



ICEBERG Interim Workshop

# **Interactions Between Imbalance Settlement and Cross-Border Balancing**

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**Introduction**

**Model Description**

**Analysis Settings**

**Simulation Results**

**Conclusions**



## **Introduction**

Regulatory Context

Imbalance Settlement Price

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Model Description

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# Introduction

## Regulatory Context

- ❑ **Imbalance Settlement** constitutes a core element of the balancing market clearing procedure
  - Charges or remunerates Balancing Responsible Parties for their imbalances
  - European directives aim at the harmonization of the imbalance settlement procedures in all regions
  
- ❑ Imbalance settlement price is an important element since it provides adequate economic signals to BRPs to either balance their position or help the system to return to balanced conditions
  - May follow the single or dual pricing methodology
  - Applied over the duration of an imbalance settlement period, i.e. 15 minutes
  
- ❑ In Belgium there are currently two major approaches:
  - ❑ The regulator CREG aims to enhance the **efficient operation** of the pan-European system
  - ❑ The transmission system operator ELIA is concerned about the **system security**

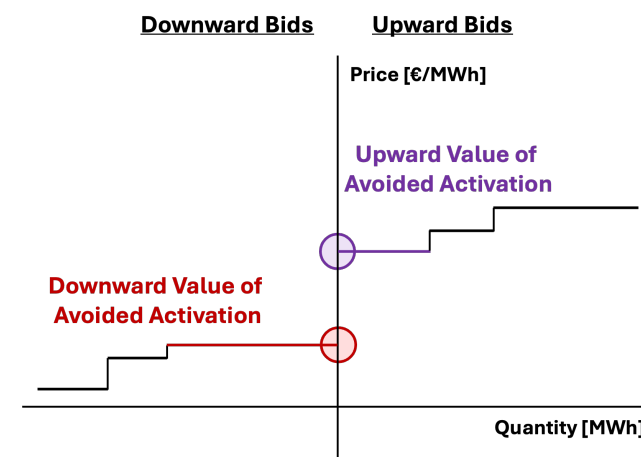
# Introduction

## Imbalance Settlement Price – Price Components

- **Weighted Average Price (WAVG):** Weighted average, with regard to absolute system imbalance, Cross Border Marginal Price
  - IMB: System Imbalance
  - CBMP: aFRR Cross Border Marginal Price

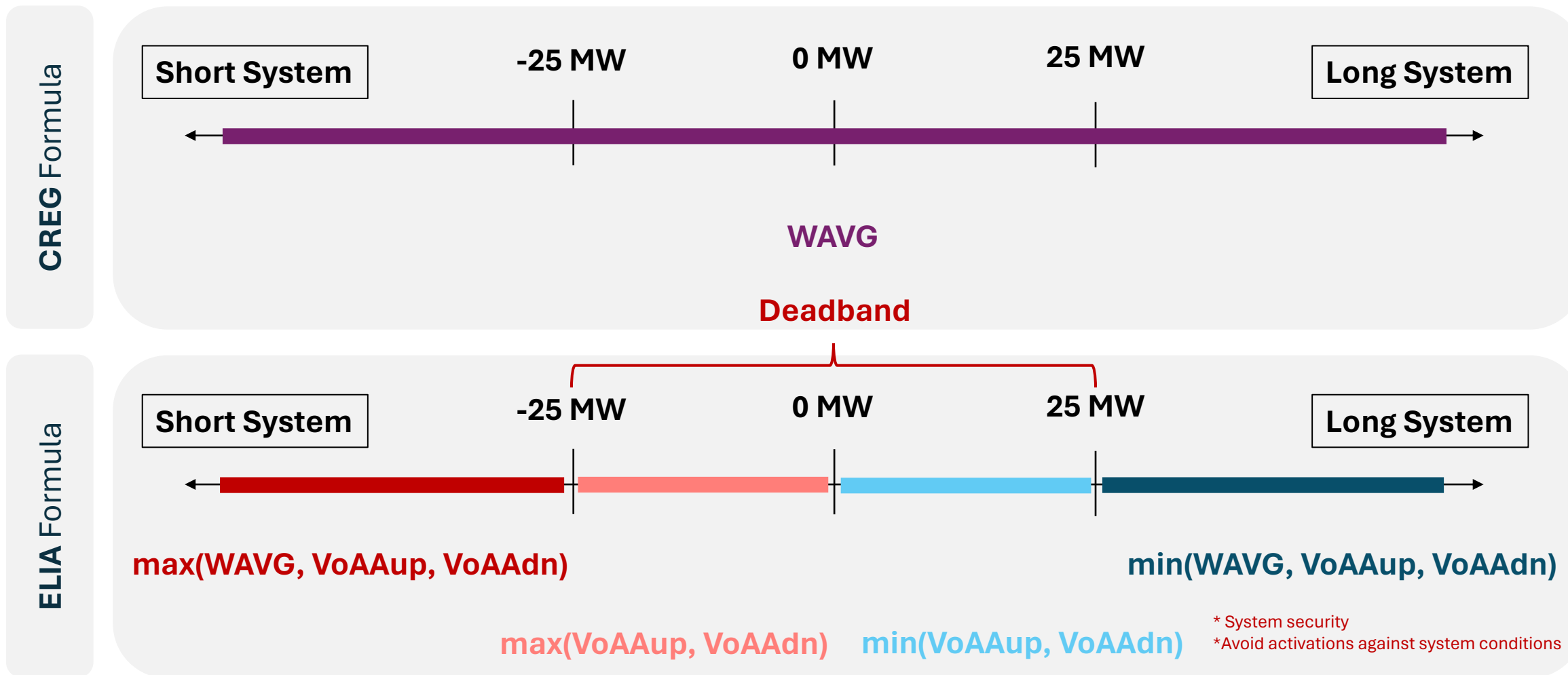
$$\text{WAVG} = \frac{\sum_t(\text{abs}(\text{IMB}_t) * \text{CBMP}_t)}{\sum_t(\text{abs}(\text{IMB}_t))}$$

- **Value of Avoided Activation in Upward/Downward direction (VoAAup and VoAAdn):** Value of the first bid for providing Upward and Downward balancing energy



