



Stochastic Energy Systems with Mechanism Choices

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01

Introductions



Energy System Fundamentals



Background


Energy systems of consumer communities

An energy system used for supplying energy to a group of energy consumer entities, all energy consumptions are in a First-In-First-Out order.

Examples of Such energy systems

e.g., power plant of an apartment, where each room requires energy supply equally.

Two parts of energy systems' cost origins

1. Energy consumption spending (**More energy consumption, more cost**)
 2. Energy supply congestion spending (**More waiting supplies, most cost**)
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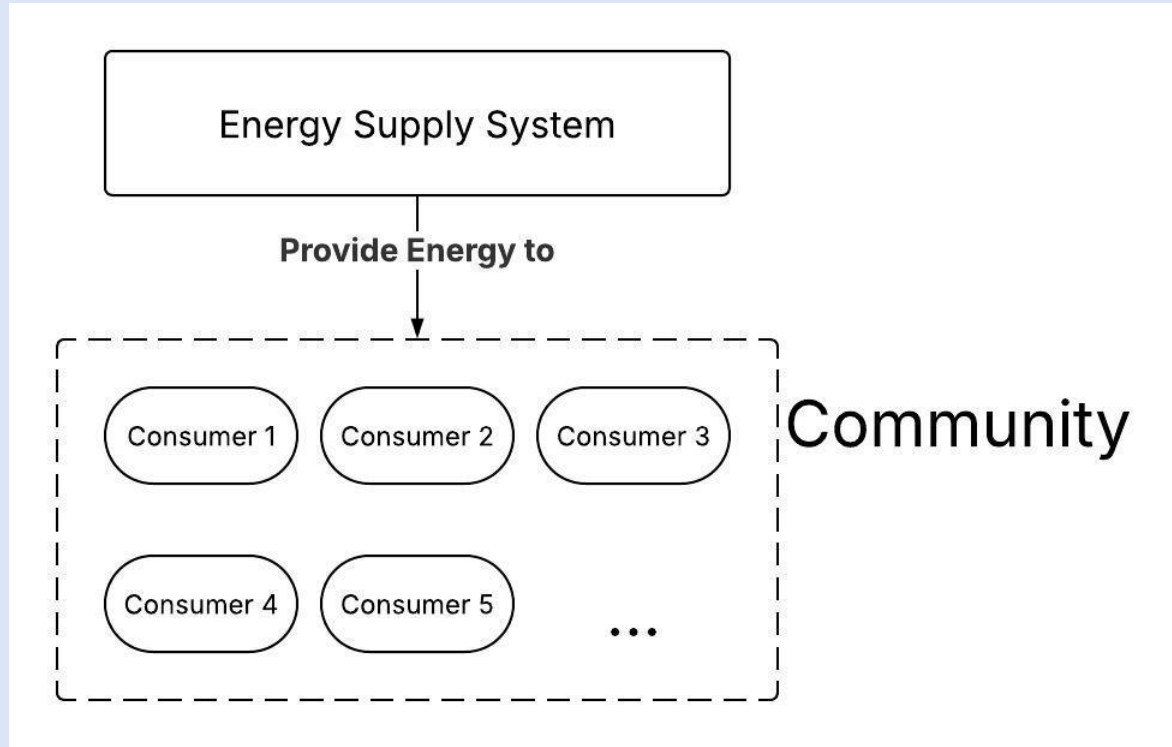



Fig 1 – Sample of Energy Systems



Engineer's Paralleled Goals

Goal 1: To minimize energy consumption of the whole system

Goal 2: To minimize waiting in this system's energy serving





02

Problem Environment



Energy System Concrete Problems' Settings

On usual cases:

1. The energy system is a continuous time Markov chain (CTMC), where job arrivals are under Poisson distributions, whose parameter is a constant.
2. The energy system will not service the arrived jobs until the waiting job amount reaches k_1 , where $k_1 \geq 1$. Then the system will turn on with γ (the turn-on rate).
3. When the waiting job amount $< k_2$, the system uses the normal process rate (μ), otherwise when it $\geq k_2$, the system will use the enhanced process rate ($\bar{\mu}$).

On usual cases:

- 4. In order to make the whole system under stable and expected performance, $\lambda < \mu < \bar{\mu}$.
- 5. Alpha is the idle rate of the energy system, available on 1 of 2 conditions. (**Mentioned on slide 11**)
- 6. Remember gamma (the turning-on rate) and alpha (the idle rate) also exists, both are greater than 0.
- 7. will_turn_off represents the ability to turn off. If it is true, then $\alpha > 0$, or it is 0. -> (the system cannot turn off)

On usual cases:

8. $1 \leq k_1 \leq k_2$, as the enhanced service rate only occurs after normal service rate (represents overload).

