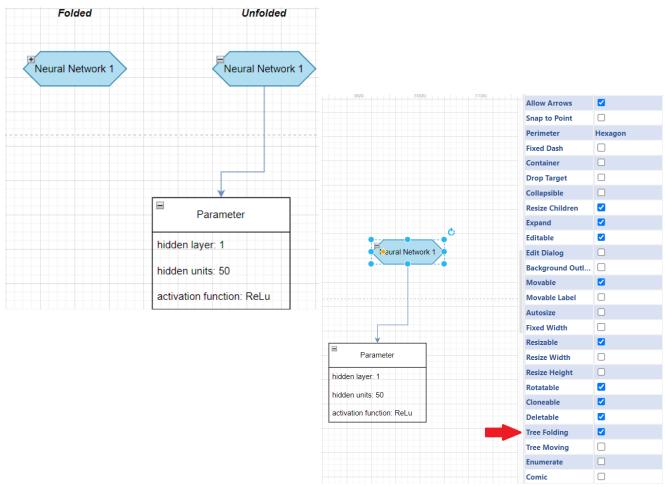
- Transformation Process (pink rounded rectangle), e.g. for data pre-processing or mapping, no new knowledge is generated, these processes are strictly transformational
- i) Actor (orange triangle), e.g. an engineer or user
- j) Container, for scoping different parts of the system
- k) Solid black arrow to draw the connection/workflow between core components



- 2. **Annotation drawing elements** used to annotating your AI system and its components, including
 - a) Note to represent any free-text annotation
 - b) **Risk** to represent risks associated with components or the system
 - c) **Risk Source** to represent what gives rise to risks
 - d) **Consequence** to represent the consequence of a risk if it materializes
 - e) **Impact** to represent the impact of risks/consequences on specific stakeholders/groups
 - f) Risk Control to represent risk controls and mitigation strategies
 - g) **System** and **Context** description to provide textual information of the system as a whole and the context it is operated in
 - h) **Dotted arrow** to associate any of the above annotation elements to the relevant component



3. Show/hide details: In case the model contains details that should not always be visible, e.g. to hide them for an overview, this can be accomplished using the Tree Folding property. The attached annotation details of elements with this property can be collapsed/expanded using the small minus/plus-icon on the element. Note: Unfortunately, draw.io does not support this for annotation details directly attached to container elements.

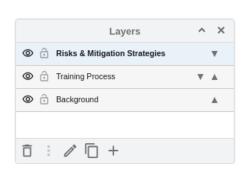


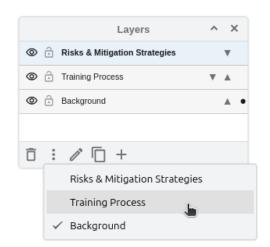
4. **Layers**: In draw.io, elements and edges can be assigned to different layers. This feature provides great flexibility, e.g. the system architecture and data flow can be on one layer, and the risk perspective can be on another. The visibility of layers can

be toggled separately, which enables showing different perspectives of the same model. The layer toolbox can be activated via View > Layers, see below.



In the layer toolbox, layers can be added, edited and removed, as well as reordered. To assign one or more elements and connectors to a layer, first select them and then click the three-dots-menu to select the layer to assign them to.





3. Modeling Risks and Risk Controls

As described in the previous section, BEAM includes special annotation elements for representing risks and their sources, consequences, impacts and controls.

Risk elements are orange, risk source elements are yellow, consequence and impact elements are red (with impact elements having a thicker border) and risk control elements are green. They each have a title bar and a description field. The title by convention should consist of the element type and a short descriptive part, e.g. *Risk Source: Poor Data Quality*.



Risks can be associated to one or more elements using annotation connectors, this includes container elements. System level risks can represented by attaching risk elements to an overarching container that includes the whole system.

Consequence elements are used to model consequences in case a risk materializes. Impact elements model the effects of a consequence on specific stakeholders. Risks, consequences and impacts are to be connected with regular workflow connectors (solid black arrow) to represent their causal relationship.

Example: A medical image classification system can have the risk of inaccurate predictions, whereas the risk source can be poor data quality. A possible consequence is then a false negative classification of a patient's medical scan, which can have a severe impact for the patient in case necessary treatments are not carried out because of the misclassification.