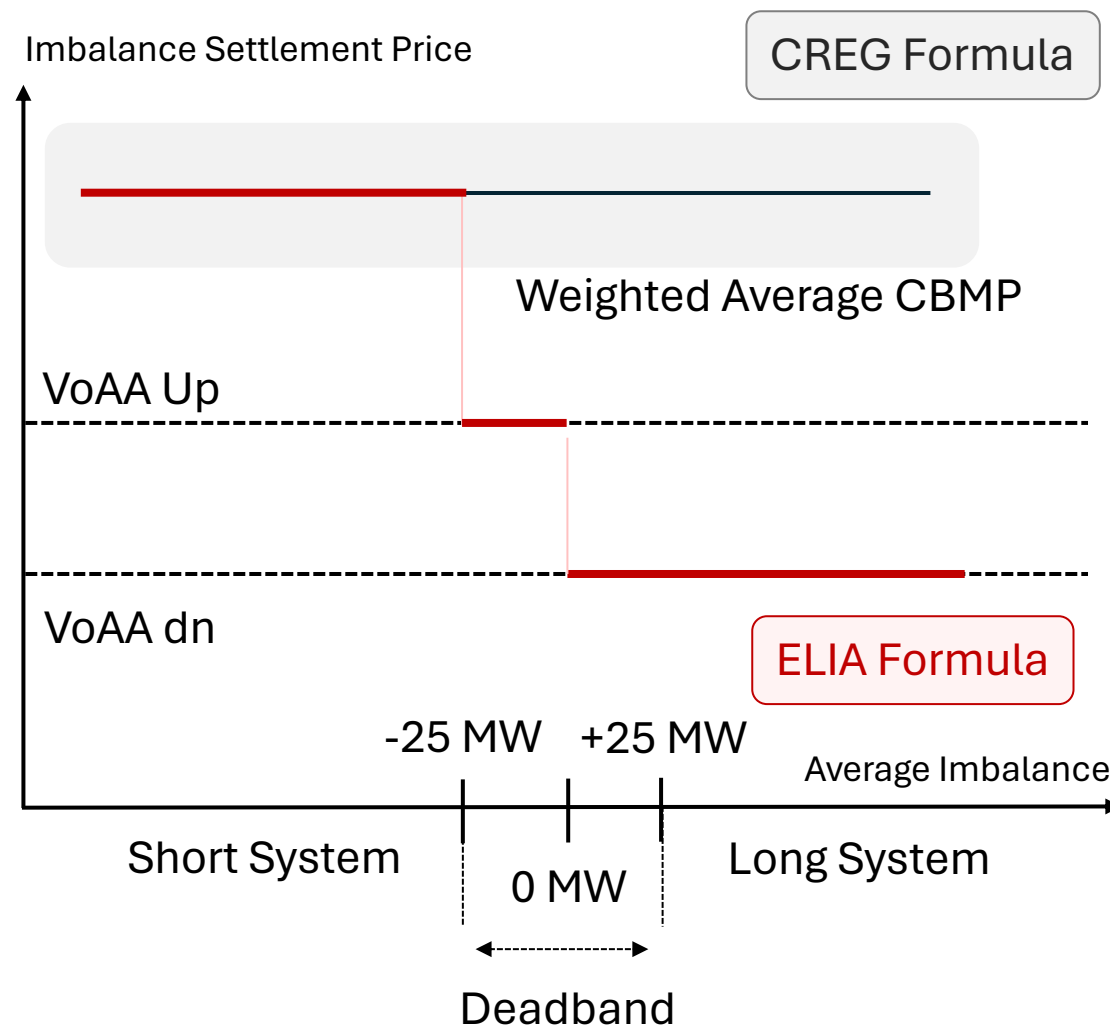
# Introduction

## Imbalance Settlement Price – Existing Formulas



# Introduction

## Imbalance Settlement Price – Graphical Representation



# Introduction

## Project Scope

Participant behavior against the various imbalance price settlement schemes

Model

Quantify

Efficient pan-European system operation and system security

Identify

Different granularity between the imbalance settlement period and the underlying price components





## Introduction

## Model Description

Imbalance Time-Series Generation Model

Balancing Platform

Reinforcement Learning Model

## Analysis Settings

## Simulation Results

## Conclusions

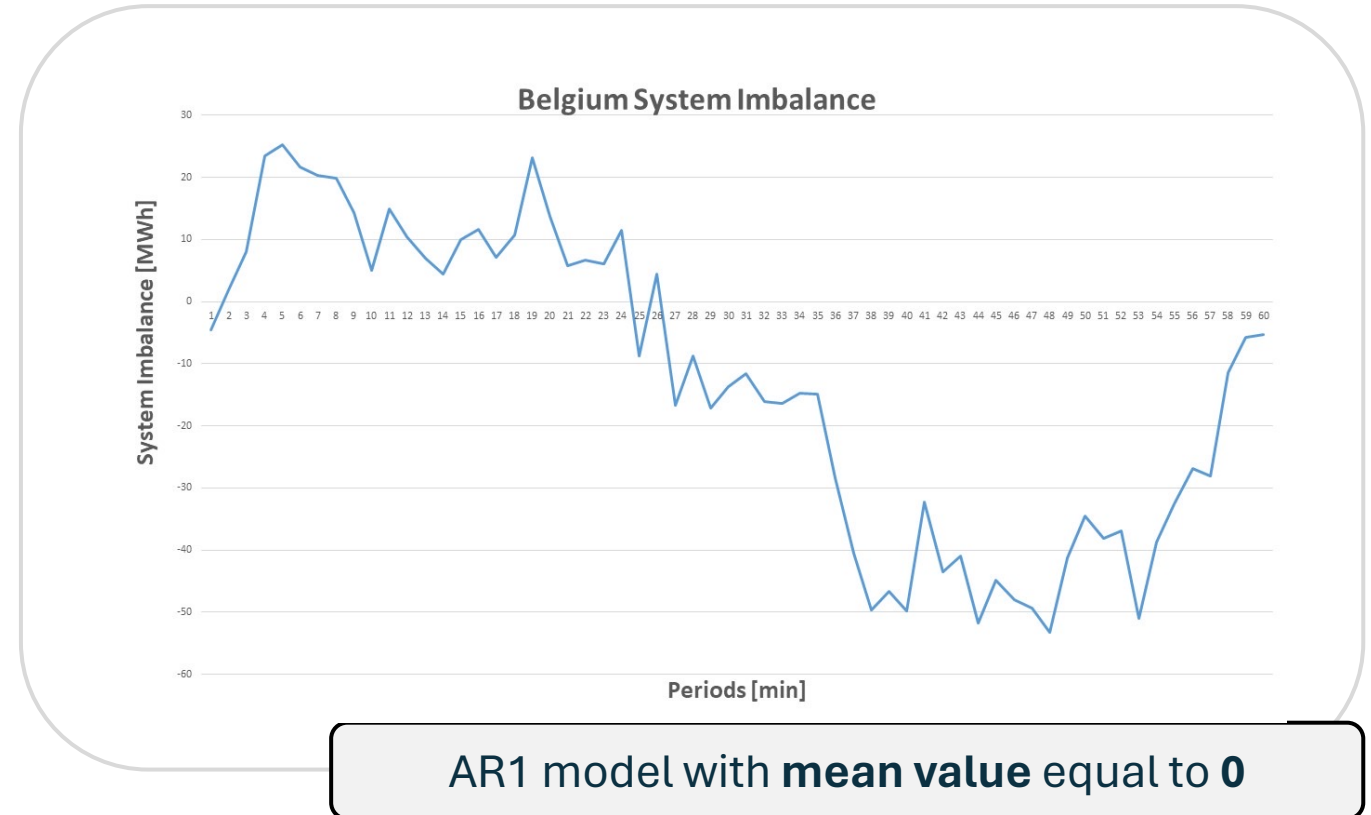
# Model Description

## Imbalance Time-Series Generation Model

- ❑ In the analysis, a **first-order stationary autoregressive model (AR1)** was used, for creating each bidding area imbalance

$$Imb_t = \varphi \cdot Imb_{t-1} + \varepsilon_t$$

- ❑  $\varepsilon_t$  is the white noise  $\rightarrow$  sampled from a normal distribution
- ❑ A continuous imbalance is created, for **X** minutes  $\rightarrow$  separated in **X/15** scenarios  $\rightarrow$  training/testing of the reinforcement learning model



# Model Description

## Balancing Platform

**min** {Balancing Energy Procurement ( $B_{eup}$ ,  $B_{Edn}$ )}

**s.t.**

Balancing Energy Activation Limits ( $B_{eup}$ ,  $B_{Edn}$ )

Zonal Power Balance Constraint ( $E_{aa'}$ ,  $B_{eup}$ ,  $B_{Edn}$ ) : ( $\lambda$ )

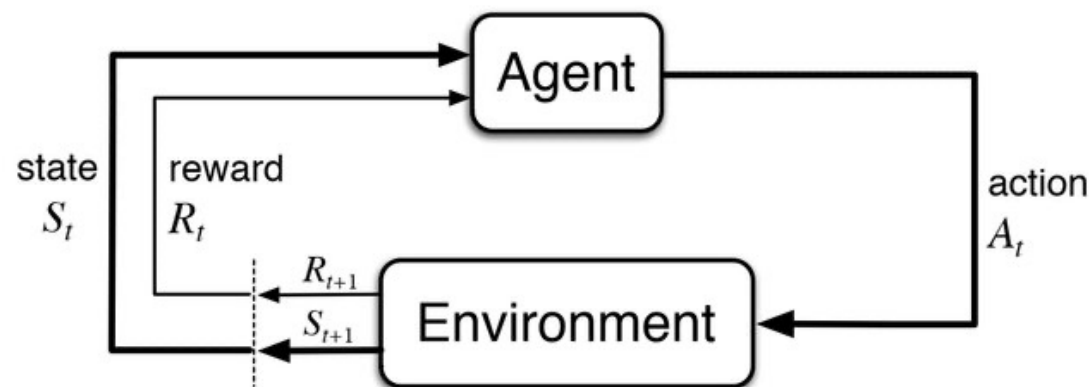
Available Transfer Capacity (ATC) Limits ( $E_{aa'}$ )

Simplified version of the PICASSO algorithm  
Linear Programming Model

- ❑ Balancing energy procurement (=)
  - cost of upward balancing energy ( $B_{eup}$ ) activation
  - (-)
  - utility of balancing energy down ( $B_{Edn}$ ) activation
- ❑ Negative/Positive imbalance translates to short/long system conditions
- ❑ Additional slack variables have been incorporated to ensure problem feasibility
- ❑ Cross Border Market Price: **dual multiplier  $\lambda$**  of the power balance constraint for each 1-minute

# Model Description

## Reinforcement Learning Model (1/5) – Basic Concepts



\* Source: Richard S Sutton and Andrew G Barto. Reinforcement learning: An introduction. MIT press, 2018

