

# Modelling

## *Linear Optimization for Improved Decision-Making*

A Modeling Agency Business Context

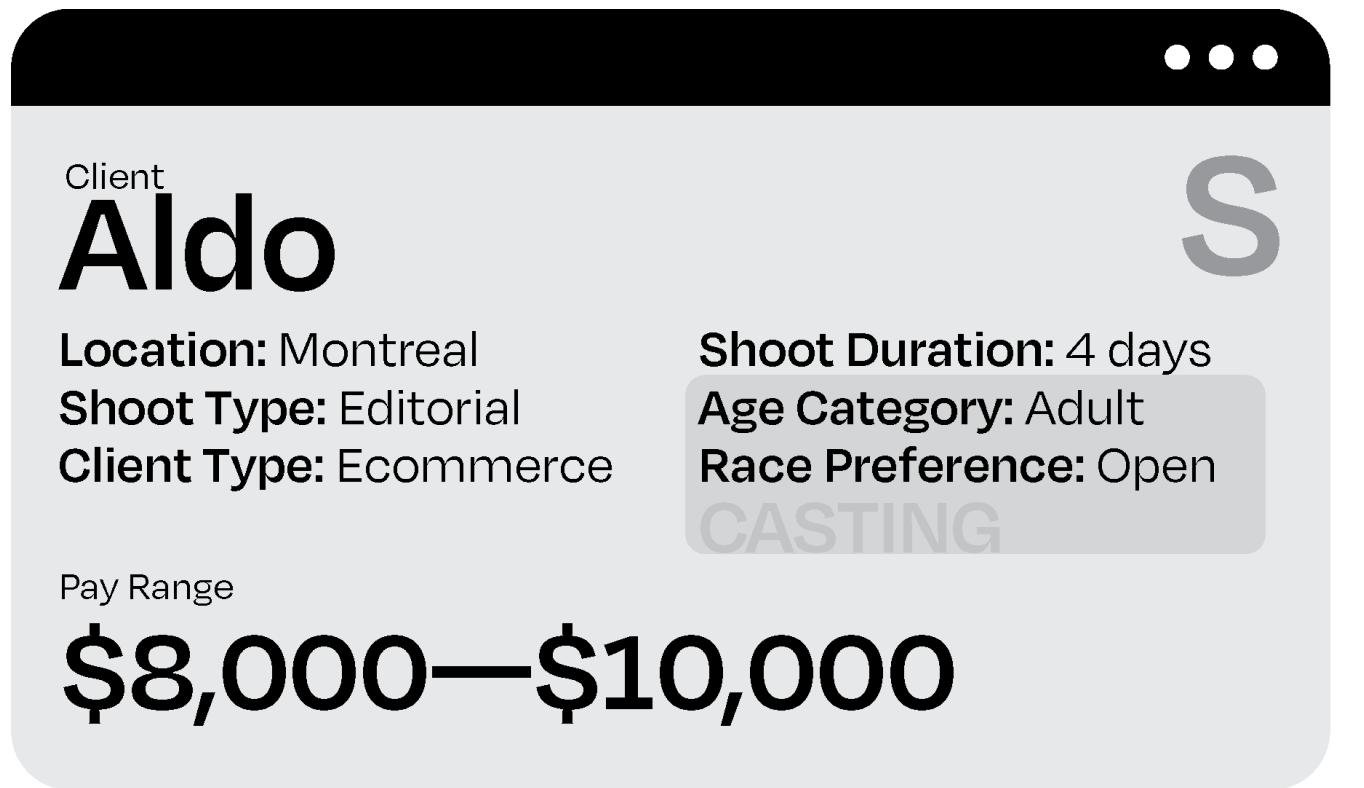
### Project Contributors

Rohan Kumar | Meriem Mehri | Avi Malhotra | Moiz Zahid Shaikh | Yvan Kammel

# AGENDA

- 1 Executive Summary**
- 2 Problem Context**
- 3 Mathematical Formulation**
  - a Variables**
  - b Objective Function**
  - c Constraints**
- 4 Results Interpretation**
- 5 Possible Extensions & Challenges**
- 6 Appendix**