Risk control elements can also be associated to one or more system elements using annotation connectors in the same way as risks (or other annotations) to indicate in which part of the system the mitigation strategy operates. Often a risk mitigation strategy will be associated to a specific risk, risk source, consequence or impact, in which case this connection should also be modeled using a regular workflow connector. Risk control elements are often connected to both a system element (or more than one) and other elements of the risk perspective (risks, risk sources, consequences, impacts).

4. Example

We provided an example of a system in the BEAM notation, which is described below and can be downloaded at [2].

The depicted system is a classifier of fashion attire images. It is used in a recommender system of an online shop that sells fashion items. Users can upload images of fashion attire they like in order to receive similar recommendations.

At the heart of the system is a machine learning model, which is trained on a dataset of labeled fashion attire images. The model is trained to classify which type of fashion attire an image shows, e.g. T-shirt, pullover, ... The BEAM notation depicts the training and inference processes separately.

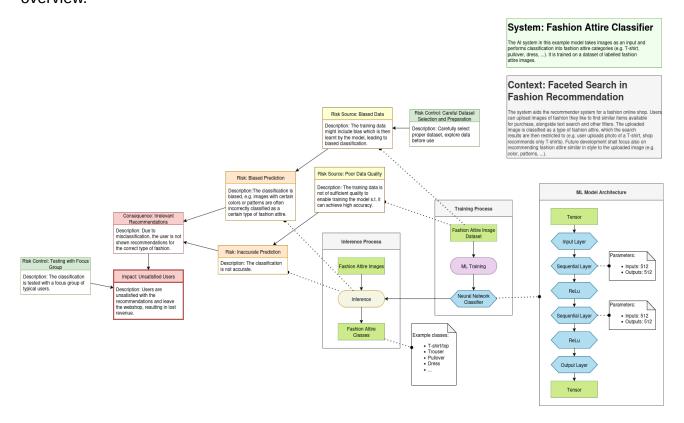
In the training process the aforementioned dataset is used to train the model. The internal model architecture is also depicted (an image is fed into the input layer of the ML model, with data flowing through the various layers of the model until the final layer produces a label as its output).

In the inference process, the ML model is used to classify the input images.

Two risks associated with this system are also modeled in the example. First, the dataset might contain some bias (for example there might be a disproportionate amount of pink dresses, which might cause the model to be biased towards classifying images of pink objects as dresses). This risk can be mitigated by selecting a high-quality dataset and/or additional preparation of the dataset.

The other depicted risk is that customers of the online shop might not be shown relevant items if the uploaded images are classified incorrectly. As this might turn away customers, monetary loss for the owner of the online shop (system provider) is modeled as an impact of this risk. The mitigation strategy for this risk is to test the system with the trained model with a focus group of potential customers and evaluate the results.

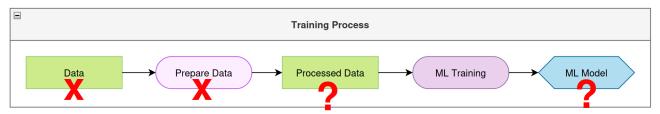
The example furthermore includes a system and context description to provide a quick overview.



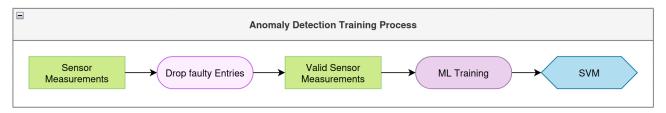
5. Dos and Don'ts - Best Practices

This section gives a few small examples with mistakes and how to avoid them.

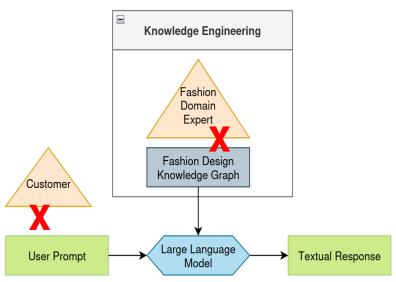
5.1 Example 1 - Unspecific Labels



In this example, very generic labels are used, s.t. the details of the modeled training process remain unclear. What type of data is used? Where does it come from? Data Preparation is too broad of a label, which concrete steps does the data preparation process consist of? Could the processed data be described more specifically? What kind of model is trained here? Below is the same example with more descriptive labels:

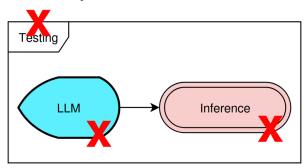


5.2 Example 2 – Missing Connectors



In this example, the actors are not connected to the rest of the model. Since BEAM models shall be automatically processable, the actors should be connected to the relevant elements using workflow connectors. Co-occurrence in the same container or proximity to the relevant elements are not enough – the workflow must be explicitly modeled.