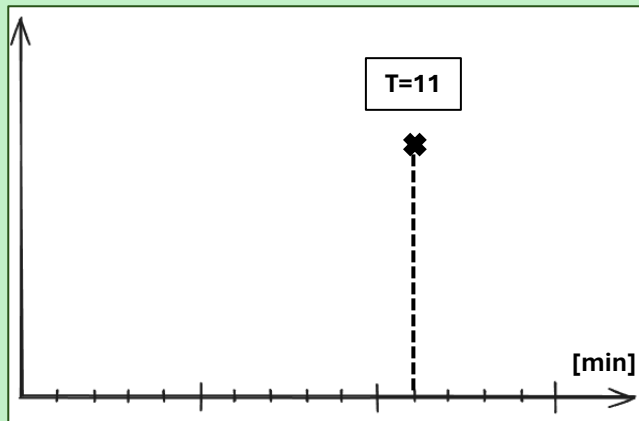
- ❑ The agent operates in a predefined environment that is described by the **state variables**
  - In this study: the course of the 15-minute imbalance settlement period
- ❑ **Actions** constitute the decision of the agent
- ❑ The result of this action is quantified through a **reward** → performance of the agent to its overall goal
- ❑ Agent is trained against a sufficient large number of 15-minute imbalance and price scenarios → **episodes**

# Model Description

## Reinforcement Learning Model (2/5) – Examined States

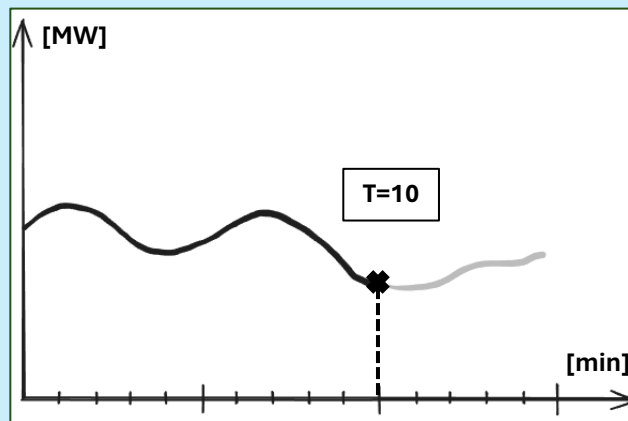
### Trading Period

Current Period



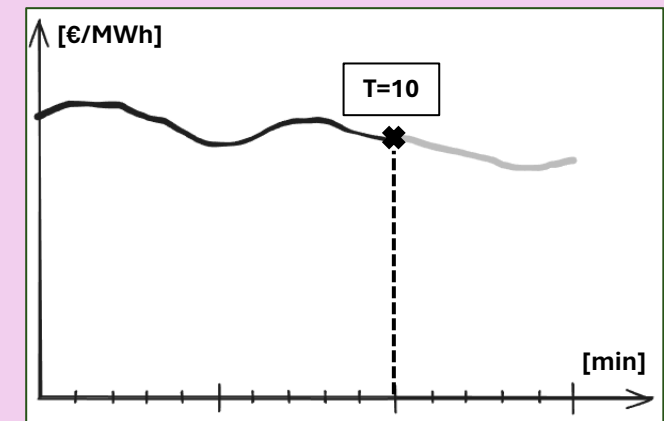
### Average Imbalance

Up to the Previous Period



### Weighted Average Price

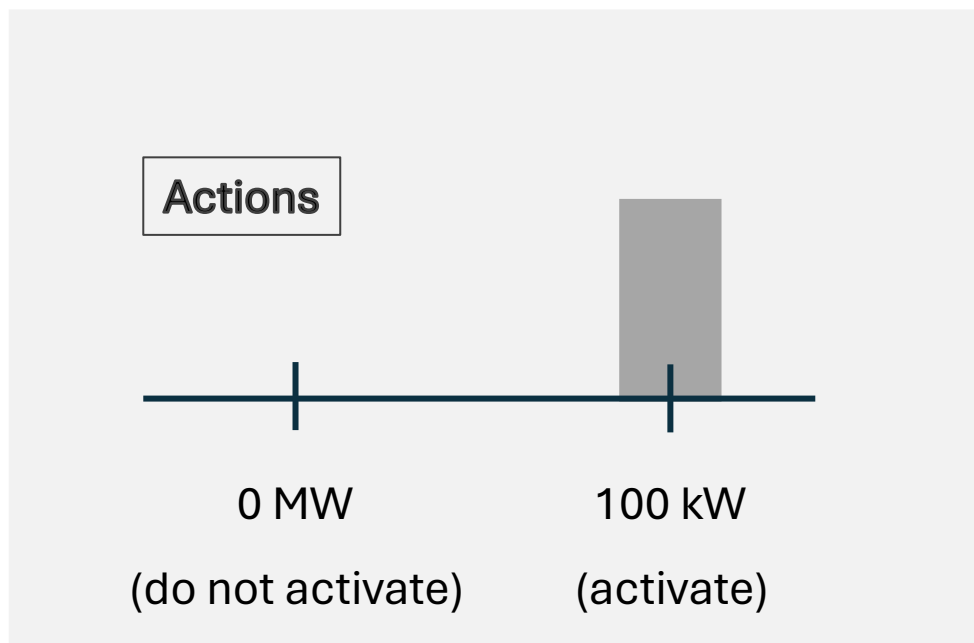
Up to the Previous Period



- Other states may also be taken into consideration, i.e.
  - Available capacity of the interconnecting line
  - Available agent capacity
  - Other
  
- Trade-off between agent behavior modelling and computational tractability

# Model Description

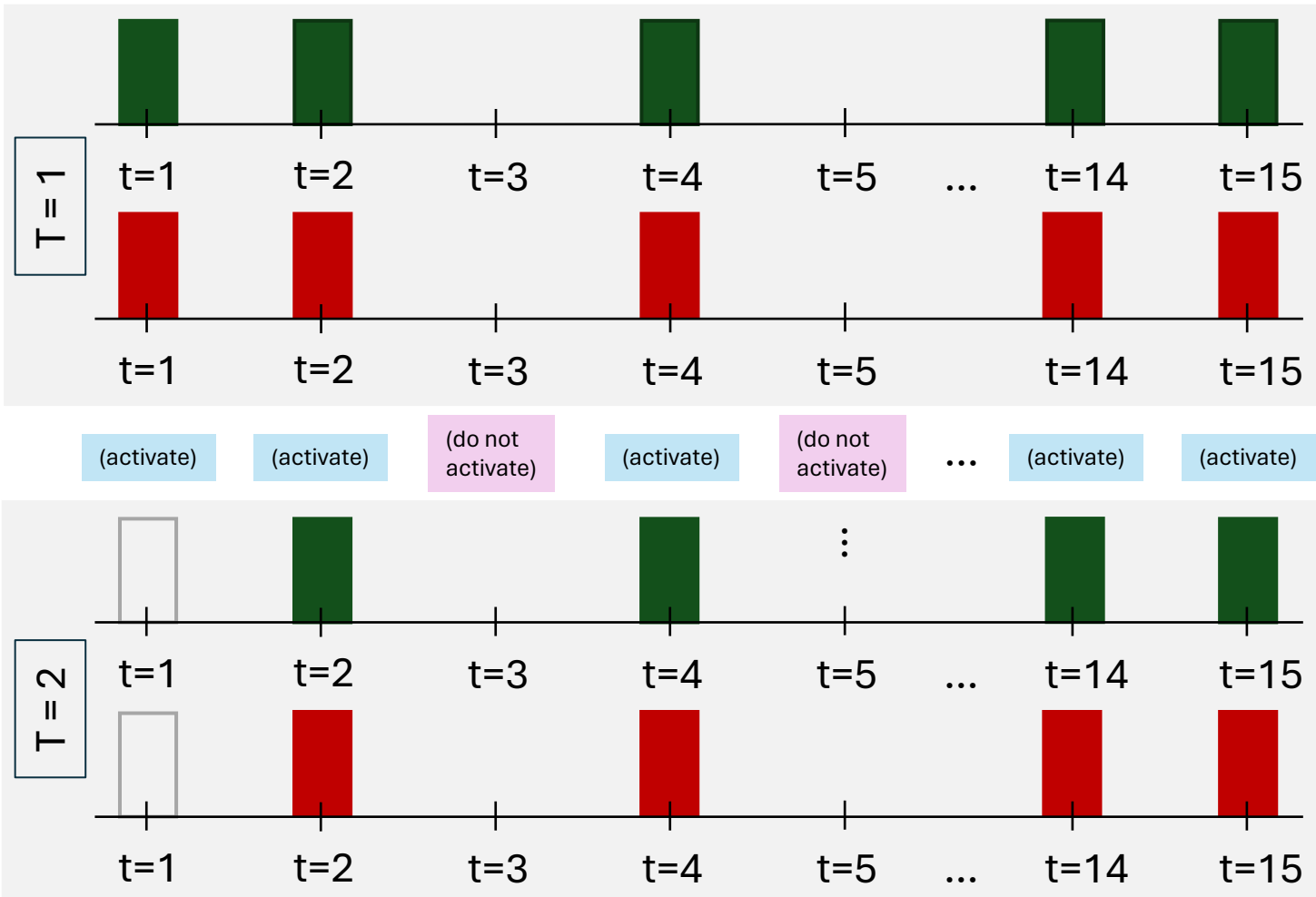
## Reinforcement Learning Model (3/5) – Examined Actions