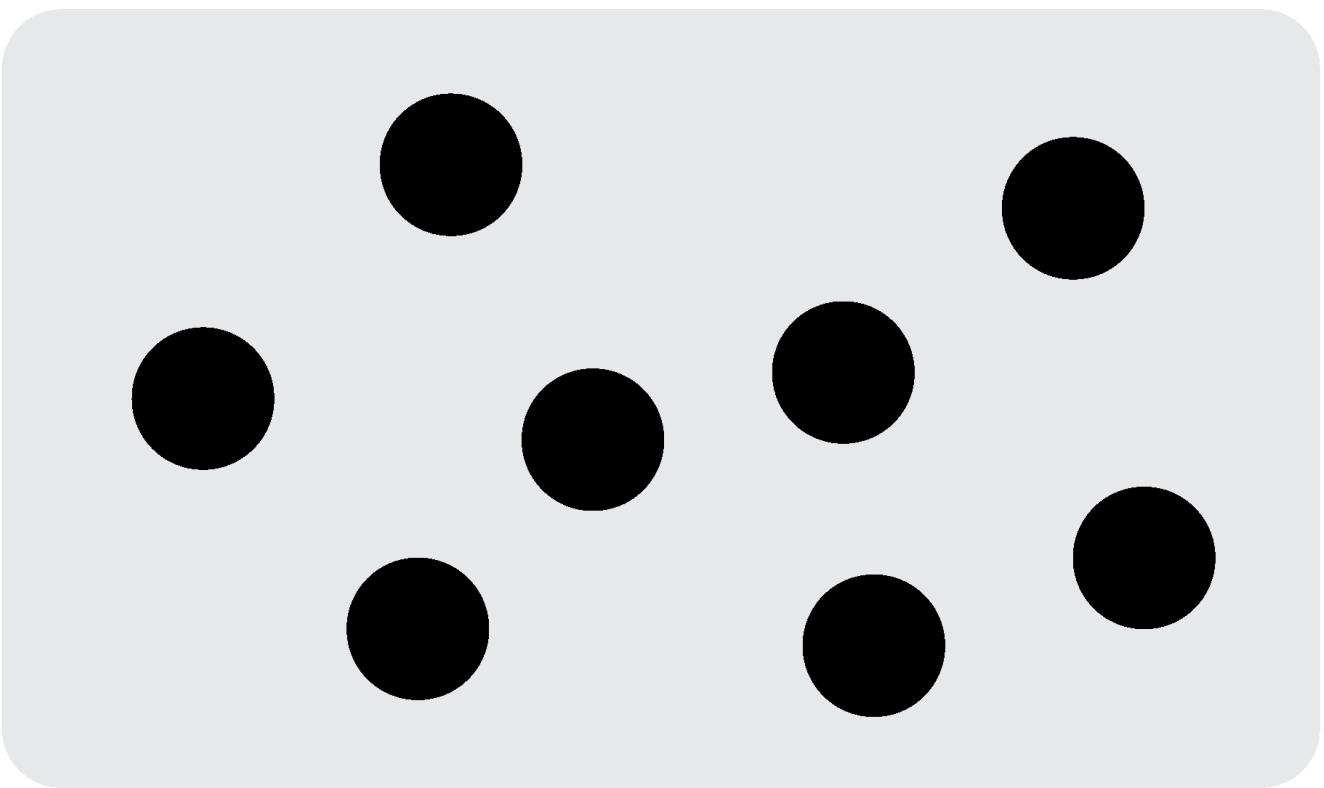
# BACKGROUND CONTEXT AND SIGNIFICANCE



**CLIENT**

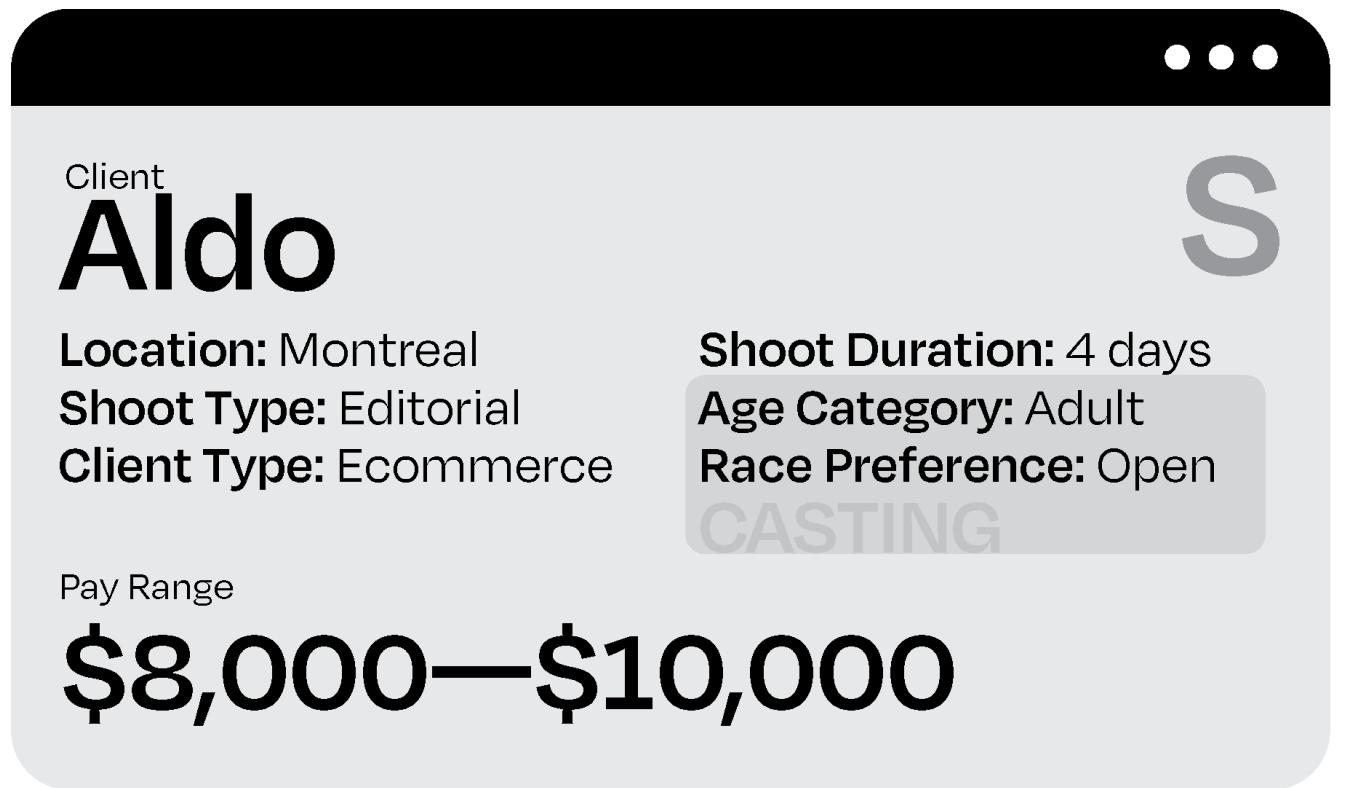


**AGENT**



**MODELS**

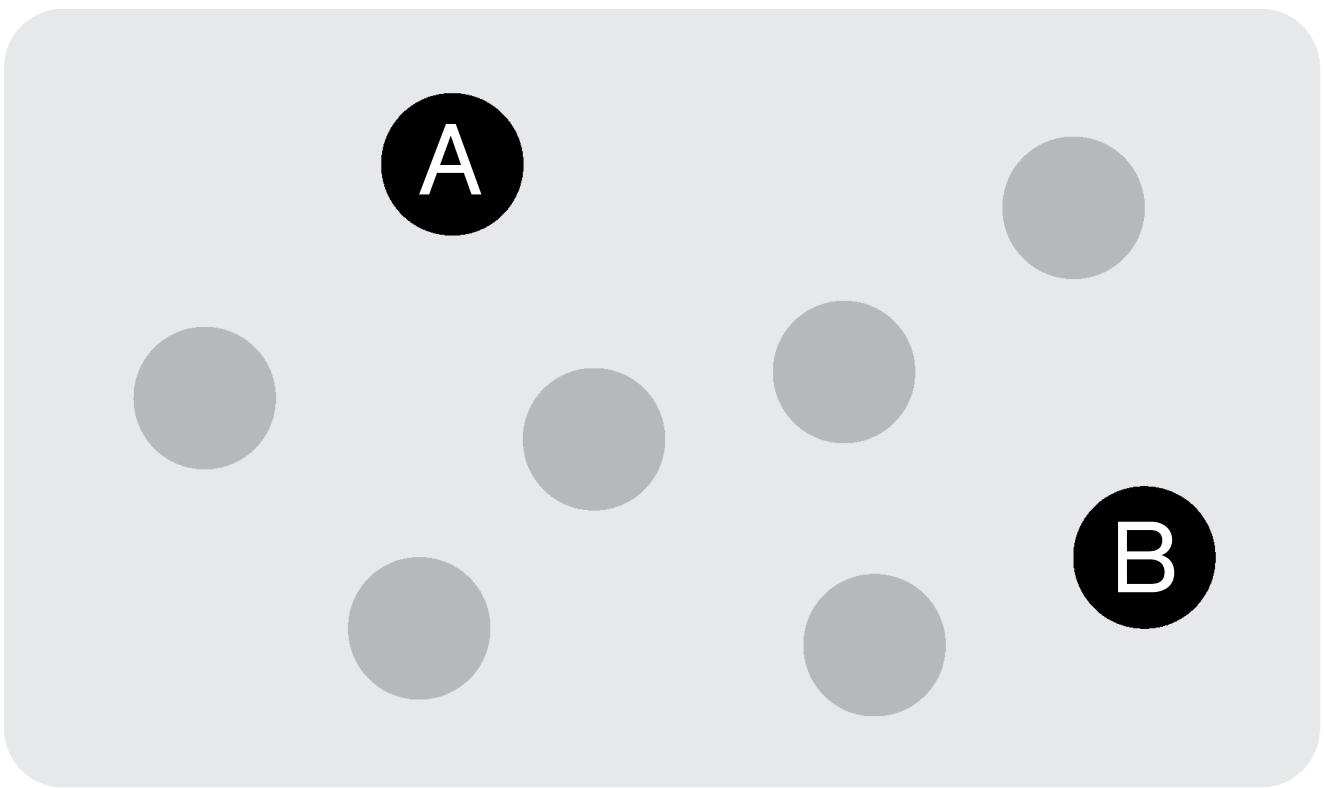
# BACKGROUND CONTEXT AND SIGNIFICANCE



**CLIENT**

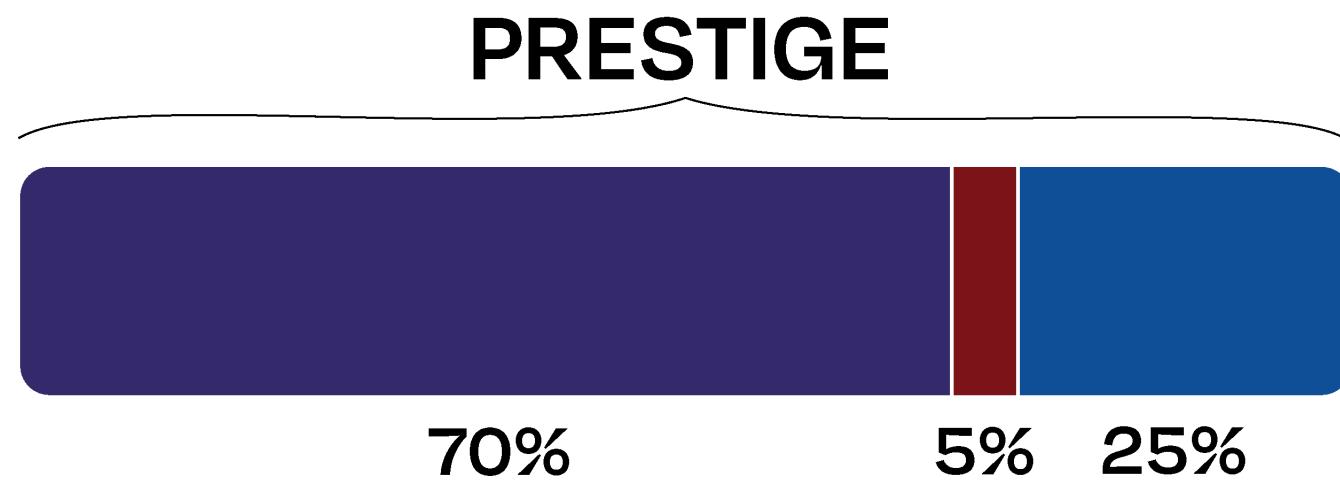


**AGENT**



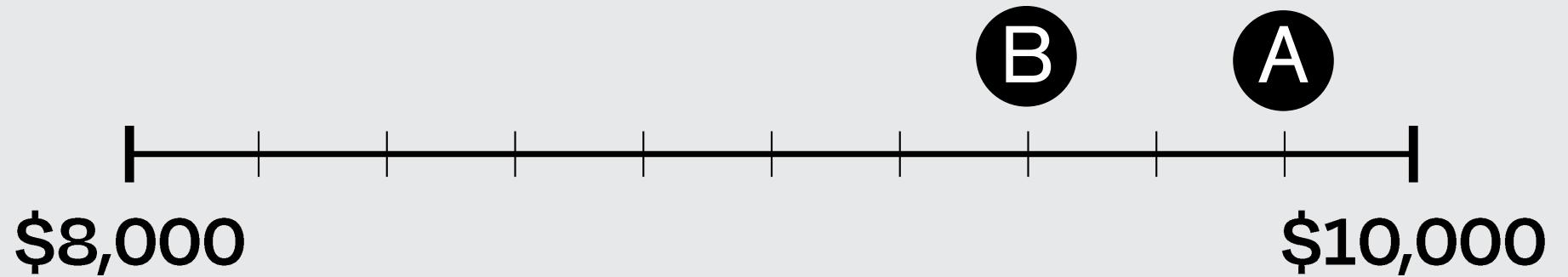
**MODELS**

# PRESTIGE SCORES



HOW MUCH WILL ALDO PAY FOR EACH MODEL?  
RECALL: THE PAY RANGE IS \$2,000

THE RANGE OF \$2,000 IS DIVIDED INTO 10 BINS



WITH MODEL A, ALDO PAYS \$9,800

WITH MODEL B, ALDO PAYS \$9,400

# WHAT ARE THE AGENCY PROFITS?



20%

80%

*The agent makes 20%  
of what the clients pays*

**WITH MODEL A, ALDO PAYS \$9,800**

**THE AGENCY GETS PAID \$1,960**  $=(0.20 \times 9800)$

**THE MODEL GETS PAID \$7,840**  $=(0.80 \times 9800)$

# EXECUTIVE SUMMARY

*In the dynamic world of fashion and modeling, agencies face the challenge of maximizing profitability while navigating ethical considerations, societal expectations, and environmental impact scrutiny.*

**Project Focus:** Our project seeks to tackle these challenges through a linear optimization lens, offering a structured approach to decision-making in modeling agencies.

## Industry Dynamics

**Goal:** Develop a comprehensive model that balances objectives such as profitability, ethics, diversity, and environmental sustainability, providing a relevant and innovative solution to talent placement and resource management complexities.

## Ethical Considerations

The modeling industry's influence on societal norms and its increasing pressure for diversity and inclusivity, coupled with growing environmental concerns, make this problem both pertinent and complex.