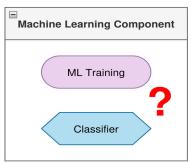
5.3 Example 3 – Unknown Shapes



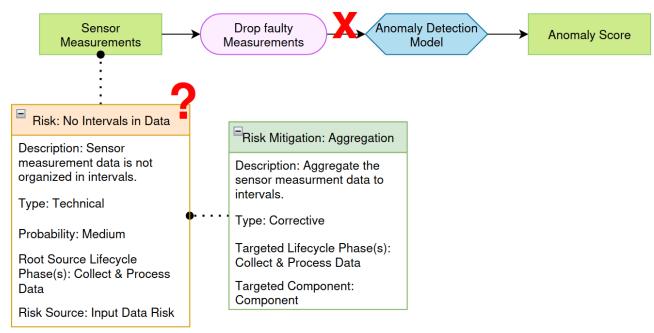
In order for BEAM models to be understandable for humans and automatically processable, elements that are not from the library shall not be used. This includes elements from other well-known modeling languages, such as UML.

5.4 Example 4 - Missing Elements and Connectors

As mentioned in a previous example, an explicit connection between the training process and the classification model is missing. Furthermore, this example is missing input to the training process – which data is the classifier trained with?

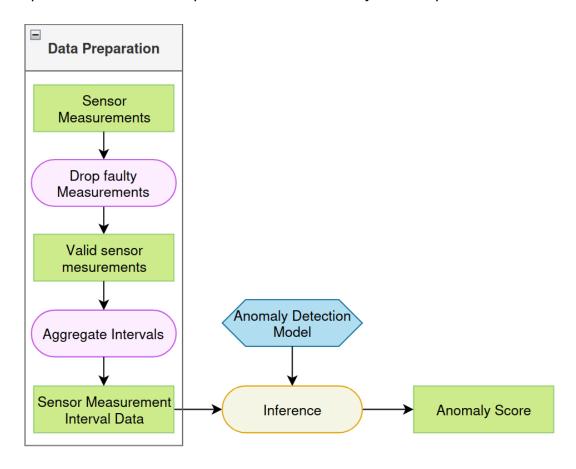


5.5 Example 5 - Missing Elements & Risk vs. Process

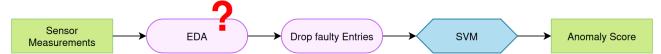


There are multiple issues in this example. Firstly, in case the input data (Sensor Measurements) is not always aggregated in a useful way, but there is an aggregation step, this should be modeled as a process and not as a risk with a mitigation strategy. Secondly, a process for data preparation (Drop faulty measurements) connects directly to the model

and it is only implied that there is an inference process. The example below correctly represents the inference process and intermediary data output.



5.6 Example 6 – Represent AI system, not Engineering Process



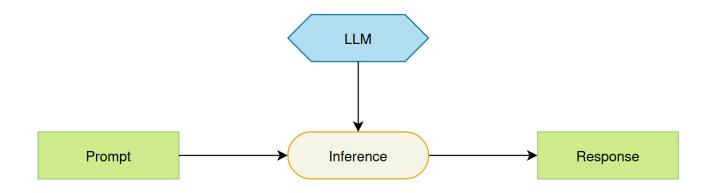
In this example, exploratory data analysis (EDA) is included as a process in the workflow. BEAM is intended to model AI systems, not the engineering process behind it. It might also be helpful to ask what the output of a process is and to model it explicitly. In the case of EDA, the results might be descriptive statistics, data visualization and insights gained about the data – is this used in the AI system or in the engineering process?

Furthermore, the mandatory process type/prefix is omitted for the EDA process – perhaps because neither transform nor infer or generate were a good fit – a possible hint that it might not belong in the BEAM model.

5.7 Example 7 – Data Flow through Model



In this example, the data flow is modeled to go through a model. This is not the intended use of the BEAM notation – rather, the dat flow should go through a process, in which a model is used, see below.



6. References

[1] https://github.com/wu-semsys/beam_tutorial/

[2]

 $\underline{\text{https://github.com/wu-semsys/beam_tutorial/blob/main/beam_example_fashion_attire_classification.xml}$