- ❑ Only two actions were examined to ensure computational tractability
- ❑ In each episode (scenario):
  - Optimal actions for each state vector are selected → **maximum Return**
  - Actions are updated based on the  **$\epsilon$ -greedy algorithm** → There is a possibility that another action will be selected with probability  $\epsilon/(\text{No of Actions})$

# Model Description

## Reinforcement Learning Model (4/5) – Calculation of Rewards



- Agent maximizes the expectation of **return**  $G_t$  over all following **rewards**  $R_t \dots R_T$

$$G_t = R_t + R_{t+1} + \dots + R_T$$

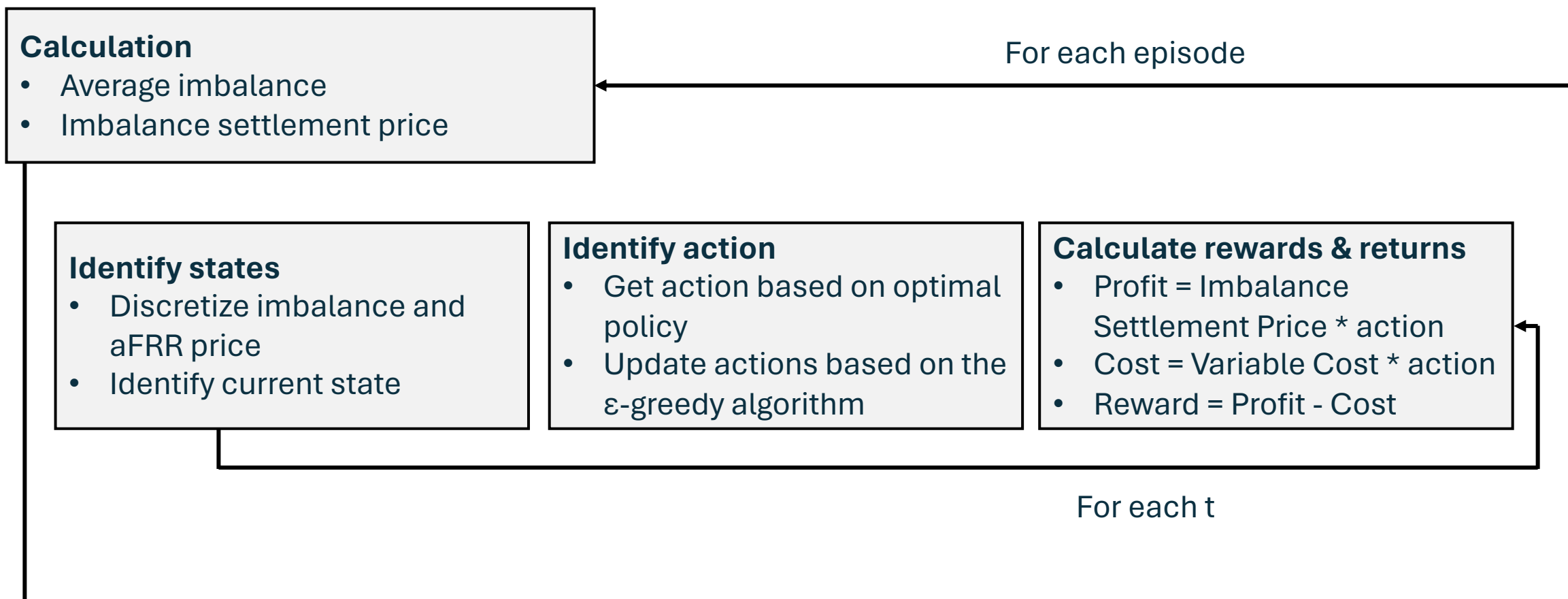
Where:

$$G_t = \sum_{\tau=t}^T p^{imb} \cdot a_{\tau} - \sum_{\tau=t}^T VC \cdot a_{\tau}$$

- The return  $G_t$  is used to update the overall return vector for the examined state and action
- Optimal action  $\alpha_t$  for an examined state is selected based on a policy  $\rightarrow$  maximum return

# Model Description

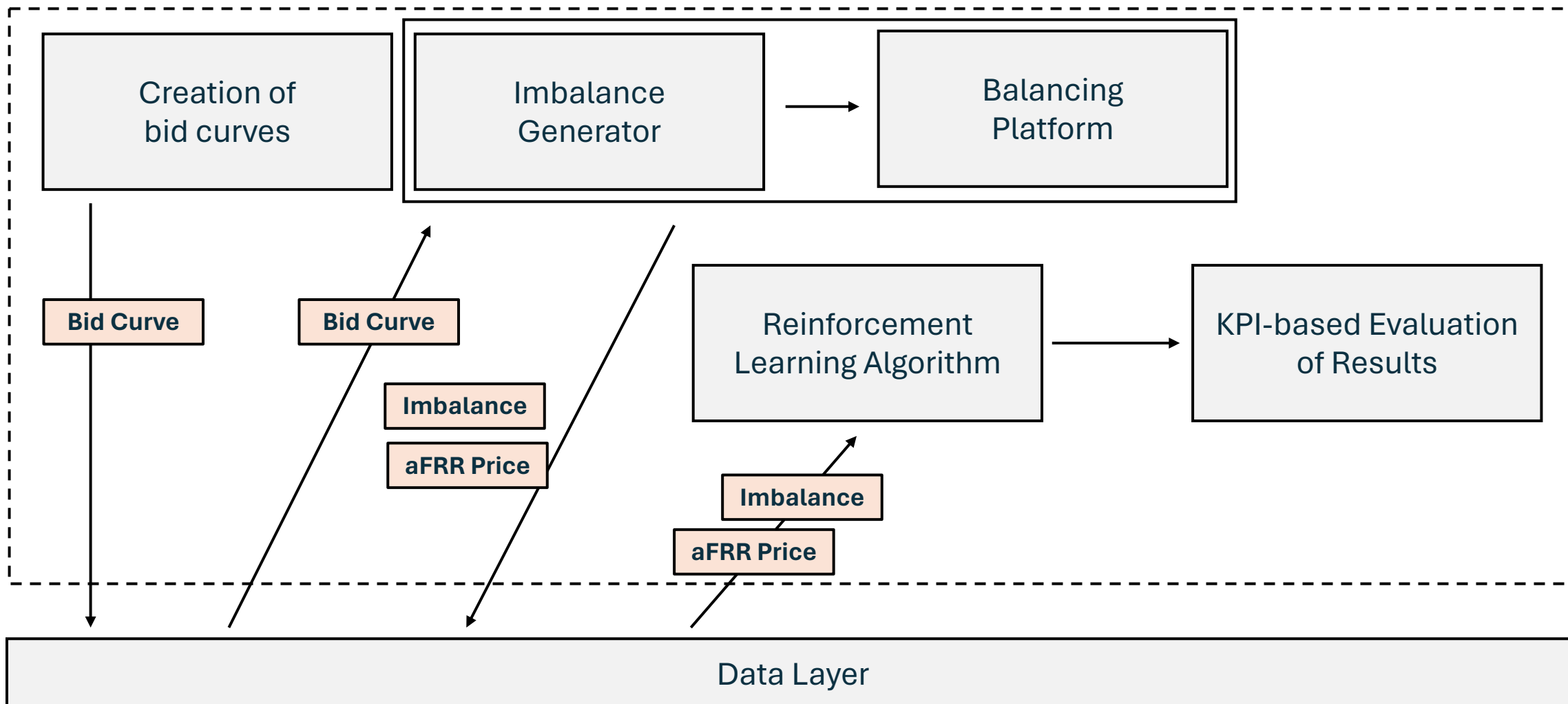
## Reinforcement Learning Model (5/5) – Overall Algorithm