## Challenges

Aiming for a balance between conflicting objectives, ensuring economic viability alongside social and environmental responsibility in talent placement decisions.



**NEW YORK**  
**LOS ANGELES**  
**MIAMI**  
**TORONTO**

ABOUT  
BECOME ELITE  
CONTACT

INSTAGRAM  
FACEBOOK  
TIKTOK  
ELITE MODEL LOOK

**WOMEN**

**MEN**

**MEN**  
**DEVELOPMENT**  
**DIRECT**  
**LIFE**

**Modelling** *Modelling*

## Project Focus

*Developing an optimization model is crucial for Elite Model Look Toronto Agency, aiming to efficiently assign models to diverse jobs while navigating the complexities of the competitive and dynamic modeling industry. This model addresses the agency's modern needs and challenges.*

*Systematic approach to match models with jobs, maximizing satisfaction and efficiency.*



**PROBLEM  
CONTEXT**

### Balancing Factors

- *Model availability*
- *Skills & Experience*
- *Social influence*
- *Client preferences*
- *Environmental Considerations*
- *Ethical Considerations*
- *Industry Practices*
- *Job Requirements*

- **Constraints:** *Involve limited availability, varying pay scales, job requirements, travel logistics, and maintaining a diverse roster of models.*
- **Expected Outcomes:** *Potential benefits include increased efficiency in model-job matching, higher client and model satisfaction, and reduced environmental impact.*

# DIVERSE DATA SOURCES

- Key variables focus on model assignments, jobs requirements & information, travel/event scheduling, etc.

