
Dynamic Data Driven Crowd Sensing Task Assignment

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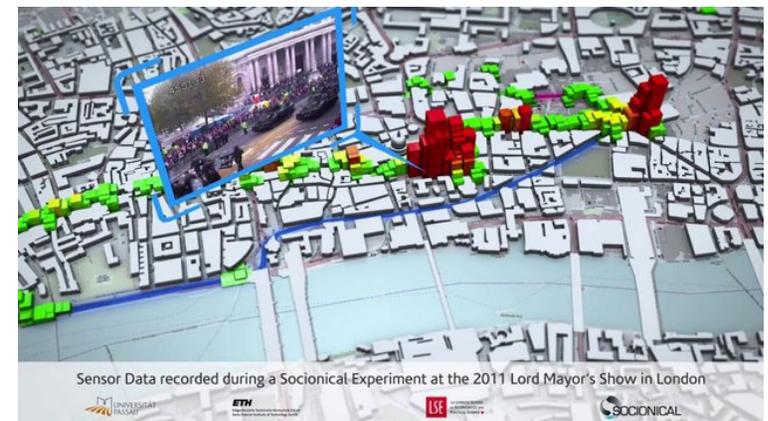
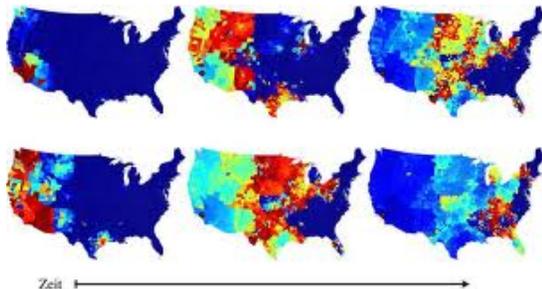
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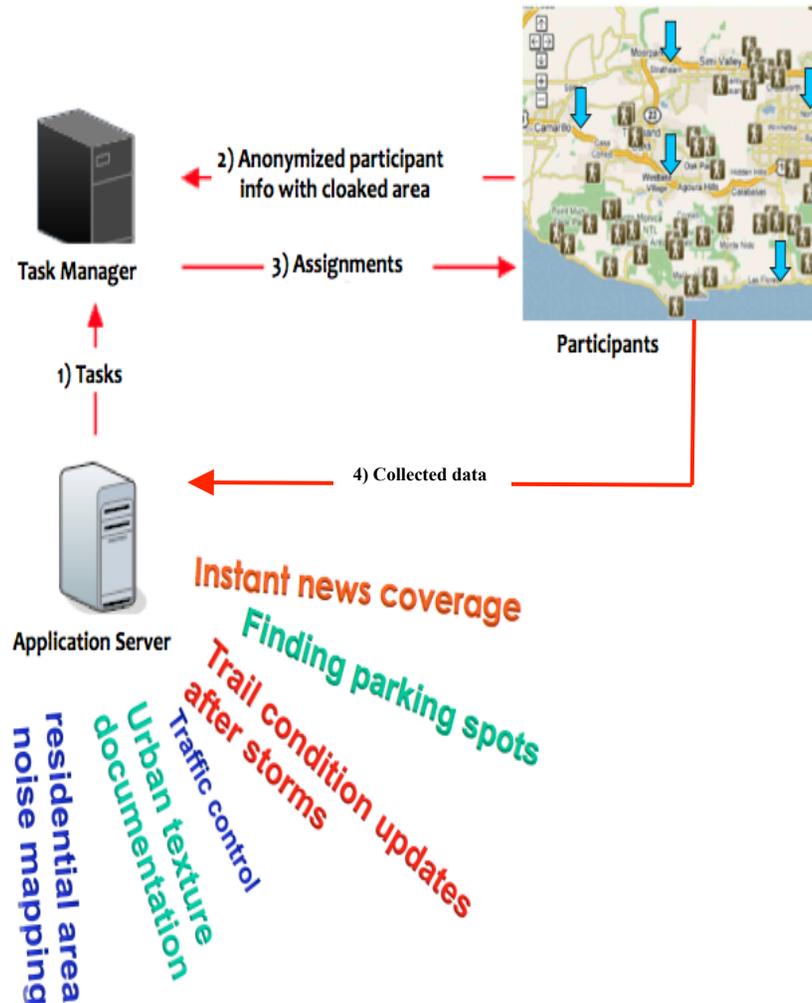
Crowd Sensing