# Model Description

## Developed Platform





## Introduction

## Model Description

## **Analysis Settings**

Settings in Two-Zone System

Bid Generation Procedure

Imbalance Settlement Price Formulas

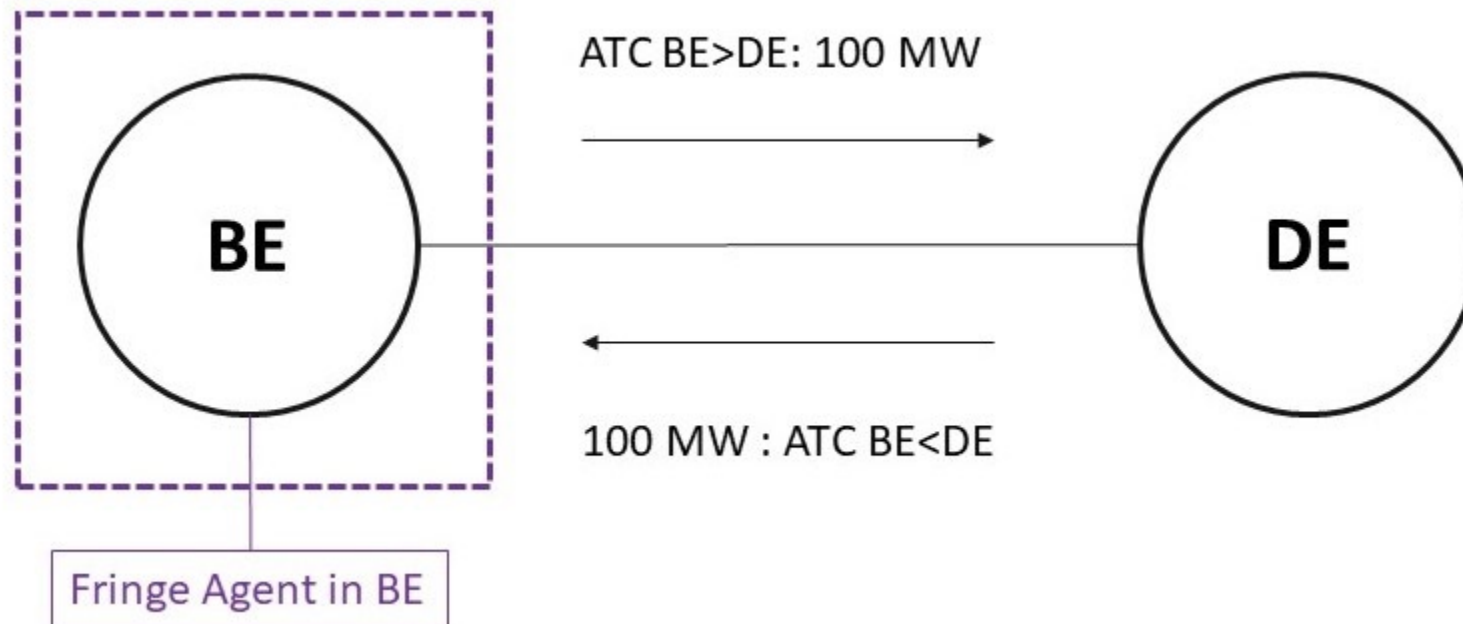
Definition of Key Performance Indicators

## Simulation Results

## Conclusions

# Analysis Settings

## Settings in Two-Zone System (1/2) – General Options



- Base case ATC between the two areas → 100 MW on both directions
- Basic parameters for our models
- Algorithm trained against a set of 1,000,000 scenarios and tested against additional 1,000,000 scenarios

# Analysis Settings

## Settings in Two-Zone System (2/2) – Step Discretization

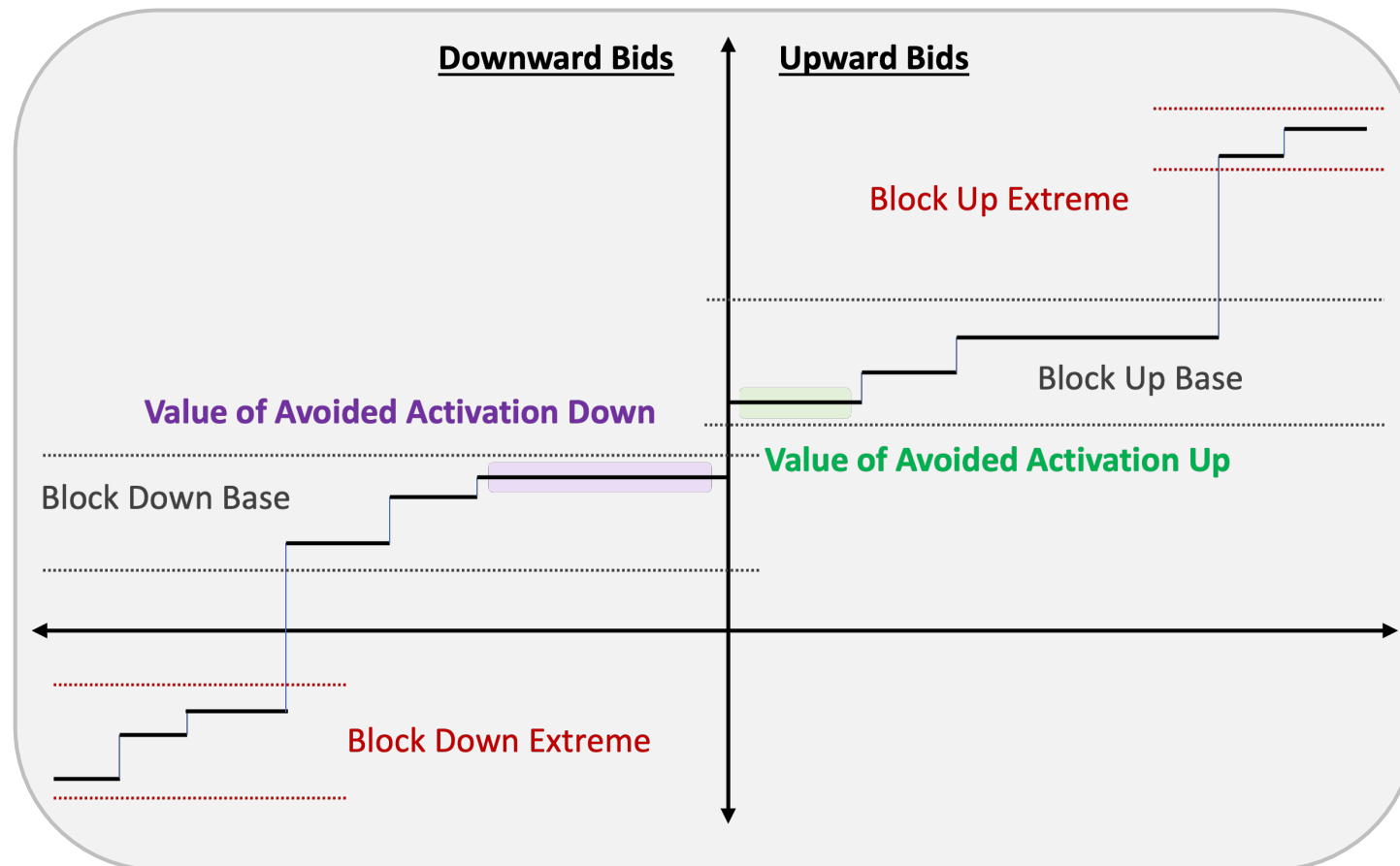
Imbalance Discretization [MW]		
From	To	Step
-10,000	-200	10,000
-200	200	5
200	10,000	10,000

Price Discretization [€/MWh]		
From	To	Step
-10,000	-500	1,000
-500	100	50
100	600	5
600	1,000	50
1,000	10,000	1,000

- Step discretization to ensure computational tractability
- Smaller steps in the imbalance and price areas with more expected instances

# Analysis Settings

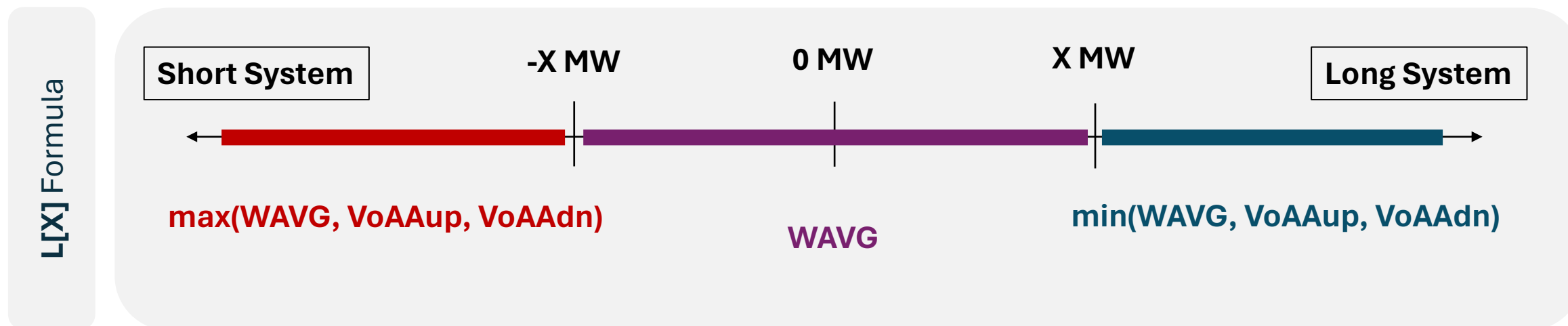
## Bid Generation Procedure - Methodology