## • Data Collection and Integration:

- Meticulous integration of:
  - Synthetic data
  - Elite Model Look website-sourced data
  - Real communications data from a former Elite Model team member enriches job-related information.
  - Social Media
  - Google Flights data, etc.

Enhances realism and consistency of our model.

*Cf. Master Dataset.*

## Model Data

- Uniquely identified by model numbers
- Diverse attributes from physical features to professional metrics
- Ethnicity, Work History, factors contributing to the 'prestige score'
- Model availability for the week

- Geographical details - X and Y coordinates
- CO2 emissions costs on major urban centers ensuring sustainable considerations
- Models location, enabling spatial analysis

## Location & Emissions Data

## Jobs Data

- Client-specific job requirements - job location, preferred age, ethnicity, payment range limits, etc.
- Job packages, shoot duration, priority, client type, non-compete agreements

*The dataset empowers the agency to align models with suitable opportunities, optimizing resource utilization for successful collaborations*

# MATHEMATICAL MODELING

## Decision Variables

Let  $x_{ij}$  be a binary variable where  $x_{ij} = 1$  if model  $i$  is assigned to job  $j$ , and  $x_{ij} = 0$  otherwise.

## Objective Function

$$\text{Maximize } \sum_{i,j} [B_{ij} + \frac{R_{ij}}{10} \cdot (\alpha R_i + \beta E_i + \gamma F_i) - \delta C_{ij}] \cdot x_{ij}$$

1. **Base Payment:**  $B_{ij}$  representing the **base** payment (i.e. 'Pay Ranger Lower' in the jobs dataset) for assigning model  $i$  to job  $j$ .
2. **Payment Range:**  $R_{ij}$  representing the **range** of payment (i.e. 'Pay Ranger Upper' - 'Pay Range Lower' in the jobs dataset) for assigning model  $i$  to job  $j$ .
3. **Reputation Score:**  $R_i$  denoting the reputation score for model  $i$ , based on the previous work of the model.
4. **Experience Score:**  $E_i$  for model  $i$ , based on the number of years worked in the industry.
5. **Following Score:**  $F_i$  for model  $i$ , based on the number of Instagram followers a model has.
6. **CO2 Emissions:**  $C_{ij}$  for assigning model  $i$  to job  $j$ .

$\alpha$ ,  $\beta$ , and  $\gamma$ , are weight factors for reputation (0.7), experience (0.25), and following scores (0.05) respectively.

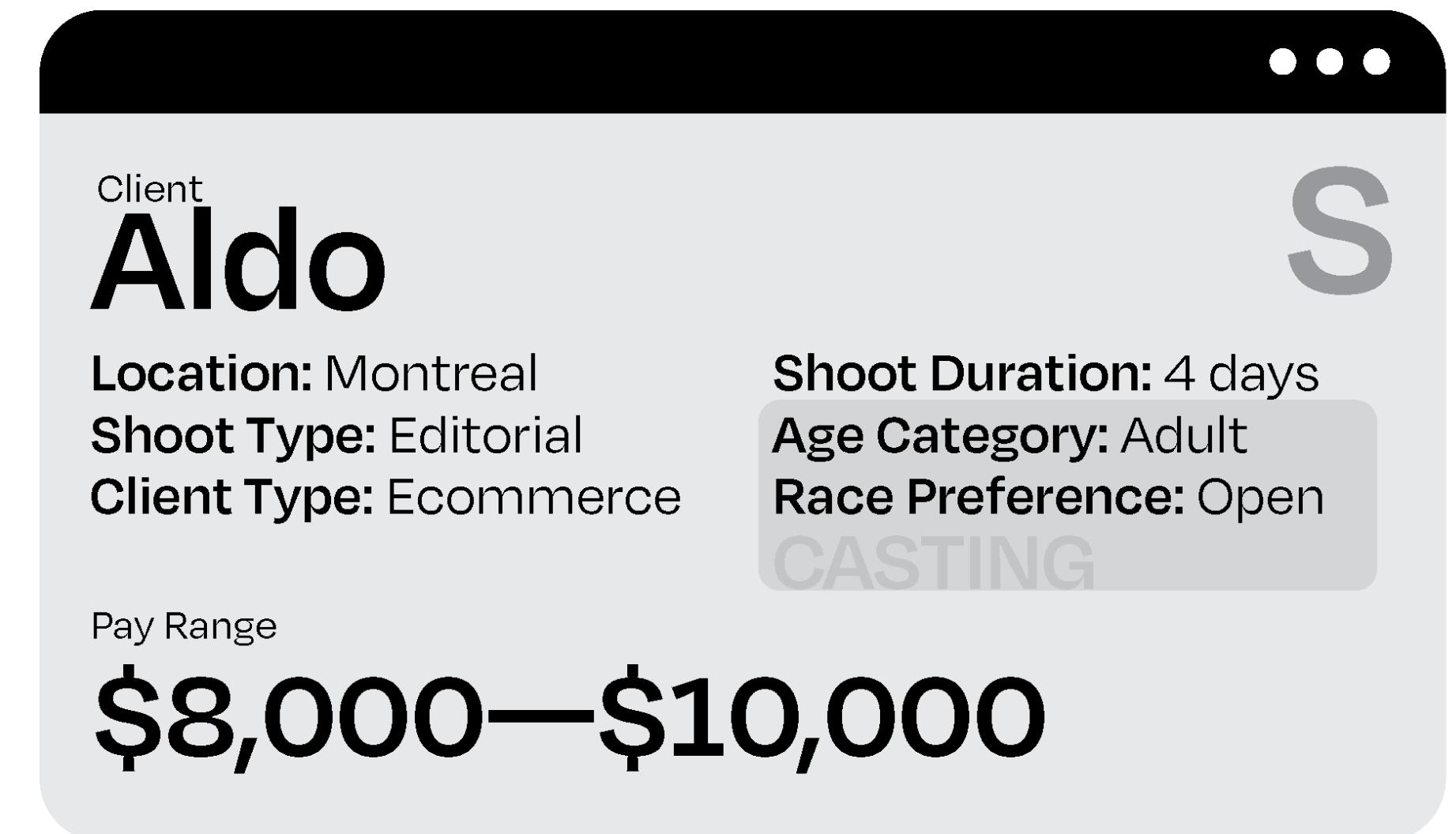
$\delta$  is the penalty factor for CO2 emissions, decided by the management of *EliteToronto*.

Note: The reputation, experience, and following scores are scaled to be between 0 and 10.

# CONSTRAINTS

- 1 Non-Conflicting Assignment**
- 2 One Job per Model**
- 3 One Model per Job**
- 4 Race Preferences**
- 5 Age-Category Preferences**

# MATHEMATICAL MODELING



# MATHEMATICAL MODELING

## Constraints

### 1. Non-Conflicting Assignments Constraint:

For each model and each day of the week, the model can be assigned to at most one job that requires availability on that day.

$$\sum_{j \in \text{Jobs}_{\text{day}}} x_{ij} \leq 1, \quad \forall i \in \text{Models}, \text{day} \in \text{Week}$$

### 2. One Model Per Job Constraint:

Each job can be assigned to at most one model.

$$\sum_{i \in \text{Models}} x_{ij} \leq 1, \quad \forall j \in \text{Jobs}$$

### 3. One Job Per Model Constraint:

Each model can be assigned to at most one job.

$$\sum_{j \in \text{Jobs}} x_{ij} \leq 1, \quad \forall i \in \text{Models}$$

# MATHEMATICAL MODELING

## Constraints

### 4. Ethnicity Constraints:

Let  $e_{ij}$  be a binary parameter that is 1 if the ethnicity of model  $i$  matches the ethnicity preference of job  $j$  or if the job preference is 'open', and 0 otherwise.