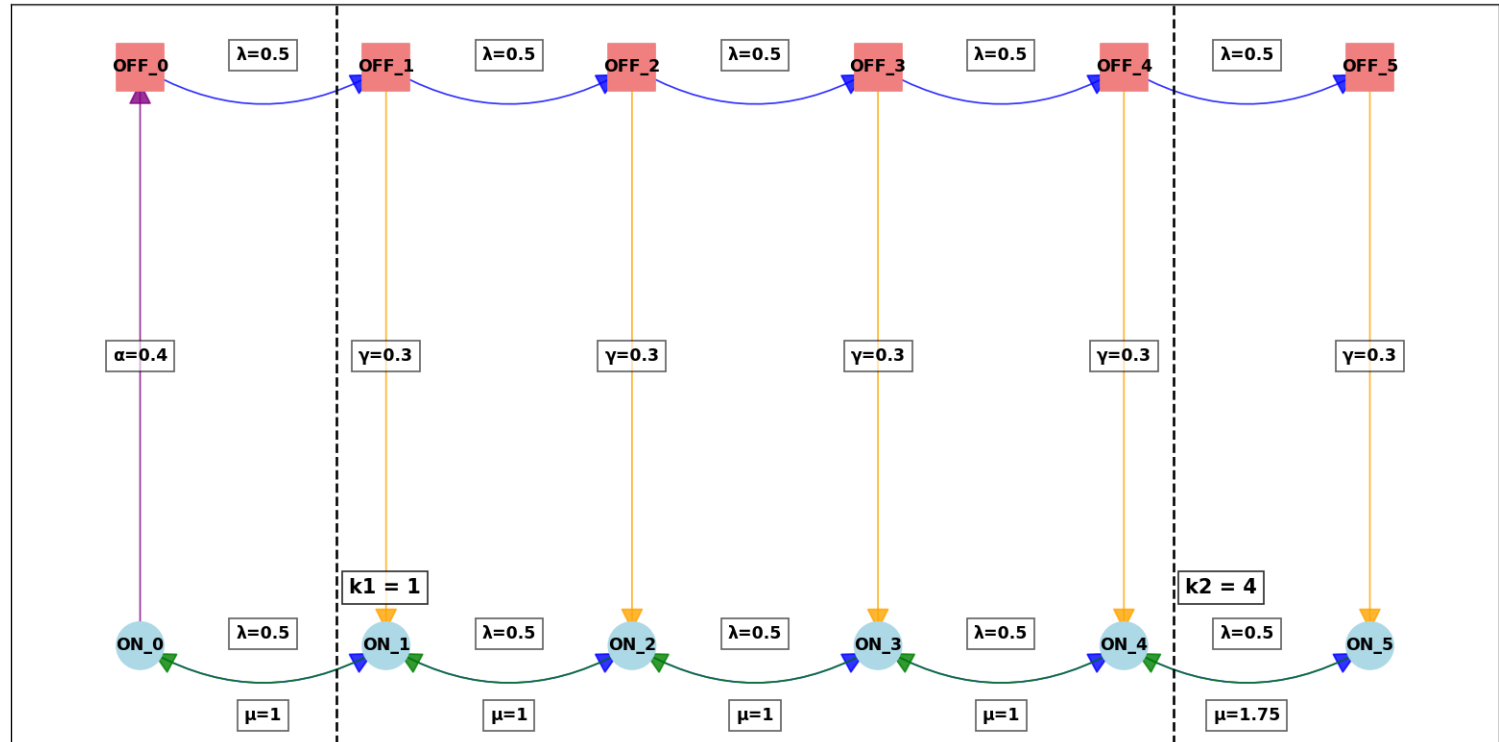
- Note that both k_1 and k_2 can reach infinity.

9. **Important:** parameters k_1 , k_2 , and will_turn_off are variables (can be adjusted by system controls), while others are all constant (cannot be changed).

• will_turn_off determines distinctions

1. When will_turn_off = False, the energy system will never go back to OFF states (to become idle), so we can ignore all states without service processing. This results in a **variant M/M/1 queue (Two service rates)**.
2. When will_turn_off = True, the energy system will go back to idle when there is no waiting job. This has more complicated structures, and it **does not belong to any special CTMC cases** (e.g., M/M/1).