- Sensing paradigm using mobile devices
- Syndromic surveillance, traffic monitoring, news coverage, intelligence data collection, crowd/congestion monitoring



Spatial Crowd Sensing Task Assignment

- Sensing tasks: collect data about specific **targets** at particular **locations**.
- Task assignment
- Maximize sensing coverage
- Minimize sensing cost



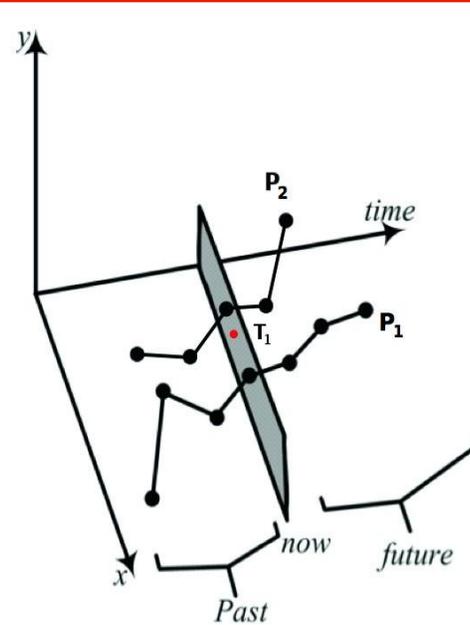
Challenges

- **Uncertain and dynamic trajectories of participants:**

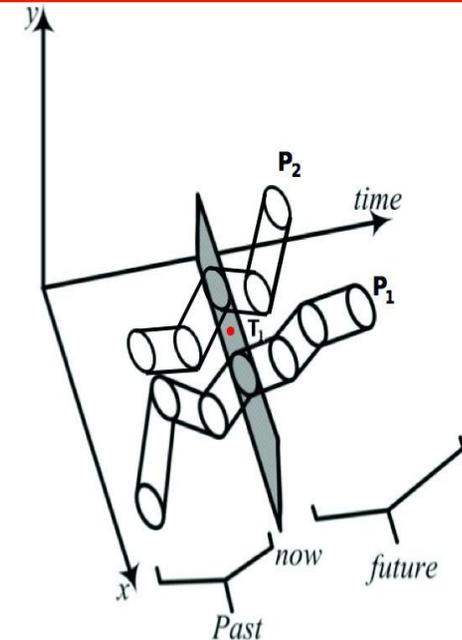
1) Error in location-detection devices, noisy transmissions, or explicit location perturbation due to privacy concerns

2) Participants may be constantly moving (e.g. commuters) and

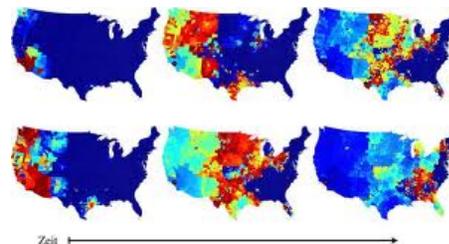
- **Dynamic tasks:** Sensing tasks may be updated (e.g. spread of disease outbreak, moving crowd in an event)