For all models  $i$  and all jobs  $j$ :

$$x_{ij} \leq e_{ij} \quad \forall i \in \text{Models}, \forall j \in \text{Jobs}$$

### 5. Age Category Preference Constraints:

Let  $a_{ij}$  be a binary parameter that is 1 if the age category of model  $i$  matches the age category preference of job  $j$  or if the job preference is 'open', and 0 otherwise.

For all models  $i$  and all jobs  $j$ :

$$x_{ij} \leq a_{ij} \quad \forall i \in \text{Models}, \forall j \in \text{Jobs}$$

# PROBLEM EXTENSIONS & CHALLENGES

## 1. Non-Compete Agreement:

If a model has a non-compete agreement with a client type, the model cannot be assigned to any other job with the same client type.

For each model  $i$  and each unique client type  $c$  in the set of all client types  $C$ , the non-compete constraint is:

$$\sum_{j \in \text{Jobs} | c_j = c} x_{ij} \leq n_{ic} \quad \forall i \in \text{Models}, \forall c \in C$$

where  $c_j$  is the client type of job  $j$ , and  $n_{ic}$  is a binary variable which is 1 if model  $i$  does not have a non-compete agreement for client type  $c$ , and 0 otherwise.

# PROBLEM EXTENSIONS & CHALLENGES

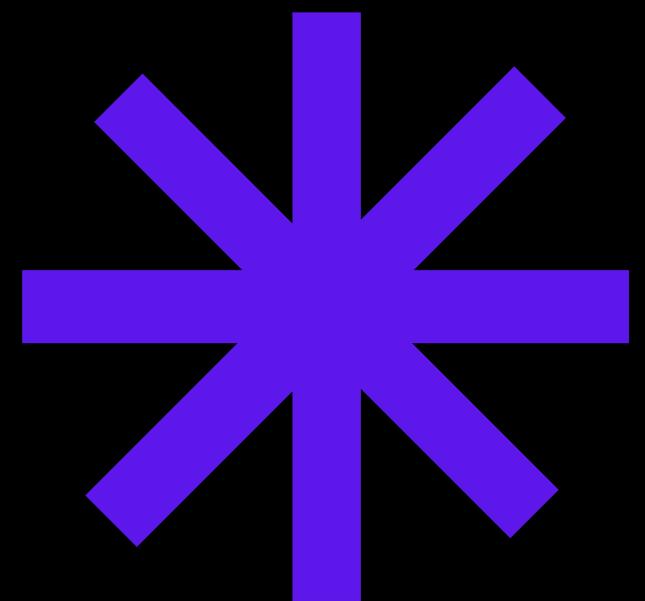
2. **Job-Package:** In reality, some jobs do not pay for the model's travel and accomodation. The revised objective function to incorporate the job-package costs would be:

$$\text{Maximize} \sum_{i,j} \left[ B_{ij} + \frac{R_{ij}}{10} \cdot (\alpha R_i + \beta E_i + \gamma F_i) - (1 - k_j) \cdot (D_j R_j + T_{ij}) - \delta C_{ij} \right] \cdot x_{ij}$$

where

- $k_j$  is a binary variable which is 1 if the job  $j$  pays for the model's travel and accomodation, and 0 otherwise.
- $D_j$  is the duration of the shoot for job  $j$  (in days).
- $R_j$  is the average cost of cost of accomodation per night in the city of job  $j$ .
- $T_{ij}$  is the air travel cost from the model  $i$ 's home city to job  $j$ 's location.