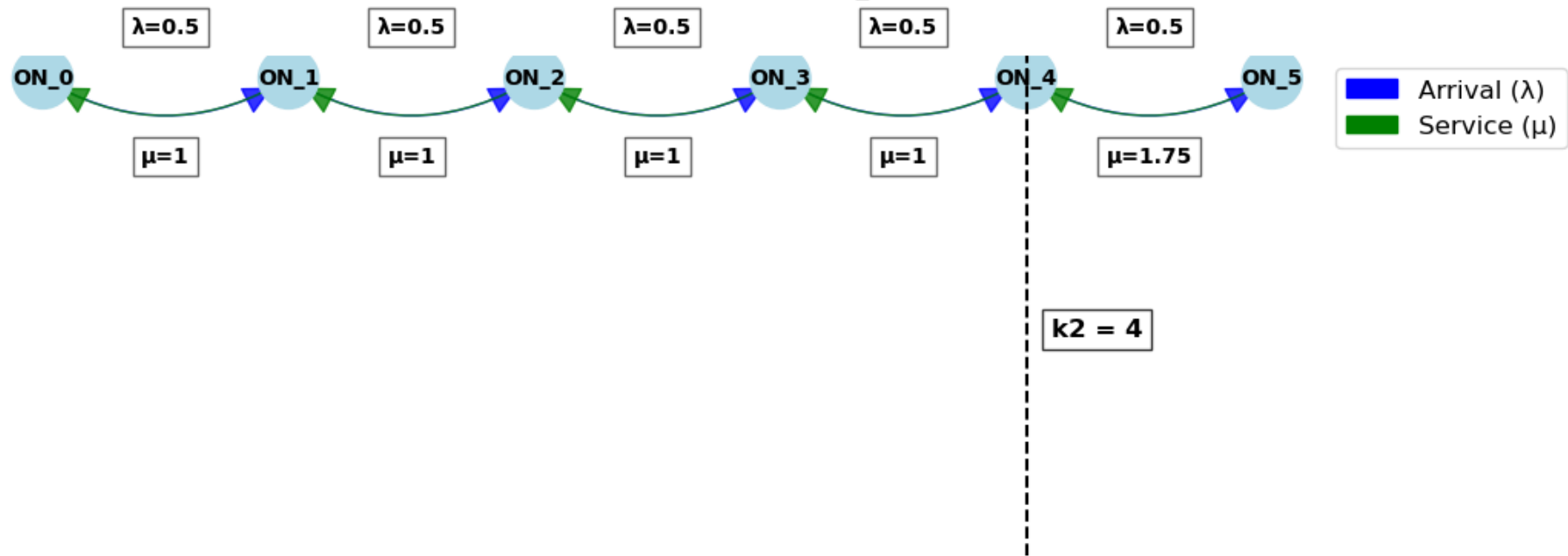
Transition Diagram with Separated Lambda and Mu Edges for max_jobs = 5



- Arrival (λ)
- Service (μ)
- Turn On (γ)
- Idle (α)



Transition Diagram (ON States Only, max_jobs = 5)



Power Consumption and Total Cost

1. Power consumption = Sum of (each state's steady-state probability * power consumption coefficient)
2. Power consumption coefficient:
0 for OFF states when the system is not turning on, 1 for OFF turning-on states. σ for ON0, 1 for ON states with the regular service rate, and $(\mu_{\text{bar}})^2$ for ON states with the enhanced service rate.
3. The idle power consumption coefficient σ is greater than 0 and less than 1.

Total Cost

1. Expected Number of Jobs (Mean) = Sum of (each state's steady-state probability * the waiting jobs)
- 2. ON0 means 0 waiting jobs, ON1 means 1, and so on...

- 1. **Total_Cost** = Power Consumption * beta + Expected Number of Jobs.
- 2. beta is a constant weight parameter, greater than 0.



03

Objectives

Problems' solution objectives



Solved Question 1

Our goal is to minimize the Total Cost of the whole energy system. Previously there are two choices:

- 1. Maximize k_2 value, which aims to minimize the energy consumption caused by enhanced service rate usage.
- 2. Setting `will_turn_off = True`.
- 3. Setting `will_turn_off = False`.
- **Question:** Can 1 and 2, or 1 and 3, co-exist in the optimal parameter group?

Solved Question 2

Question 2 is a **variant** problem of Question 1:

1. Parameter k_2 is a non-polarized value, which means it is neither maximized nor minimized.
2. Setting `will_turn_off = True`.
3. Setting `will_turn_off = False`.

Question: Can 1 and 2, or 1 and 3, co-exist in the optimal parameter group?