- Bid generation procedure that creates aFRR random bids for both Belgium and Germany
- Bids are separated in blocks based on the **direction** (upward/downward) and their **price** (base/extreme)
- Basic bid curve parameters:
  - Total offered quantity
  - Minimum offered price
  - Maximum offered price
  - Total number of available bids

# Analysis Settings

## Imbalance Settlement Price Formulas – Alternative Pricing Scheme



- ❑ **Alternative Pricing Formula:** Up until a predefined level L (i.e. 10MW) the imbalance settlement price is calculated based on the CREG formula (only the Weighted Average Price) and above this threshold it is based on the ELIA one (max/min of the three price components)
- ❑ Attempting to find a compromise between the European regulations and the TSO's objective to safeguard the system security

# Analysis Settings

## Definition of Key Performance Indicators

### Efficient System Operation

- KPI-1:** Percentage of time that the interconnecting line is congested in the BE-DE direction
- KPI-2:** Percentage of time that the interconnecting line is congested in the DE-BE direction
- KPI-3:** Conditional percentage of time when the interconnecting line is congested in the BE-DE direction and the examined fringe agent opts to increase its production
- KPI-4:** Percentage of time that the agent is helping the short pan-European system

- KPI-5:** Percentage of time that the interconnected line is congested in the BE-DE direction, the fringe agent is activated, and the downward bid curve in Belgium is depleted
- KPI-6:** Percentage of time that the upward Belgium bid curve is depleted
- KPI-7:** Percentage of time that the downward Belgium bid curve is depleted
- KPI-8:** Percentage of time that the agent is activated when the downward cleared quantity is more than X% of the downward bid curve

### System Security



## Introduction

## Model Description

## Analysis Settings

## **Simulation Results**

Two-Zone System Analysis

KPI-based Analysis

Available Transfer Capacity Effect

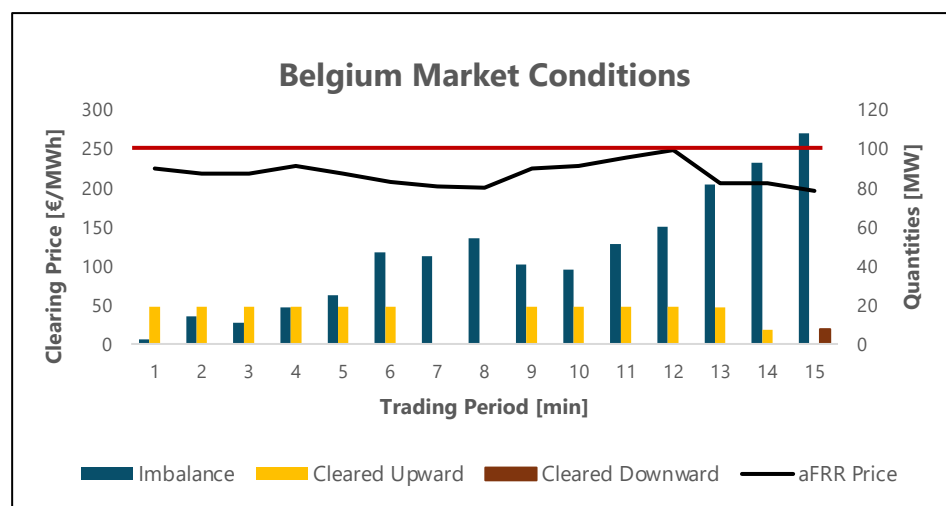
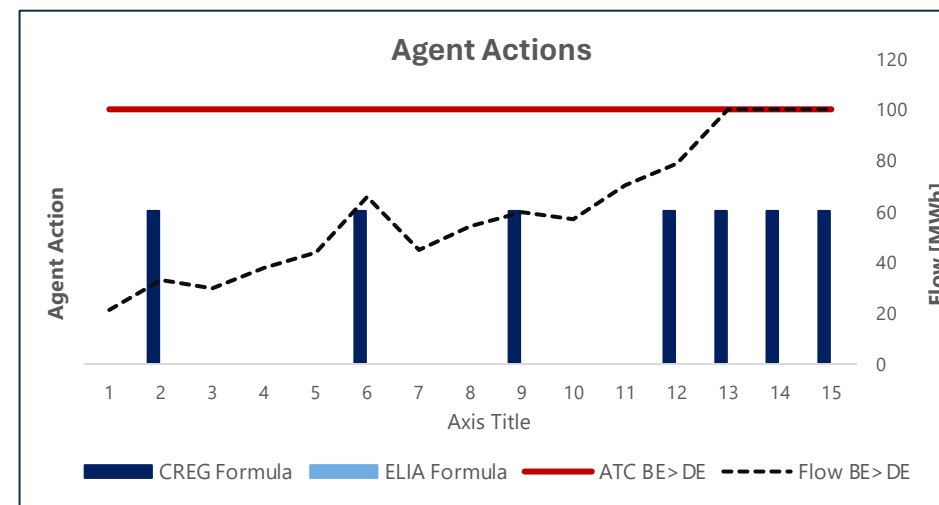
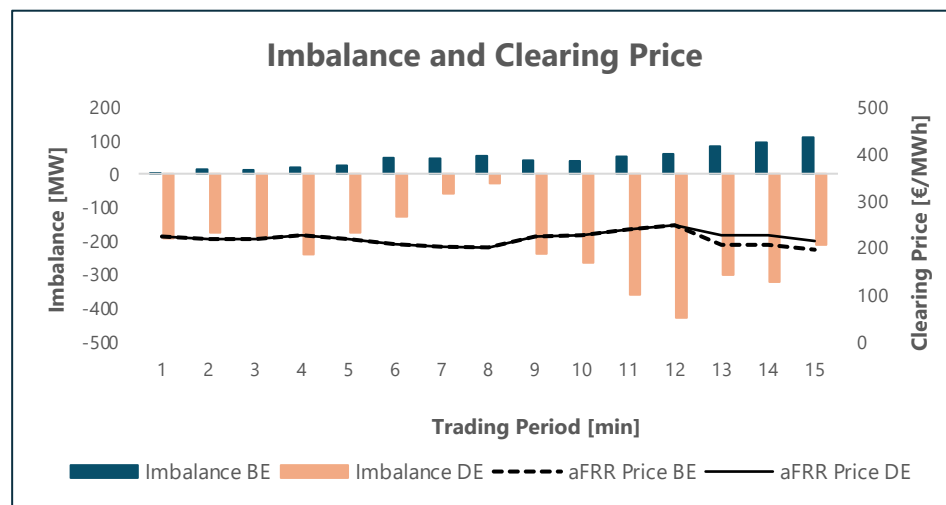
Imbalance Settlement Period Granularity Effect