# RESULTS & RECOMMENDATIONS



# MODEL ALLOCATION MAPPING

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MODEL TO JOB ASSIGNMENT WHEN THE EMISSION WEIGHT = 0  
1 UNASSIGNED



MODEL TO JOB ASSIGNMENT WHEN THE EMISSION WEIGHT = 0.50  
1 UNASSIGNED



MODEL TO JOB ASSIGNMENT WHEN THE EMISSION WEIGHT = 1  
1 UNASSIGNED



MODEL TO JOB ASSIGNMENT WHEN THE EMISSION WEIGHT = 15  
1 UNASSIGNED



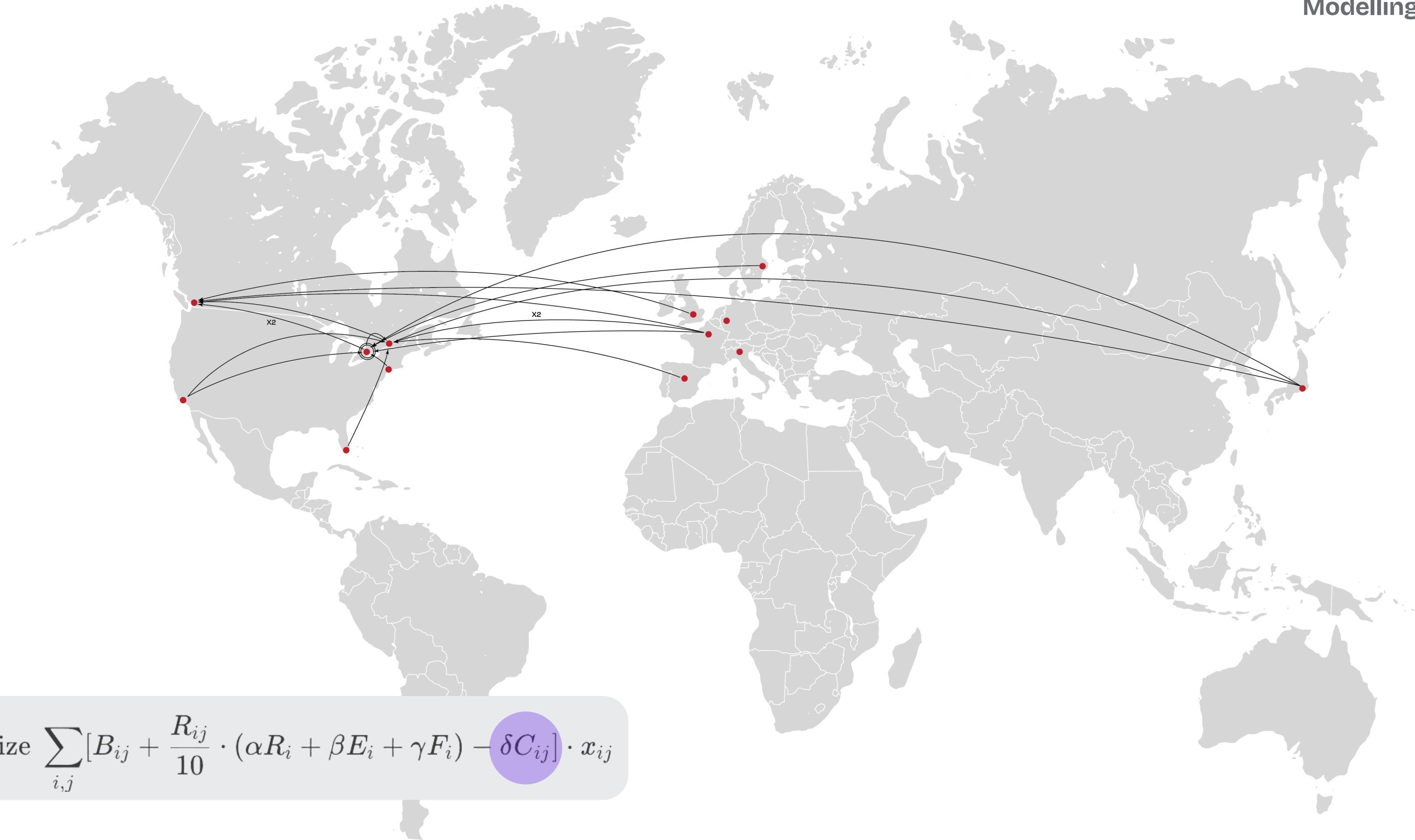
MODEL TO JOB ASSIGNMENT WHEN THE EMISSION WEIGHT = 30  
8 UNASSIGNED



# MODEL TO JOB ASSIGNMENT WHEN THE EMISSION WEIGHT = 0

1 UNASSIGNED

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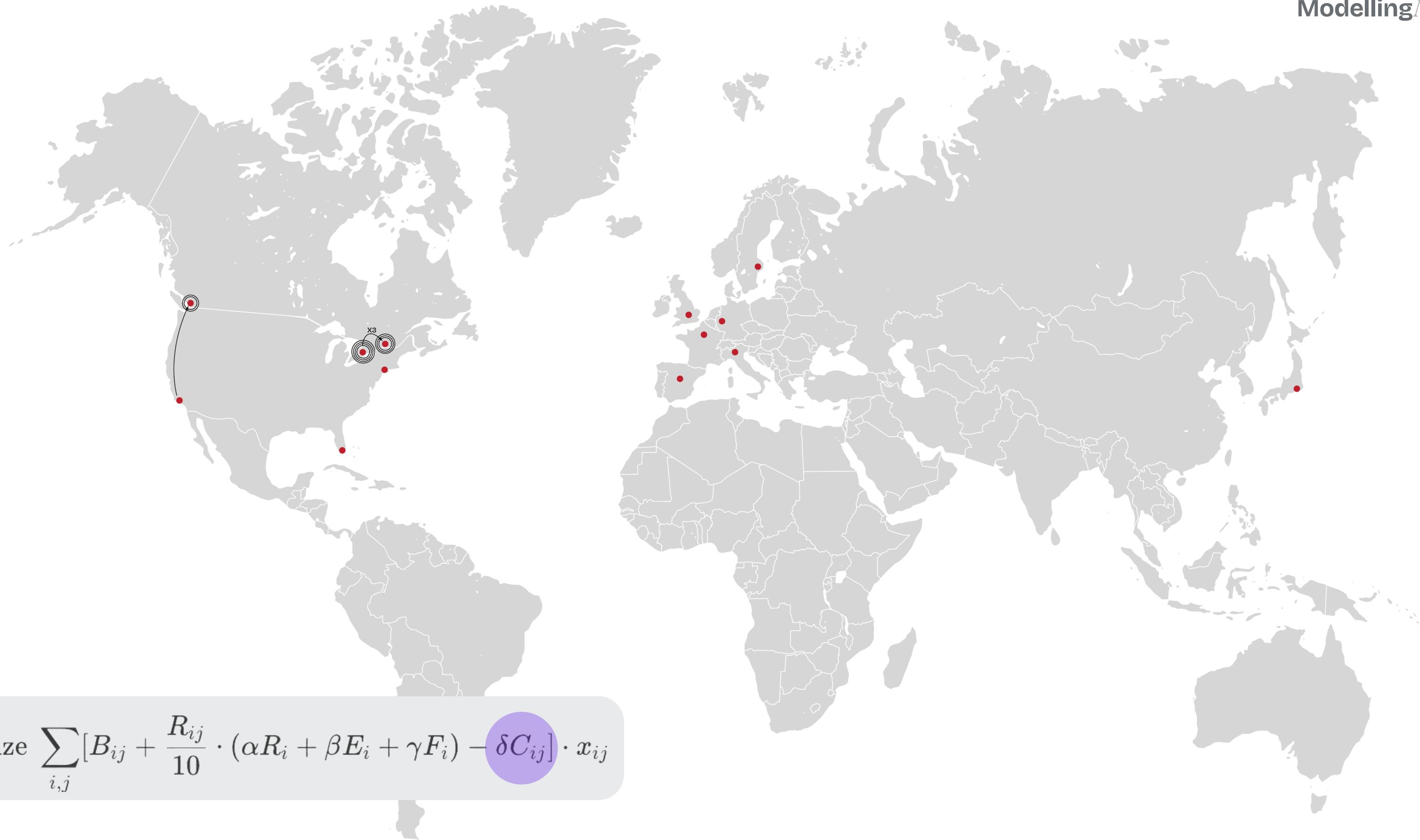


$$\text{Maximize } \sum_{i,j} [B_{ij} + \frac{R_{ij}}{10} \cdot (\alpha R_i + \beta E_i + \gamma F_i) - \delta C_{ij}] \cdot x_{ij}$$