a) Exact Trajectories



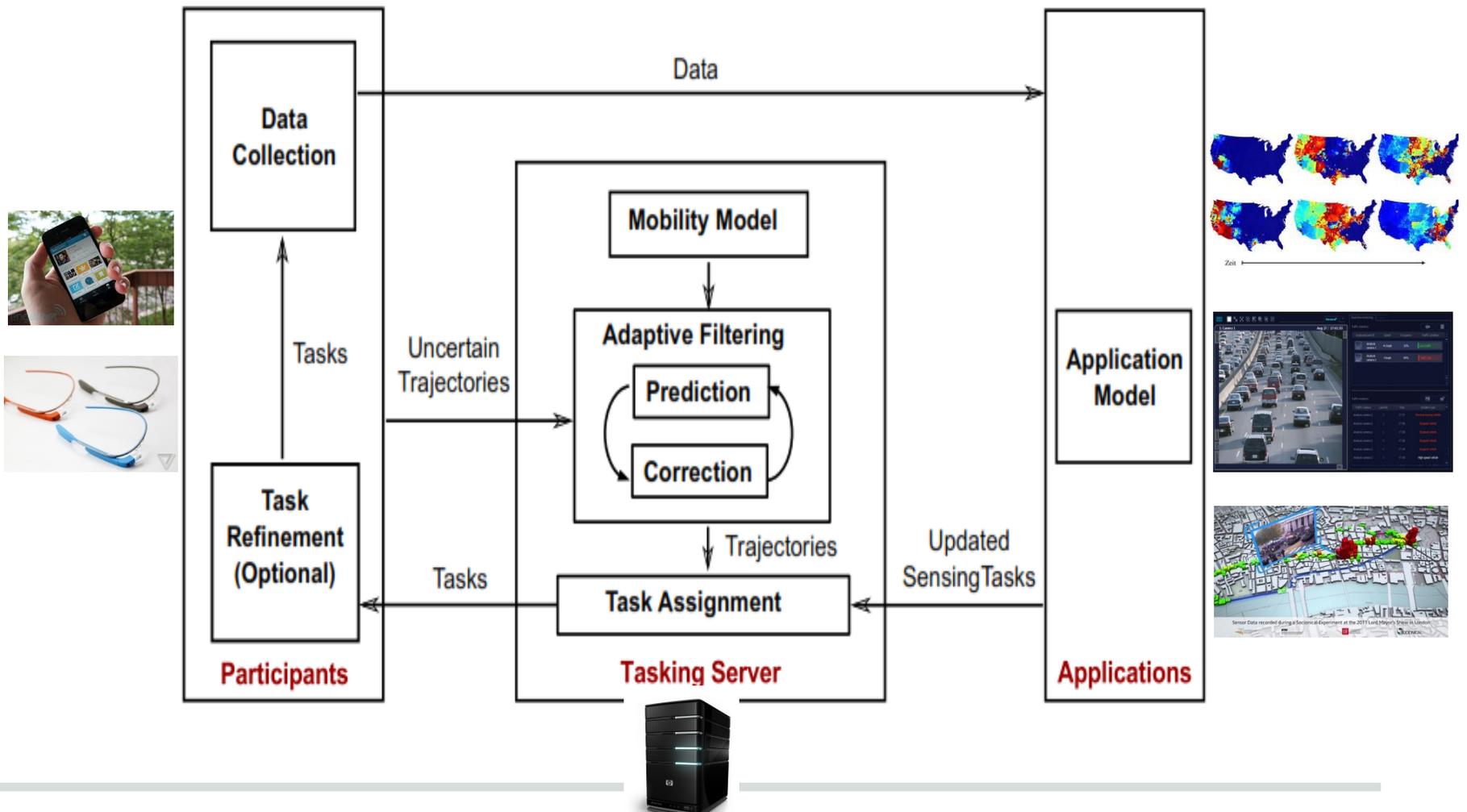
b) Uncertain Trajectories



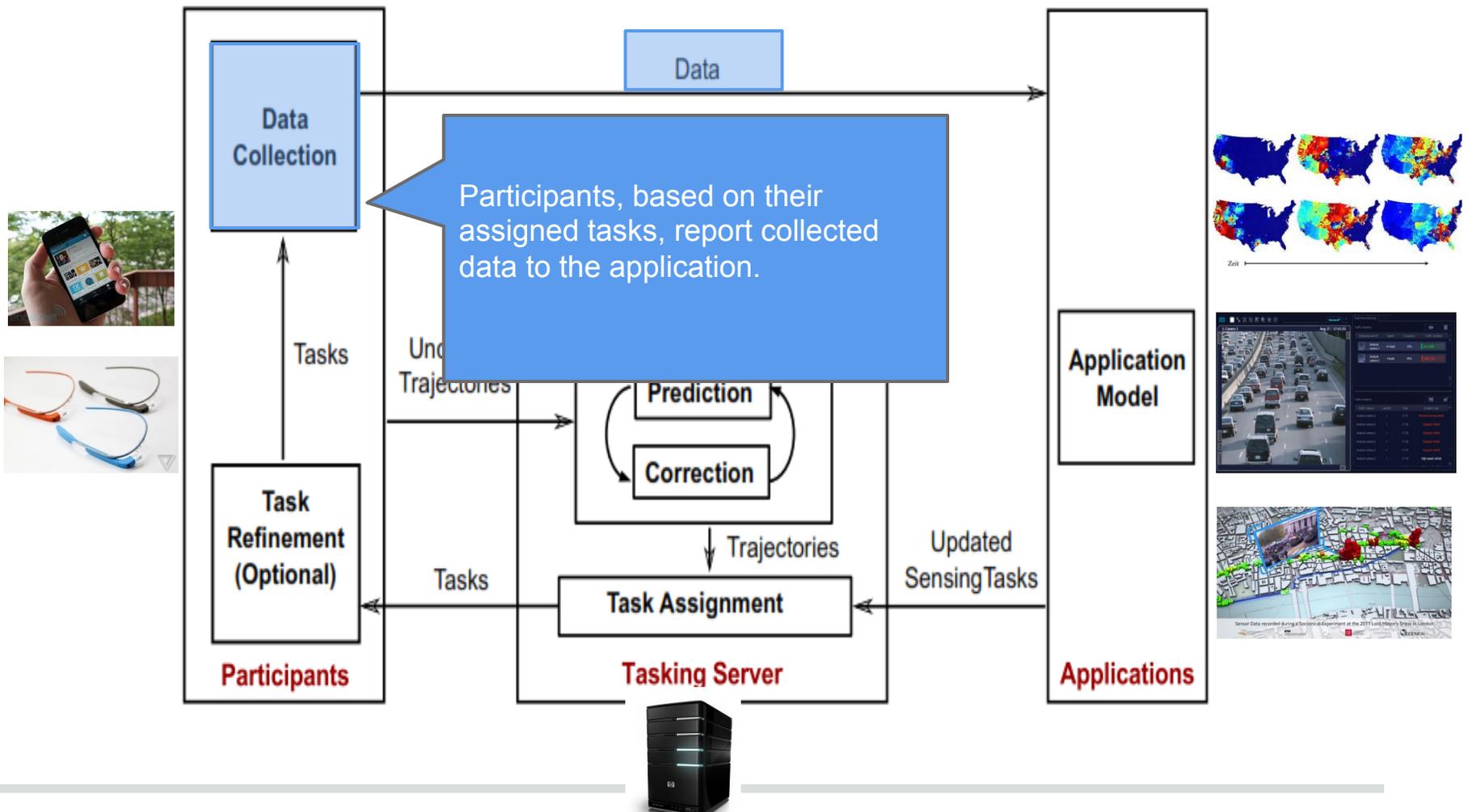
Contributions

- A dynamic data driven framework for spatial task