## Conclusions



# Simulation Results

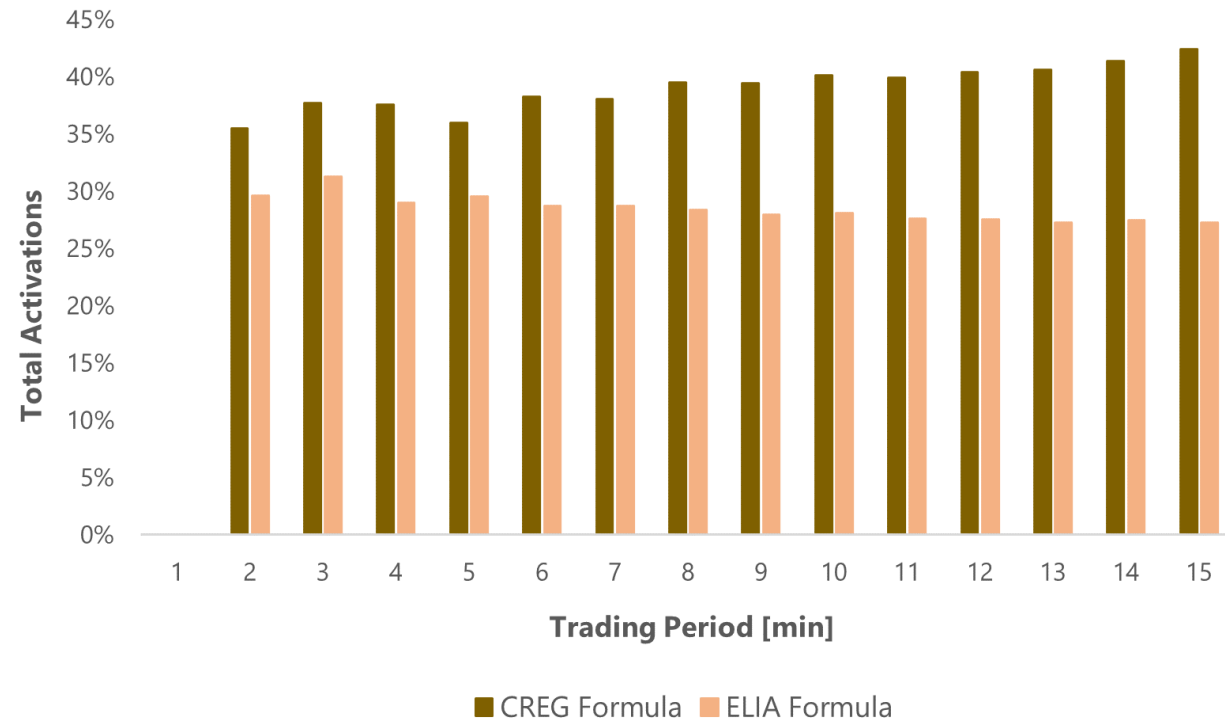
## Two-Zone System Analysis



- Even though the Belgium system is long, upward energy is activated to help the overall long system
- Belgium price is set due to the activation of upward bids that are submitted either in Germany or in Belgium
- Long system → ELIA formula provides incentives to not activate

# Simulation Results

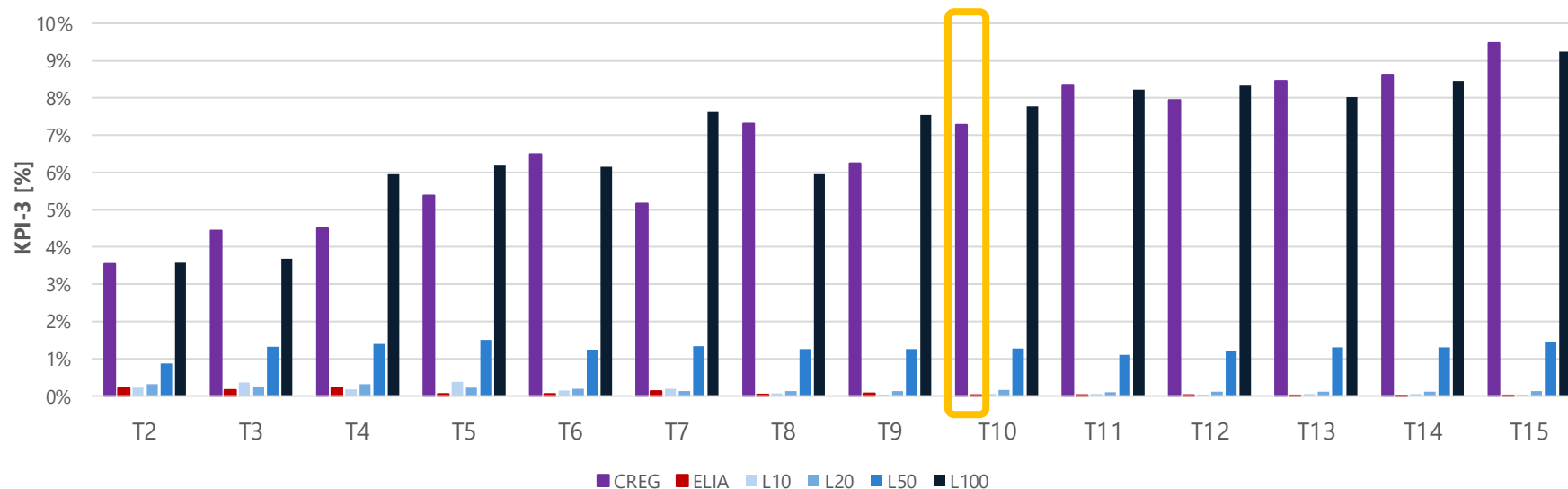
## Two-Zone System Analysis – Total Agent Activations



- ❑ **CREG** formula activates the agent more during the end of the interval → activations associated only with the **weighted average price**
- ❑ **ELIA** formula activates the agent less as we move closer to the end of the interval → activations depend both to the **weighted average price** and the **average imbalance**

# Simulation Results

## KPI-based Analysis – KPI-3 Results

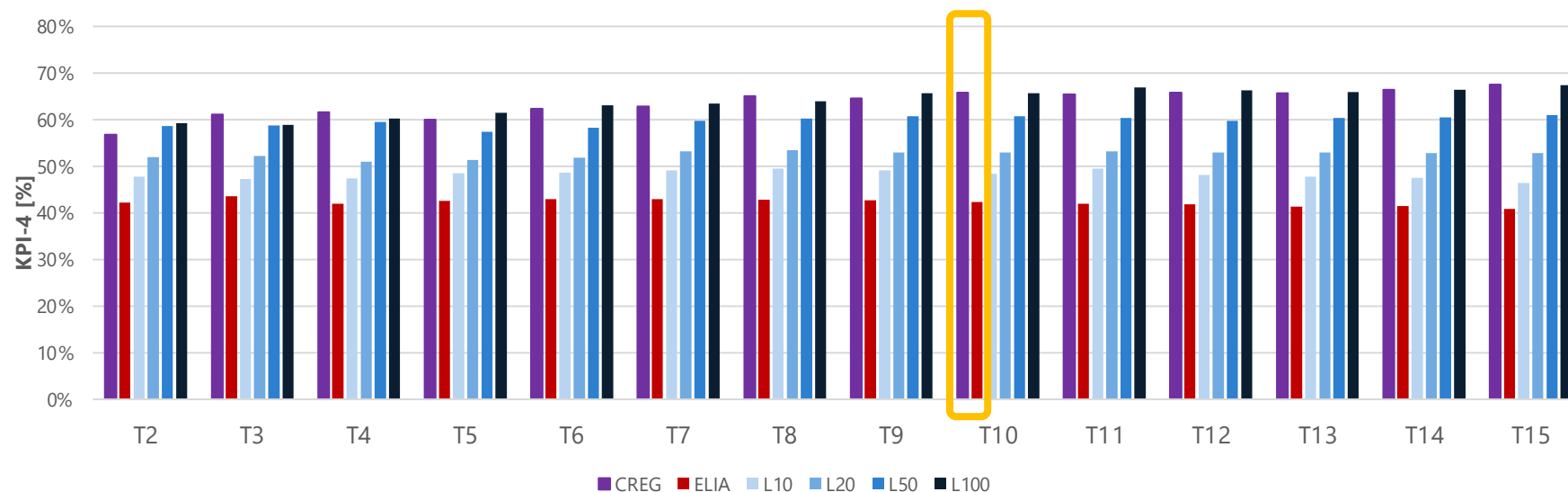


***KPI-3: Conditional percentage of time when the interconnecting line is congested in the BE-DE direction and the examined fringe agent opts to increase its production***

- CREG** formula leads to an increased activation of the agent relative to the **ELIA** one
- CREG** formula activates the agent more during the end of the interval but no more than 10%
- L50** formula provides similar to the ELIA formula results

# Simulation Results

## KPI-based Analysis – KPI-4 Results



### ***KPI-4: Percentage of time that the agent is helping the short pan-European system***

- CREG** formula outperforms the **ELIA** formula → aims at containing the activation of Belgian resources for local imbalances only
- In the CREG formula the agent is activated more during the end of the ISP interval → agent is reacting to average and not instantaneous conditions

# Simulation Results

## KPI-based Analysis – KPI-5 & KPI-8 Results

***KPI-5: Percentage of time that the interconnected line is congested in the BE-DE direction, the fringe agent is activated, and the downward bid curve in Belgium is depleted***