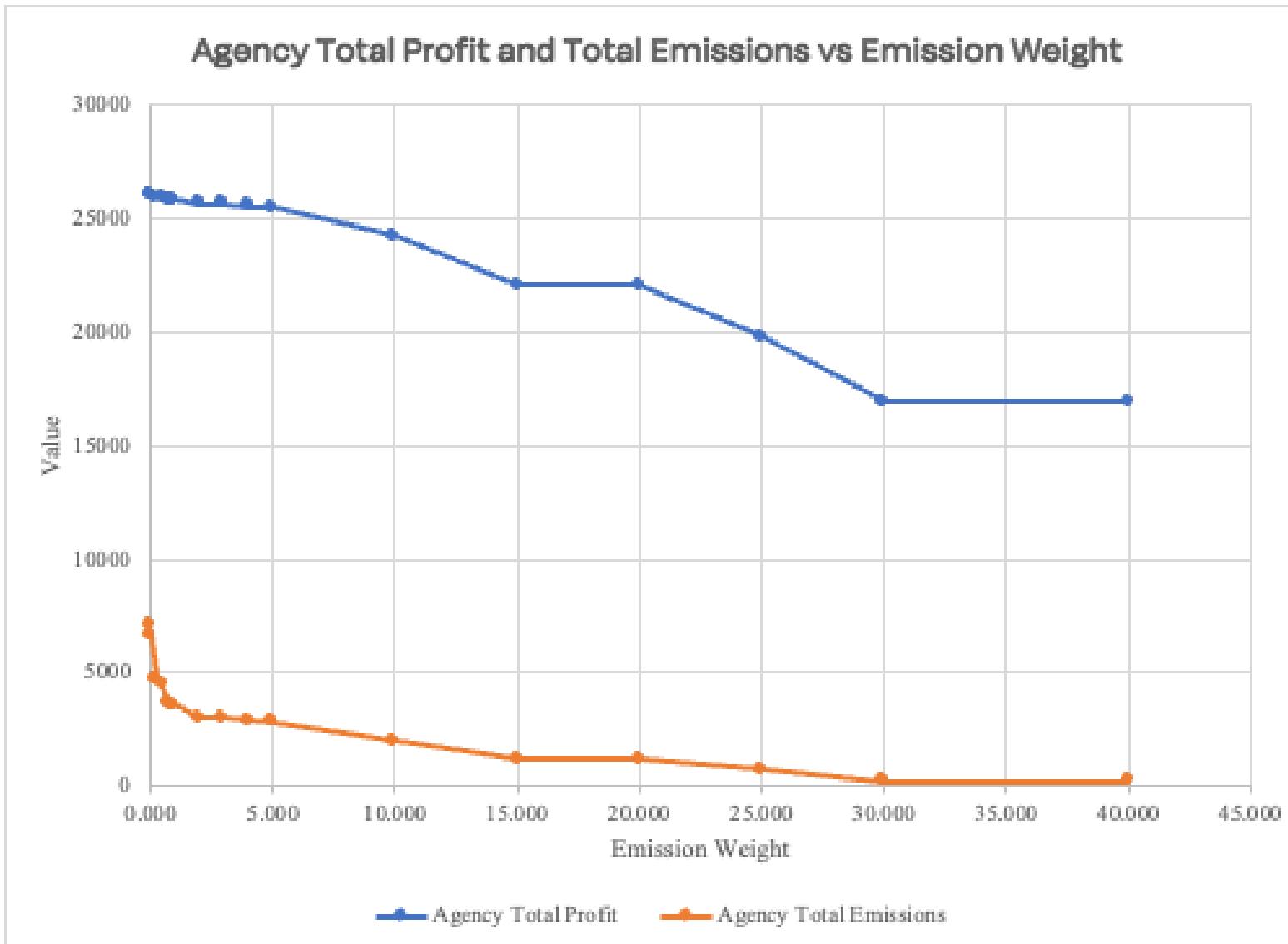
# MODEL TO JOB ASSIGNMENT WHEN THE EMISSION WEIGHT = 30

8 UNASSIGNED

Modelling *Modelling*

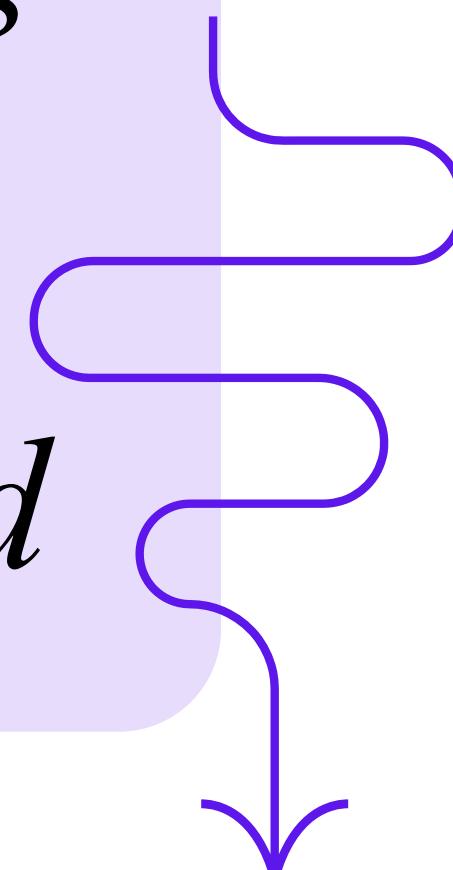
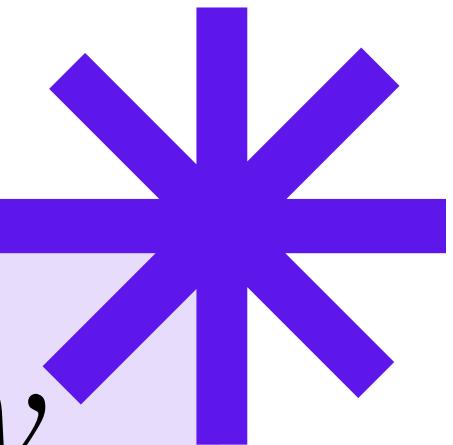


# KEY TAKEAWAY



*the more the agency prioritizes reducing CO<sub>2</sub> emissions*

*the more their total profits are impacted*



# RELEVANT INSIGHTS



## TALENT POOL

The optimization model has highlighted limitations in our talent pool as some jobs remain unassigned.

This indicates gaps in our agency's capacity to fulfill certain job criteria, possibly due to a lack of diversity in our models' skills or scheduling constraints.



## COST OPTIMIZATION

The project's focus on incorporating travel and accommodation expenses into the optimization model offers a nuanced view of cost-efficiency.

The agency can strategically select model assignments that not only meet the job's requirements but also minimize financial outlay.



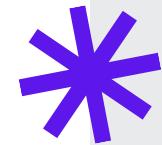
## SUSTAINABILITY

Incorporating CO<sub>2</sub> emissions into the model promotes environmentally conscious decision-making. It allows the agency to reduce its carbon footprint by opting for local talent and collaborating with eco-friendly clients. This strategic move not only advances sustainability goals but also strengthens the agency's reputation as an environmentally responsible entity in the marketplace.