assignment in mobile crowd sensing with dynamic and uncertain participant locations
 - Utilize DDDAS to steer and assign the sensing tasks in targeted ways, adapting dynamically to
 - 1) sensing/application needs
 - 2) uncertain and moving locations of participants
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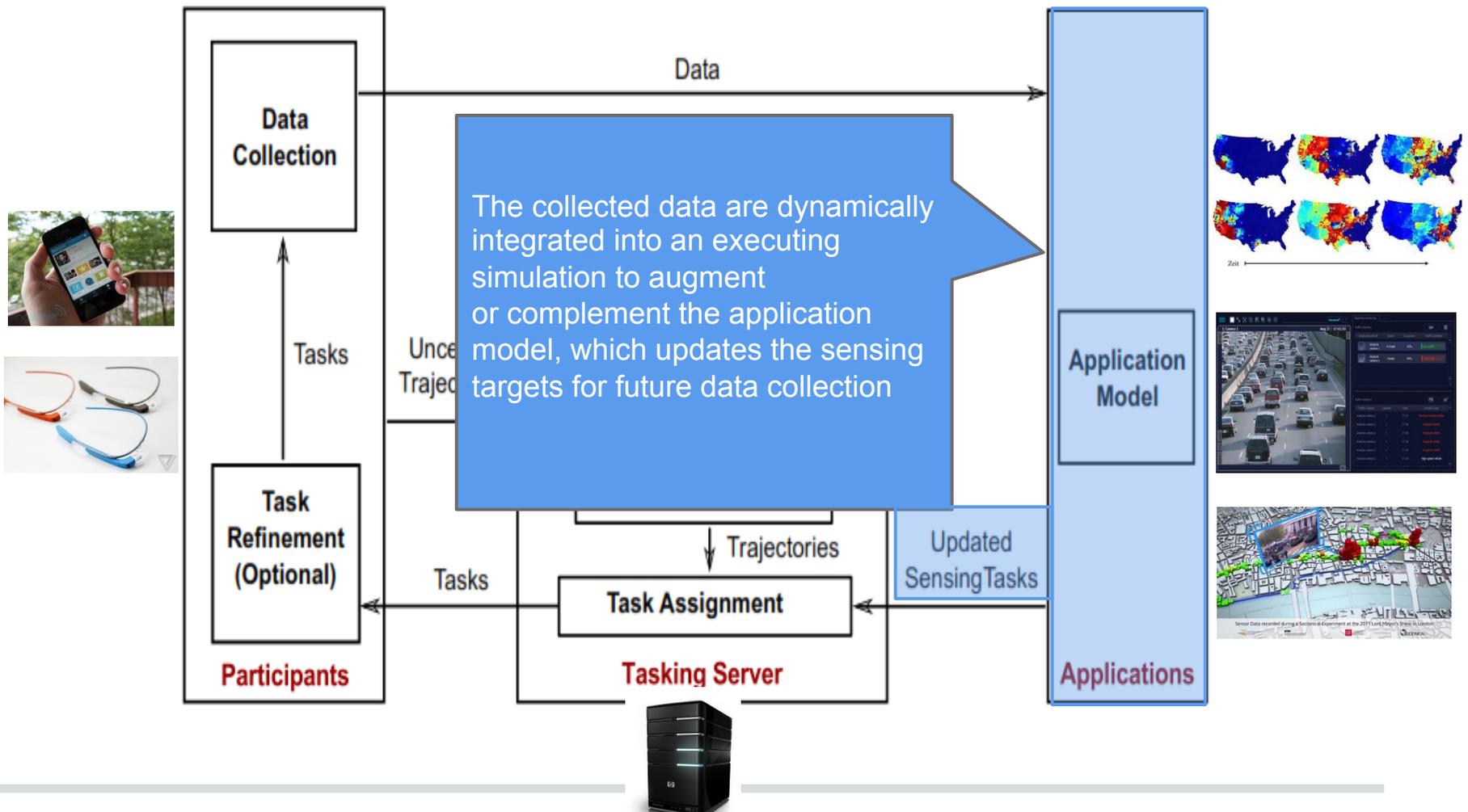
Adaptive dynamic data driven framework