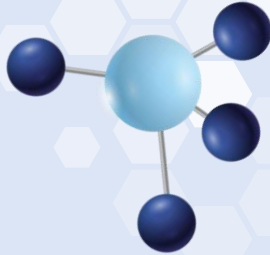
04

Methodologies



Methods used to solve these problems

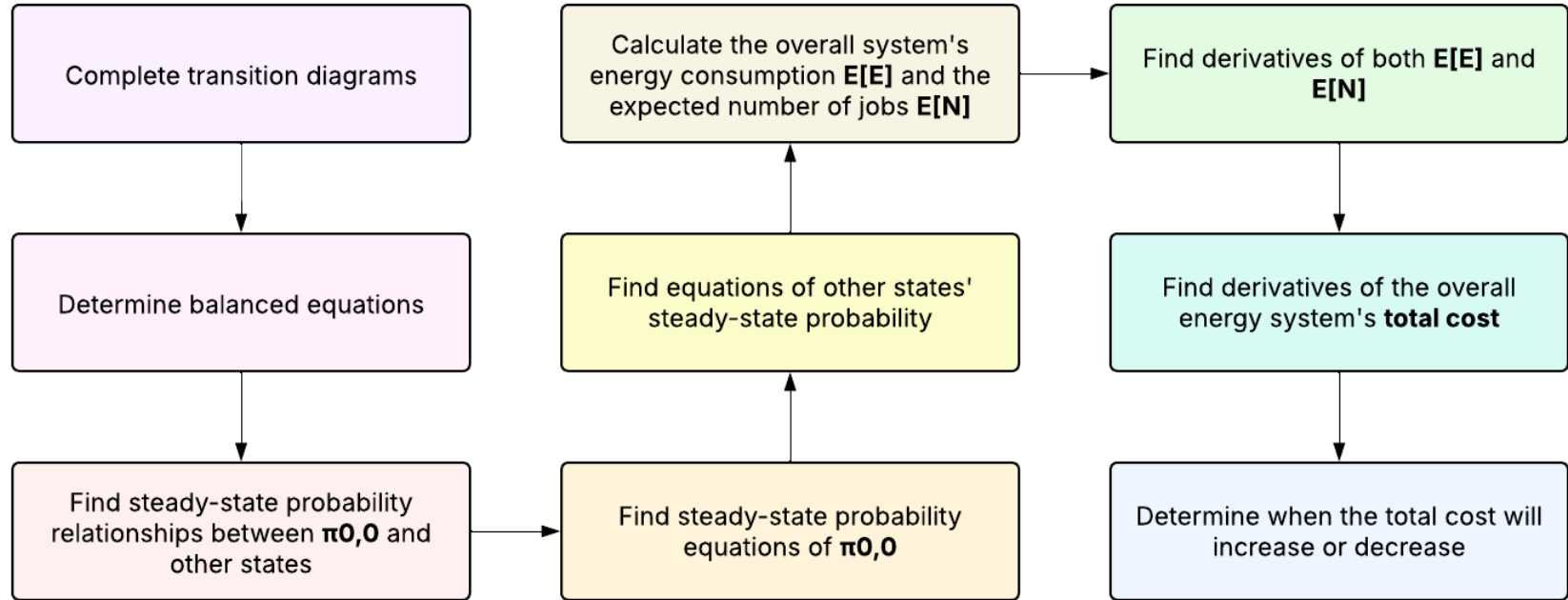
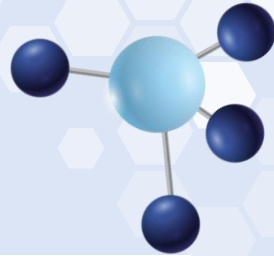
Key – Solving CTMC



1. Both cases are under CTMC (continuous-time Markov chain), so the method of solving CTMC balanced equations will be efficient.
2. After getting equations of all states' steady-state probability, we can determine the overall system's equation of energy consumption and # of Jobs, which points out the equation of total cost.
3. By discovering the k2 derivative of such system's total cost, we can find local max/min positions (which k2 is).



Mandatory Processes



Step 1 – Complete Transition Diagrams

1. Listed all states on both conditions.

$\pi 0$, x represent OFF states.

$\pi 1$, x represent ON states.

2. Determine all states' transmission rates and targets
(λ , μ , $\bar{\mu}$, γ , α)

Step 2 & 3 – Solving balanced equations

1. For each state, find all IN and OUT rates, and summarize them to an equation between two nearby states' steady-state probability.
2. Solve these equations, to get relationships between different states' steady-state probability.
3. By substitutions, find all other states' steady-state probability equations with $\pi_0, 0$.

Step 4 – Determine the formula of $\pi_{0,0}$

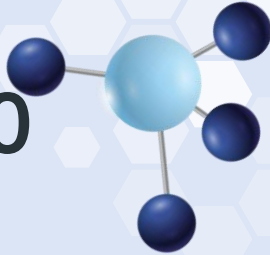
Now all other states can be expressed by $(...) * \pi_{0,0}$.