# POTENTIAL AREAS FOR FUTURE EXPLORATION



**SUSTAINABILITY INDEXING**



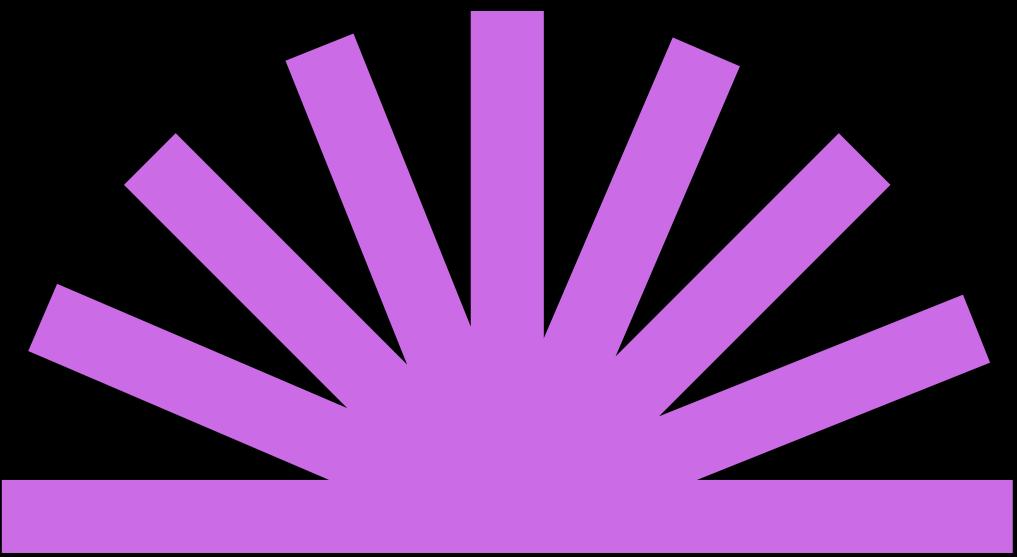
**MODEL CAREER ADVANCEMENT**



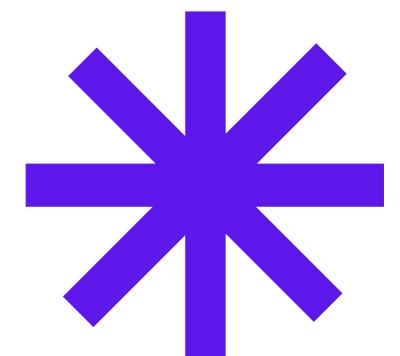
**EMERGENCY BACK-UPS**



# THANK YOU



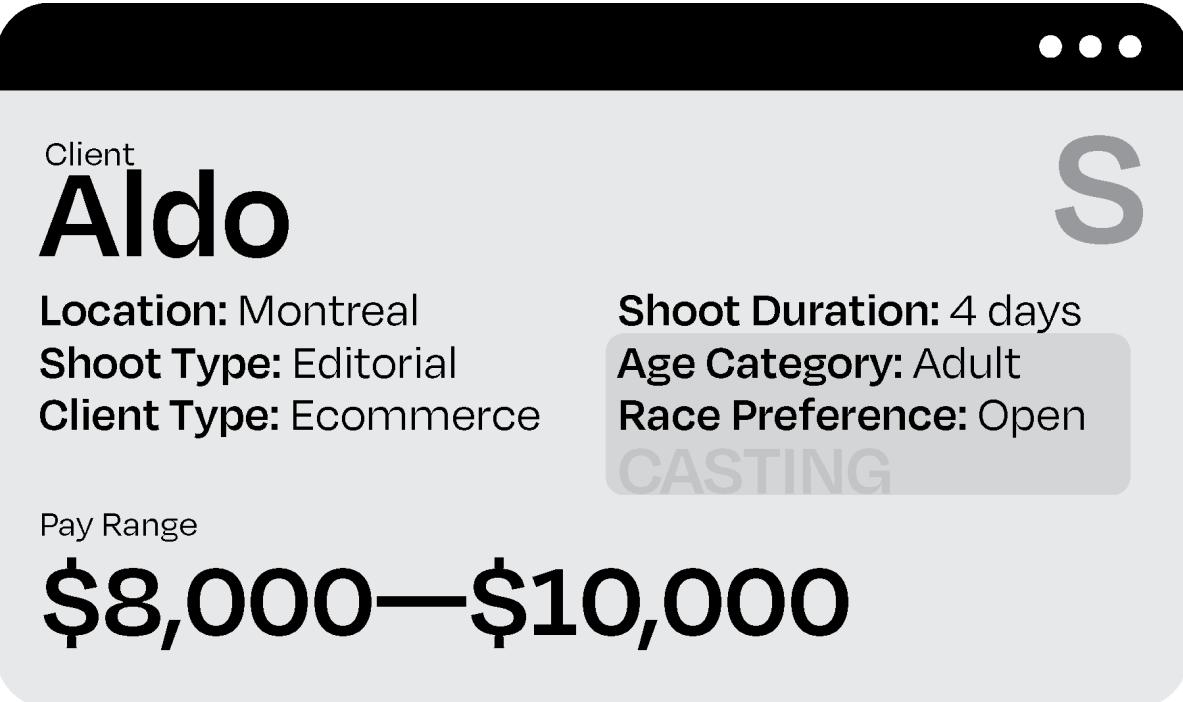
# APPENDIX



# RESULTS

<u>Model Name</u>	<u>Model Location</u>	<u>Job Name</u>	<u>Job Location</u>	<u>Job Package</u>	<u>Total Job Pay</u>	<u>Total Travel Cost</u>	<u>Model Net Income</u>	<u>Agency Profit</u>	<u>CO2 Emissions</u>
Brodie Scott	Tokyo	Frank And Oak	Vancouver	Yes	5149.35	0	4119.48	1029.87	639
Dwight Ireland	Toronto	Maple Leaf Diamonds	Toronto	Yes	6192.55	0	4954.04	1238.51	0
Hector Raptis	Toronto	Herschel Supply Co.	Vancouver	Yes	6472.43	0	5177.95	1294.49	229
Jacob Cheng	Toronto	Lululemon	Vancouver	Yes	6124.68	0	4899.75	1224.94	229
Justin Lyons	Toronto	Flare Magazine	Toronto	Yes	4523.09	0	3618.47	904.62	0
Malik Lindo Ireland	New York	Canadian Tire	Toronto	No	6883.33	762	4897.07	1224.27	54
Steven Smith	Los Angeles	Holt Renfrew	Toronto	No	7189.56	1016	4938.85	1234.71	244
Michael Gonzalez	Los Angeles	Canada Goose	Vancouver	No	6168.7	615	4442.96	1110.74	117
Steven Lee	Paris	Blue Ruby	Toronto	No	5488.38	799	3751.51	937.88	397
Robert Smith	Paris	Bombardier	Montreal	No	8122.32	1495	5301.86	1325.46	376
Michael Lopez	Tokyo	Aldo	Montreal	No	8182.86	2556	4501.49	1125.37	929
Joshua White	Montreal	Lincoln	Vancouver	No	8937	963	6379.2	1594.8	252
Daniel Jones	Tokyo	Rudsak	Montreal	No	5120.73	2412	2166.98	541.75	929
Ronald Harris	Tokyo	Peace Collective	Toronto	Yes	5953.64	0	4762.91	1190.73	782
Andrew Jackson	Paris	Smythe	Toronto	Yes	5240.41	0	4192.33	1048.08	397
Richard Jones	Paris	Roots Canada	Vancouver	Yes	6547.5	0	5238	1309.5	579
Michael Smith	Miami	Arc'teryx	Montreal	Yes	7400.05	0	5920.04	1480.01	205
David Smith	Stockholm	Magazine "Elle Canada"	Toronto	Yes	4843.3	0	3874.64	968.66	455
Kevin Moore	Madrid	Birks	Montreal	Yes	7330.65	0	5864.52	1466.13	445
David Thomas	Toronto	Mackage	Montreal	No	7615.77	732	5507.02	1376.75	40
Unassigned	N/A	Moose Knuckles	Vancouver	No	3500 - 6900	N/A	0	0	N/A

# REPUTATION CALCULATION



A 8 reputation score

## Model's Work History

- Job 1: A 8
- Job 2: S 10
- Job 3: A 8
- Job 4: B 6

## Calculation

$$8+10+8+6=32$$
$$32/4=8$$