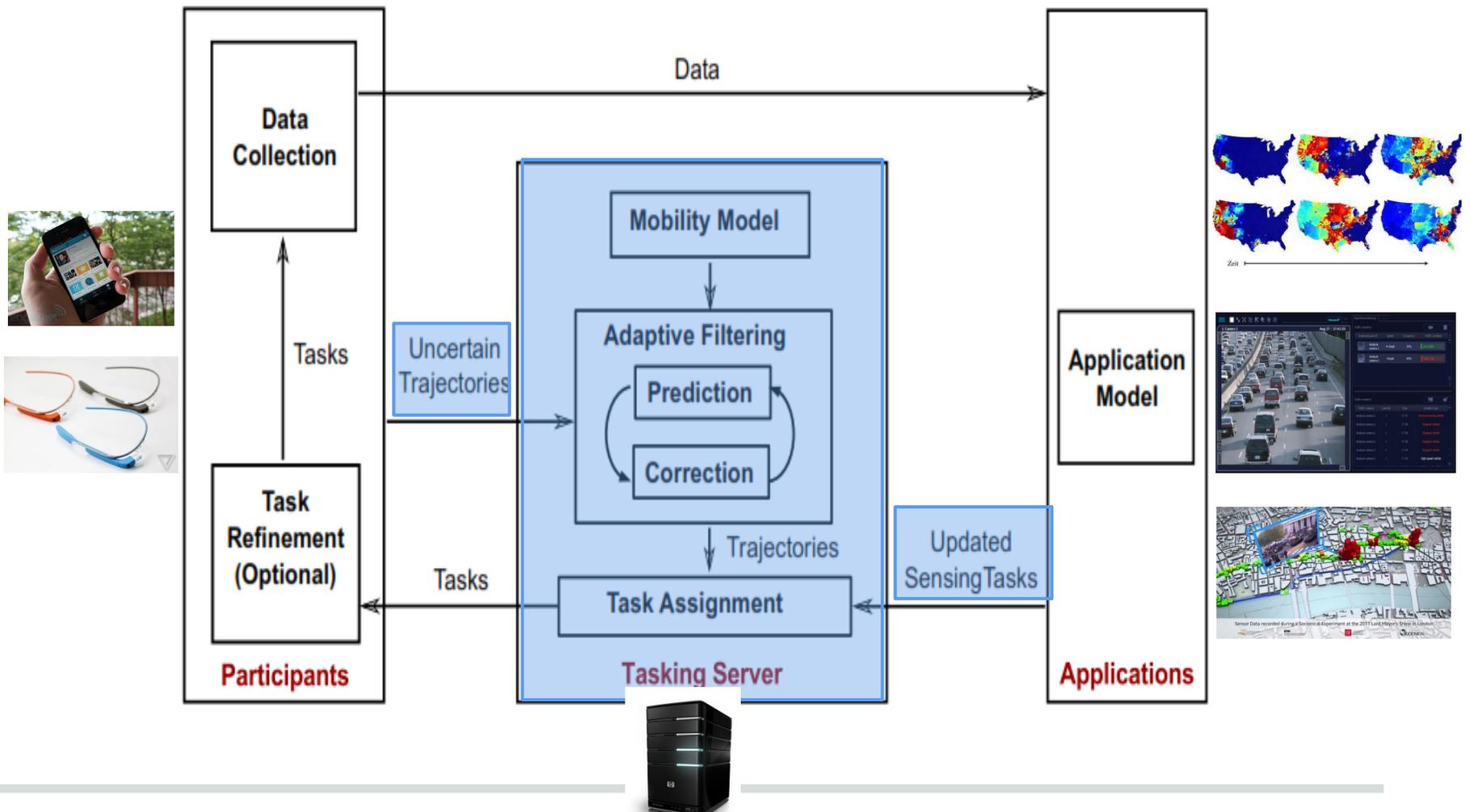
Adaptive dynamic data driven framework