By adding them together, we have:

$$(\text{Net Term}) * \pi_{0,0} = 1.$$

So $\pi_{0,0}$ will equal $1/(\text{Net Term})$.

Reminder: Whatever the system's parameters $\pi_{0,0}$ change, will always be positive.





Step 5 – Determine the formula of states

From Step 3, we already have the relationship equation of different formulas between each other state and $\pi_0, 0$.
e.g., $(...) * \pi_0, 2 = (...) * \pi_0, 0$.

By substituting $\pi_0, 0$ and transforming the formula, we can get equations of all other states. E.g., $\pi_0, 2 = (...)/(...)$
↳ here $\pi_0, 0$, are already substituted.



Step 6 – Determine $E[E]$ and $E[N]$

To calculate the expected jobs $E[N]$:

$E[N] = \text{sum of \{each state's steady-state probability * its corresponding number of jobs\}}$

e.g., $\pi_{1, 0}$ has the steady-state probability of 0.3, and its expected number of jobs is 0, so therefore its expected number of jobs is $0.3 * 0 = 0$.

Step 7 – Find derivatives

Some interesting things exist on both $E[E]$ and $E[N]$:

1. They have the same denominator, because both of them are calculated from $\pi_0, 0$
2. On both numerator side, a lot of terms look similar, since there are a lot of common terms inside.

Step 7 – Find k2 derivatives

First use the quotient rule for both body, and then simplify both sides' numerator terms.

(Not need to simplify the denominator terms. Since by the quotient rule, both **E[E]** and **E[N]** derivative's denominator is **D²**, where **D** is the denominator of **E[E]**)

Step 8 – Find total cost and its derivatives

From slide 15, we already know **total cost** = $E[E] * \text{beta} + E[N]$, where beta is the weight coefficient (**always** > 0).

Substitute the total cost functions with actual equations of **$E[E]$** and **$E[N]$** . On the numerator side, a lot of terms can be combined or cancelled.

Step 9 – Find total cost's k^2 derivatives

From slide 15, we already know **total cost** = $E[E] \cdot \text{beta} + E[N]$, where beta is the weight coefficient (**always** > 0).

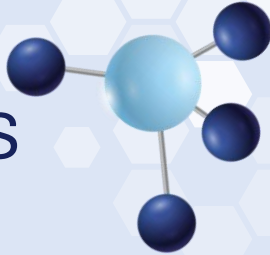
Considering beta as an abstract constant, then we have the total cost's k^2 derivative here (using **constant rule**):

$$\frac{d}{dk^2}(\text{total cost}) = \frac{d}{dk^2}(E[E]) \cdot \beta + \frac{d}{dk^2}(E[N])$$

Step 9 – Find total cost's k_2 derivatives

With such k_2 derivative, now we are able to find when it will be **positive** (total cost is increasing), and when it will be **negative** (total cost is decreasing).

We can also determine the energy system's optimal k_2 values on both conditions.



Formal Proof (on both problems' solving)



Both conditions have considered three cases independently:

1. β is a very small value, meaning only $E[N]$ is important for determining the total cost's values and tendencies.
2. β is a very large value, meaning only $E[E]$ is important for determining the total cost's values and tendencies.
3. β is a balanced value, meaning both $E[E]$ and $E[N]$ are important for determining the total cost's values and tendencies.





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Analysis



Experimental results

Case A – will_turn_off = False

$$E[E] = \frac{\sigma \cdot \left(1 + \frac{\lambda}{\mu}\right) \left(1 - \frac{\lambda}{\mu}\right) + \left(\frac{\lambda}{\mu}\right)^{k_2} \cdot \left(\left(1 - \frac{\lambda}{\mu}\right) \cdot \bar{\mu}^2 - \left(1 - \frac{\lambda}{\bar{\mu}}\right)\right)}{\left(1 - \left(\frac{\lambda}{\mu}\right)^{k_2}\right) \left(1 - \frac{\lambda}{\bar{\mu}}\right) + \left(\frac{\lambda}{\mu}\right)^{k_2} \cdot \left(1 - \frac{\lambda}{\mu}\right)}$$

$$E[N] = \frac{PartA + PartB}{PartD}$$

$$PartA = k_2 \cdot \left(\frac{\lambda}{\mu}\right)^{k_2} \cdot \left(\frac{1}{(1-\frac{\lambda}{\bar{\mu}})} - \frac{1}{(1-\frac{\lambda}{\mu})}\right) \cdot \left(1 - \frac{\lambda}{\bar{\mu}}\right) \cdot \left(1 - \frac{\lambda}{\mu}\right)$$