- Equal to 0% in all minutes of the examined ISP
- The system is not endangered

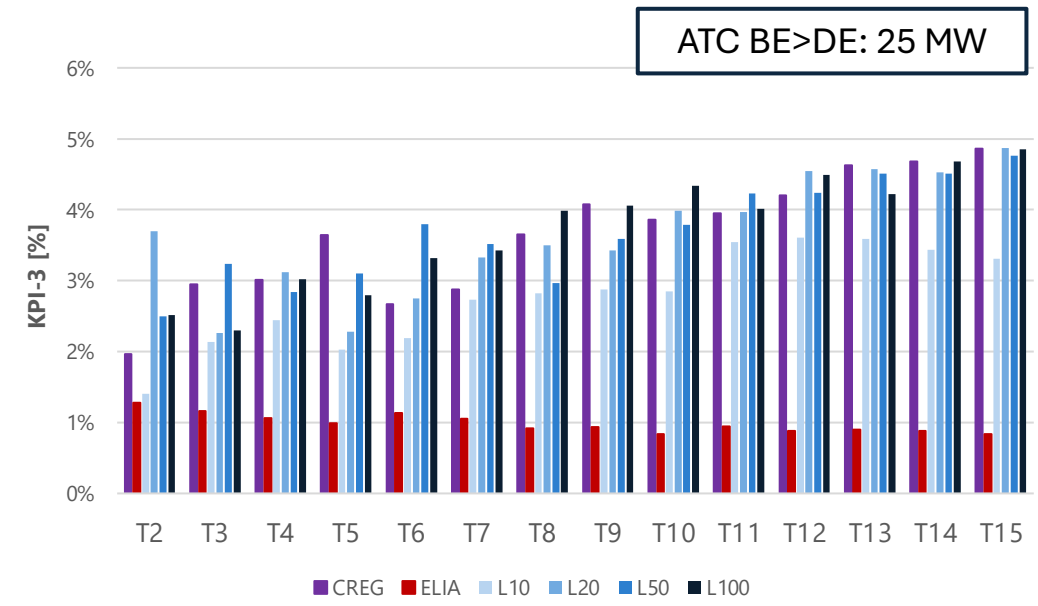
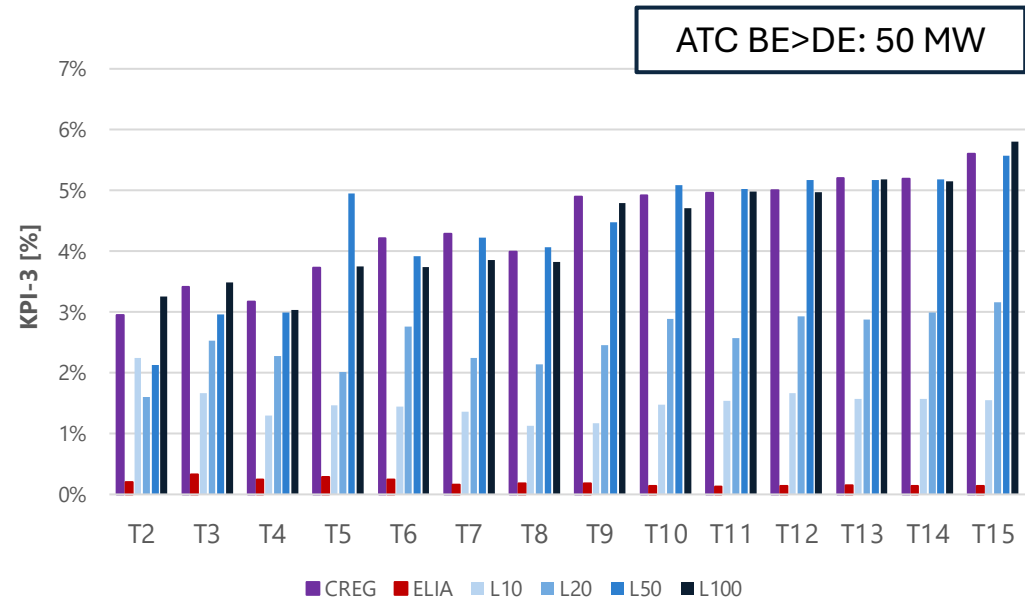
X	30	50	70	90
CREG	0.04%	0.01%	0.00%	0.00%
ELIA	0.12%	0.02%	0.00%	0.00%
L10	0.10%	0.01%	0.00%	0.00%
L20	0.08%	0.01%	0.00%	0.00%
L50	0.05%	0.01%	0.00%	0.00%
L100	0.04%	0.01%	0.00%	0.00%

***KPI-8: Percentage of time that the agent is activated when the downward cleared quantity is more than X% of the downward bid curve***

- Almost half the downward bid curve is available, even when the agent is activated
- Overall system security is ensured

# Simulation Results

## Available Transfer Capacity Effect



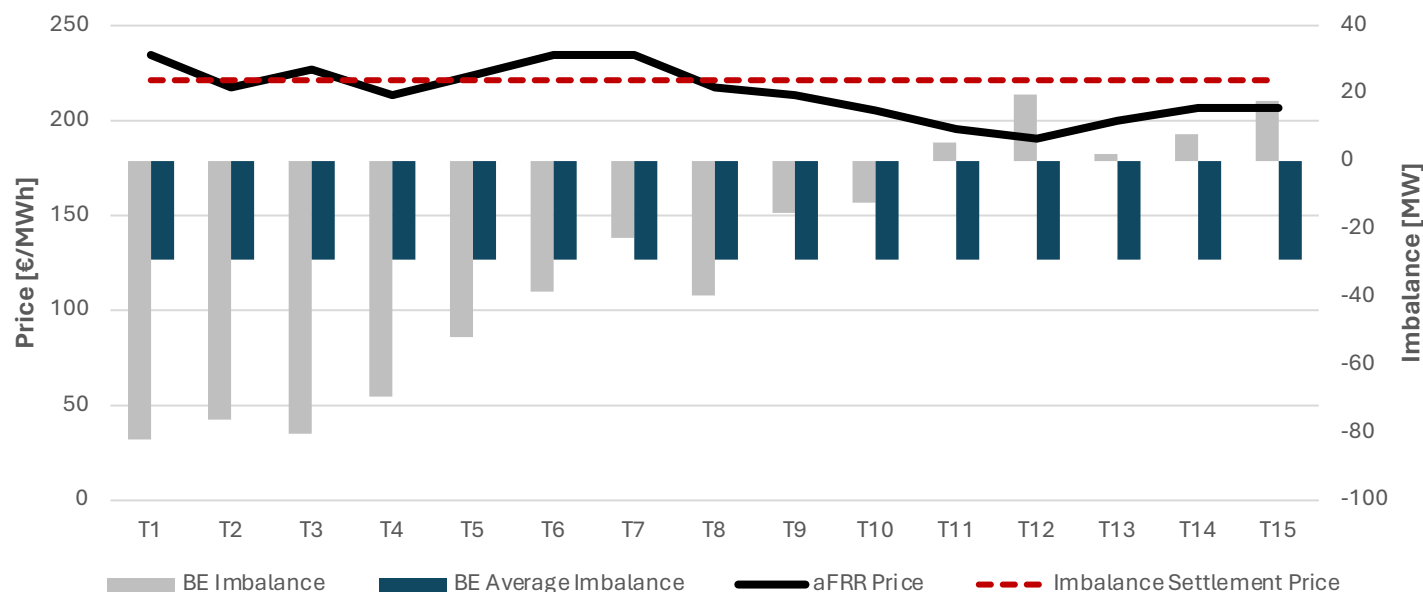
☐ Limited available interconnecting capacity

- Interconnecting line becomes quickly congested
- Long position cannot be exported → activation of downward resources
- Low weighted average prices

Participants avoid self-scheduling

# Simulation Results

## Imbalance Settlement Period Granularity Effect



Avg. Imbalance: **-28.99 MW**  
 VoAA Up: **221.24 €/MWh**  
 VoAA Down: **221.24 €/MWh**  
 Variable Cost: **212 €/MWh**

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CREG Price: **221.24 €/MWh**  
 ELIA Price: **221.24 €/MWh**

- ❑ Imbalance direction may change towards the end of the imbalance settlement period
- ❑ Imbalance settlement price defined mainly by the price that is derived due to the dominant imbalance direction
- Arbitrage opportunities from flexible resources that may adapt their output



Introduction

Model Description

Analysis Settings

Simulation Results

**Conclusions**

Policy Conclusions

Future Directions



# Conclusions

## Policy Conclusions

□ Results from extensive simulations demonstrate that:

- The CREG-based formula may activate the BRPs opposite to the imbalance conditions of the Belgian system  
→ support the needs of the overall European system
- Belgian system security is not endangered → available downward capacity in the Belgian zone.
- The ELIA-based formula leads market participants to respond mainly to local imbalances.
- Additional simulations → for cross-zonal transmission capacity of 100 MW and an imbalance interval of [-50 MW, +50 MW ] on which the CREG formula is applied we can achieve similar results to the formula proposed by ELIA. → Combination of the advantages and basic characteristics of both imbalance settlement pricing formulas

# Conclusions

## Future Directions

- ❖ Perform analysis on a multi-area test bed
- ❖ Examine additional state-action pairs
- ❖ Apply parallel programming to speed-up the training and analysis
- ❖ Examine other currently existing European imbalance settlement price schemes



# Thank you very much!

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