$$PartB = \left(\frac{\frac{\lambda}{\bar{\mu}}}{(1-\frac{\lambda}{\bar{\mu}})^2} \cdot \left(1 - \left(\frac{\lambda}{\mu}\right)^{k_2}\right) + \frac{\frac{\lambda}{\bar{\mu}}}{(1-\frac{\lambda}{\bar{\mu}})^2} \cdot \left(\frac{\lambda}{\mu}\right)^{k_2}\right) \cdot \left(1 - \frac{\lambda}{\bar{\mu}}\right) \cdot \left(1 - \frac{\lambda}{\mu}\right)$$

$$PartD = \left(1 - \left(\frac{\lambda}{\mu}\right)^{k_2}\right) \left(1 - \frac{\lambda}{\bar{\mu}}\right) + \left(\frac{\lambda}{\mu}\right)^{k_2} \cdot \left(1 - \frac{\lambda}{\mu}\right)$$

Case A – will_turn_off = False

$$\frac{d}{dk_2} (TotalCost) = \frac{Numerator}{NewDenominator^2}$$

$$Numerator = \left(\frac{\lambda}{\mu}\right)^{k_2} \cdot \ln\left(\frac{\lambda}{\mu}\right) \cdot \left(\frac{\lambda}{\bar{\mu}} - \frac{\lambda}{\mu}\right) \cdot \left(\left(k_2 - \sigma \cdot \left(1 + \frac{\lambda}{\mu}\right) \cdot \left(1 - \frac{\lambda}{\mu}\right) \right) \cdot \beta + \left(1 - \frac{\lambda}{\bar{\mu}}\right) - \left(1 - \frac{\lambda}{\mu}\right) \cdot \left(\bar{\mu}^2 + \frac{\frac{\lambda}{\bar{\mu}}}{1 - \frac{\lambda}{\bar{\mu}}} \right) \right)$$

$$NewDenominator = \left(1 - \frac{\lambda}{\mu} + \left(\frac{\lambda}{\mu}\right)^{k_2} \cdot \left(\frac{\lambda}{\bar{\mu}} - \frac{\lambda}{\mu}\right) \right)^2$$

Case A – will_turn_off = False

Problem 1 solution: Yes, 1 and 3 can co-exist in the optimal parameter group.

Observations:

1. Maximum k_2 minimizes $E[E]$ but maximizes $E[N]$, and will_turn_off = False increase $E[E]$
2. However, a small k_1 value can lower $E[N]$ and offset the effects of the maximum k_2 .

Case A – will_turn_off = False

Evidence:

$\lambda = 0.2$, $\mu = 1$, $\mu_{\text{bar}} = 10000$, $\gamma = 0.5$,
 $\alpha = 0.5$, $\sigma = 0.5$, $\beta = 1$.

```
Best Parameters: k1=1, k2=500, will_turn_off=False  
Minimum Total Cost: 0.85000000000086299  
Elapsed Time: 154.80 seconds
```

Case A – will_turn_off = False

Problem 2 solution: Yes, 1 and 3 can co-exist in the optimal parameter group.

Observations:

1. On slide 32 case 1 & 2, the k_2 derivative of total cost is continuously positive or negative, which disallow non-polarized k_2 value to exist in the optimal parameter group.
2. However, on case 3, the k_2 derivative of total cost is first negative then positive, allowing non-polarized k_2 vals.

Case A – will_turn_off = False

Evidence:

$\lambda = 0.2$, $\mu = 1$, $\mu_{\text{bar}} = 100$, $\gamma = 0.5$,
 $\alpha = 0.5$, $\sigma = 0.5$, $\beta = 1$.

```
Best Parameters: k1=1, k2=80, will_turn_off=False  
Minimum Total Cost: 0.85000000000026108  
Elapsed Time: 149.54 seconds
```

Case A – Methods to find optimal k2 values

There are several terms in the k2 derivative of total cost, but only one term possible to change sign:

$$\left(\left(k_2 - \sigma \cdot \left(1 + \frac{\lambda}{\mu} \right) \cdot \left(1 - \frac{\lambda}{\mu} \right) \right) \cdot \beta + \left(1 - \frac{\lambda}{\bar{\mu}} \right) - \left(1 - \frac{\lambda}{\mu} \right) \cdot \left(\bar{\mu}^2 + \frac{\frac{\lambda}{\bar{\mu}}}{1 - \frac{\lambda}{\bar{\mu}}} \right) \right)$$

So, we need to solve:

$$\left(\left(k_2 - \sigma \cdot \left(1 + \frac{\lambda}{\mu} \right) \cdot \left(1 - \frac{\lambda}{\mu} \right) \right) \cdot \beta + \left(1 - \frac{\lambda}{\bar{\mu}} \right) - \left(1 - \frac{\lambda}{\mu} \right) \cdot \left(\bar{\mu}^2 + \frac{\frac{\lambda}{\bar{\mu}}}{1 - \frac{\lambda}{\bar{\mu}}} \right) \right) < 0$$

Case A – Methods to find optimal k_2 values

Example

Given $\lambda = 0.2$, $\mu = 1$, $\gamma = 0.8$, $\alpha = 1.5$, $\bar{\mu} = 2$, $\beta = 1$, $\sigma = 0.5$, $k_1 = 1$, and k_2 from 1 to ∞ . (k_2 is the only variable.)

By substituting all these into this formula:

$$\left(\left(k_2 - \sigma \cdot \left(1 + \frac{\lambda}{\mu} \right) \cdot \left(1 - \frac{\lambda}{\mu} \right) \right) \cdot \beta + \left(1 - \frac{\lambda}{\bar{\mu}} \right) - \left(1 - \frac{\lambda}{\mu} \right) \cdot \left(\bar{\mu}^2 + \frac{\frac{\lambda}{\bar{\mu}}}{1 - \frac{\lambda}{\bar{\mu}}} \right) \right) < 0$$

We obtained $k_2 - 2.8689 < 0$, which is $k_2 < 2.8689$. Since k_2 can only be an integer, the greatest k_2 value fulfils this is $k_2 = 2$, which is regionally optimal.

Case B – will_turn_off = True

$$E[E] = \frac{\text{Numerator1}}{\text{Denominator}}$$

$$\begin{aligned}\text{Numerator1} = & \frac{\lambda}{\gamma} - \sigma \times \frac{\lambda}{\alpha} + \left(\left(\frac{\lambda}{\alpha} - \frac{\rho}{1-\rho} \right) \times \frac{1-\rho^{k_2}}{1-\rho} - \frac{\lambda}{\alpha} + k_2 \times \frac{\rho}{1-\rho} \right) \\ & + \frac{\lambda}{\bar{\mu} - \lambda} \times \left(\frac{\lambda}{\alpha} \times \rho^{k_2} + \frac{\rho^{k_2-k_1+1}}{1-\rho} \times (1-\rho^{k_1}) \right) \\ & + \frac{\lambda}{\bar{\mu} - \lambda} \times \left(\frac{\frac{\lambda}{\lambda+\gamma} \times \rho^{k_2-k_1}}{1 - \frac{\lambda}{\lambda+\gamma} \times \rho} \times \left(1 - \frac{\lambda}{\lambda+\gamma} \times \rho^{k_2-k_1} \right) \right) \\ & + \frac{\lambda + \gamma}{\lambda + \gamma - 1} * \left(\frac{\lambda}{\lambda + \gamma} \right)^{k_2-k_1+1} * \bar{\mu}^2\end{aligned}$$

$$\begin{aligned}\text{Denominator} = & k_1 + \frac{\lambda}{\gamma} + \left(k_2 - \frac{1-\rho^{k_2}}{1-\rho} \right) \times \frac{\lambda}{\bar{\mu} - \lambda} \\ & + \frac{\lambda}{\bar{\mu} - \lambda} \times \left(\frac{\lambda}{\alpha} \times \rho^{k_2} + \frac{\frac{\lambda}{\lambda+\gamma} \times \rho^{k_2-k_1}}{1 - \frac{\lambda}{\lambda+\gamma} \times \rho} \times \left(1 - \frac{\lambda}{\lambda+\gamma} \times \rho^{k_2-k_1} \right) \right) \\ & + \frac{\lambda}{\bar{\mu} - \lambda} \times \frac{\rho^{k_2-k_1+1} \times (1-\rho^{k_1})}{1-\rho}\end{aligned}$$

Case B – will_turn_off = True

$$E[N] = \frac{\text{Numerator2}}{\text{Denominator}}$$
$$\begin{aligned} \text{Numerator2} = & \frac{k_1 * (k_1 - 1)}{2} + \left(\frac{\lambda * (\lambda + \gamma)}{\gamma} + (k_1 - 1) * \frac{\lambda}{\gamma} \right) \\ & + \frac{\rho * (1 - \rho^{k_1} * (1 + k_1 (1 - \rho)))}{(1 - \rho)^2} + \frac{\rho}{1 - \rho} * \frac{k_1 * (k_1 + 1)}{2} \\ & + \frac{(k_2 - k_1) (k_2 + k_1 + 1)}{2} * \left(\frac{\lambda}{\alpha} * \rho^{k_2} + \rho^{k_2 - k_1 + 1} * \frac{1 - \rho^{k_1}}{1 - \rho} \right) \\ & + \frac{(k_2 - k_1) (k_2 + k_1 + 1)}{2} * \left(\frac{\lambda}{\lambda + \gamma} * \rho^{k_2 - k_1} * \frac{1 - \frac{\lambda}{\lambda + \gamma} * \rho^{k_2 - k_1}}{1 - \frac{\lambda}{\lambda + \gamma} * \rho} \right) \\ & + \frac{\lambda}{\bar{\mu} - \lambda} * \left(k_2 + \frac{1}{\bar{\mu} - \lambda} \right) * \left(\frac{\lambda}{\alpha} * \rho^{k_2} + \frac{\rho^{k_2 - k_1 + 1} * (1 - \rho^{k_1})}{1 - \rho} \right) \\ & + \frac{\lambda}{\bar{\mu} - \lambda} * \left(k_2 + \frac{1}{\bar{\mu} - \lambda} \right) * \left(\frac{\frac{\lambda}{\lambda + \gamma} * \rho^{k_2 - k_1}}{1 - \frac{\lambda}{\lambda + \gamma} * \rho} * \left(1 - \frac{\lambda}{\lambda + \gamma} * \rho^{k_2 - k_1} \right) \right) \end{aligned}$$

The denominator of E[N] is the same as E[E].

Case B – will_turn_off = True

Formulas of the k2 derivative of total cost is available at:

<https://www.overleaf.com/read/rpbgzbfpsktm#65aff2>

Case B – will_turn_off = True

Problem 1 solution: Yes, 1 and 2 can co-exist in the optimal parameter group.

Observations:

1. Selecting higher idle rates and turn-on rates trigger preferences on will_turn_off = True
2. Parameter μ and $\bar{\mu}$ are significant on determining the total cost actual values, but they are trivial in making preferences on will_turn_off options.

Case B – will_turn_off = True

Evidence:

$\lambda = 0.5$, $\mu = 1$, $\mu_{\text{bar}} = 17.5$, $\gamma = 1.5$,
 $\alpha = 1.5$, $\sigma = 0.5$, $\beta = 500$.

Best Parameters: $k_1=1$, $k_2=500$, will_turn_off=True
Minimum Total Cost: 326.26666666667114
Elapsed Time: 152.24 seconds

Case B – will_turn_off = True

Problem 2 solution: Yes, 1 and 2 can co-exist in the optimal parameter group.

Observations:

1. On slide 32 all three cases, there exists cases allowing the non-polarized k_2 value to exist as optimal.
2. To have such solutions, parameter λ should have a small value (**$\lambda \leq 10\%$ of μ**)

Case B – will_turn_off = True

Case 1 Evidence:

Counter-example: $\lambda = 0.05$, $\mu = 1$, $\bar{\mu} = 1.1$, $\gamma = 0.8$, $\alpha = 0.1$, $\sigma = 0.5$, and $\beta = 500$.

Actual optimal parameter: Best Parameters: $k_1 = 1$, $k_2 = 493$, will_turn_off=True
Minimum Total Cost: 120.09513157895385.

Case B – will_turn_off = True

Case 2 Evidence:

Counter-example: $\lambda = 0.02$, $\mu = 1$, $\bar{\mu} = 1000000$, $\gamma = 3$, $\alpha = 4$, $\sigma = 0.5$, and $\beta = 0.08$.

Actual optimal parameter: Best Parameters: $k_1 = 1$, $k_2 = 10$, will_turn_off=True
Minimum Total Cost: 0.02935225936.

Case B – will_turn_off = True

Case 3 Evidence:

Counter-example: $\lambda = 0.2$, $\mu = 1$, $\bar{\mu} = 10$, $\gamma = 3$, $\alpha = 4$, $\sigma = 0.5$, and $\beta = 1$.

Actual optimal parameter: Best Parameters: $k_1 = 1$, $k_2 = 8$, will_turn_off=True
Minimum Total Cost: 0.5793526693.



06

Conclusions



Final Conclusions

Conclusions

1. Non-polarized k_2 values exist on all conditions' most cases. This means **selecting polarized k_2 values may not always contributes to optimal total cost.**
2. Interesting discoveries: although these non-polarized k_2 values exist, the k_2 values greater than them don't have a much larger total cost (**their differences are trivial**)

A photograph of two scientists, an older woman with grey hair and a younger woman with dark hair, both wearing blue surgical masks and blue gloves. They are in a laboratory setting, looking at a small object held by the older woman. The background is slightly blurred, showing lab equipment. There are decorative elements: a hexagonal pattern on the left and a molecular structure on the right.

**End of
Presentation**