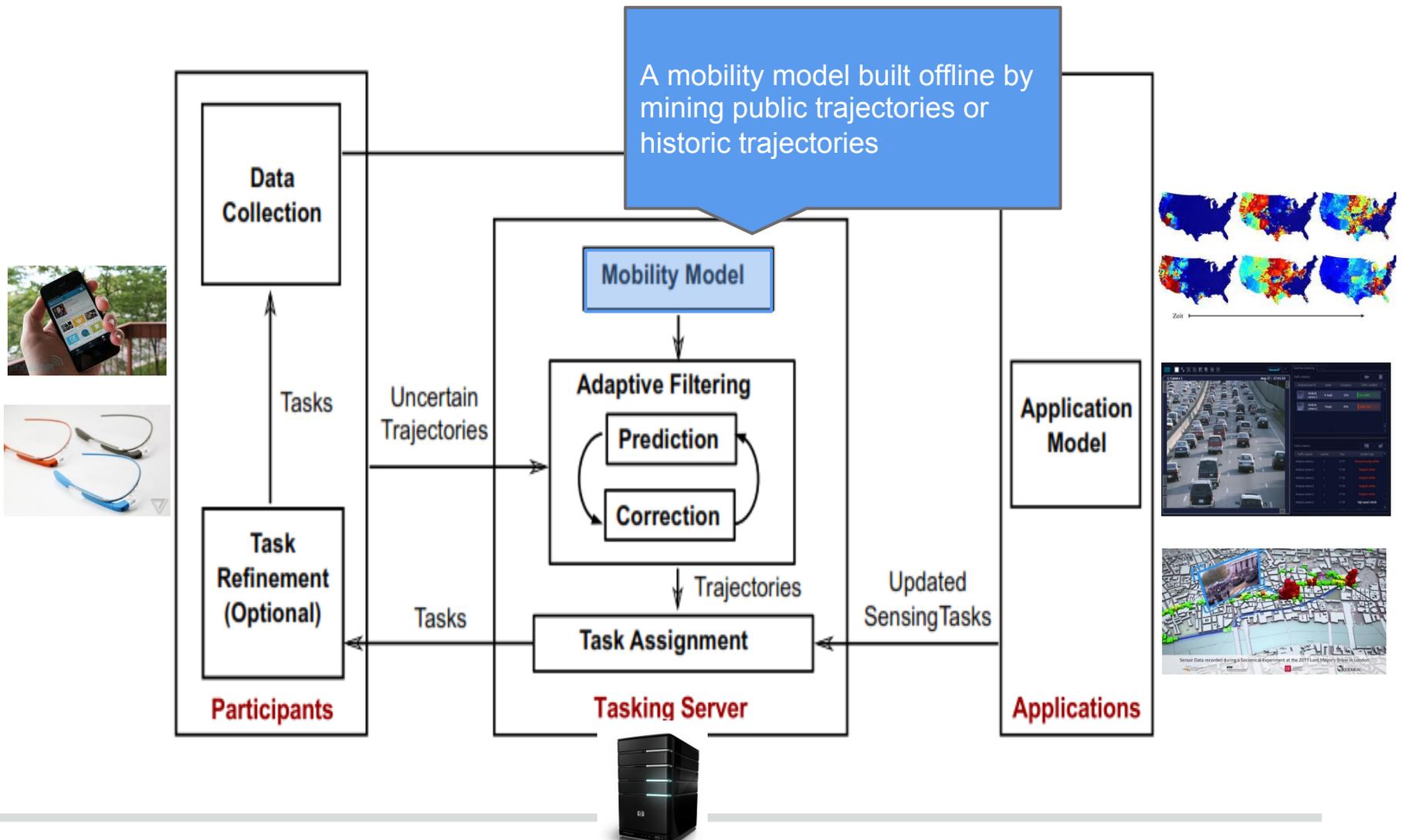
Adaptive dynamic data driven framework



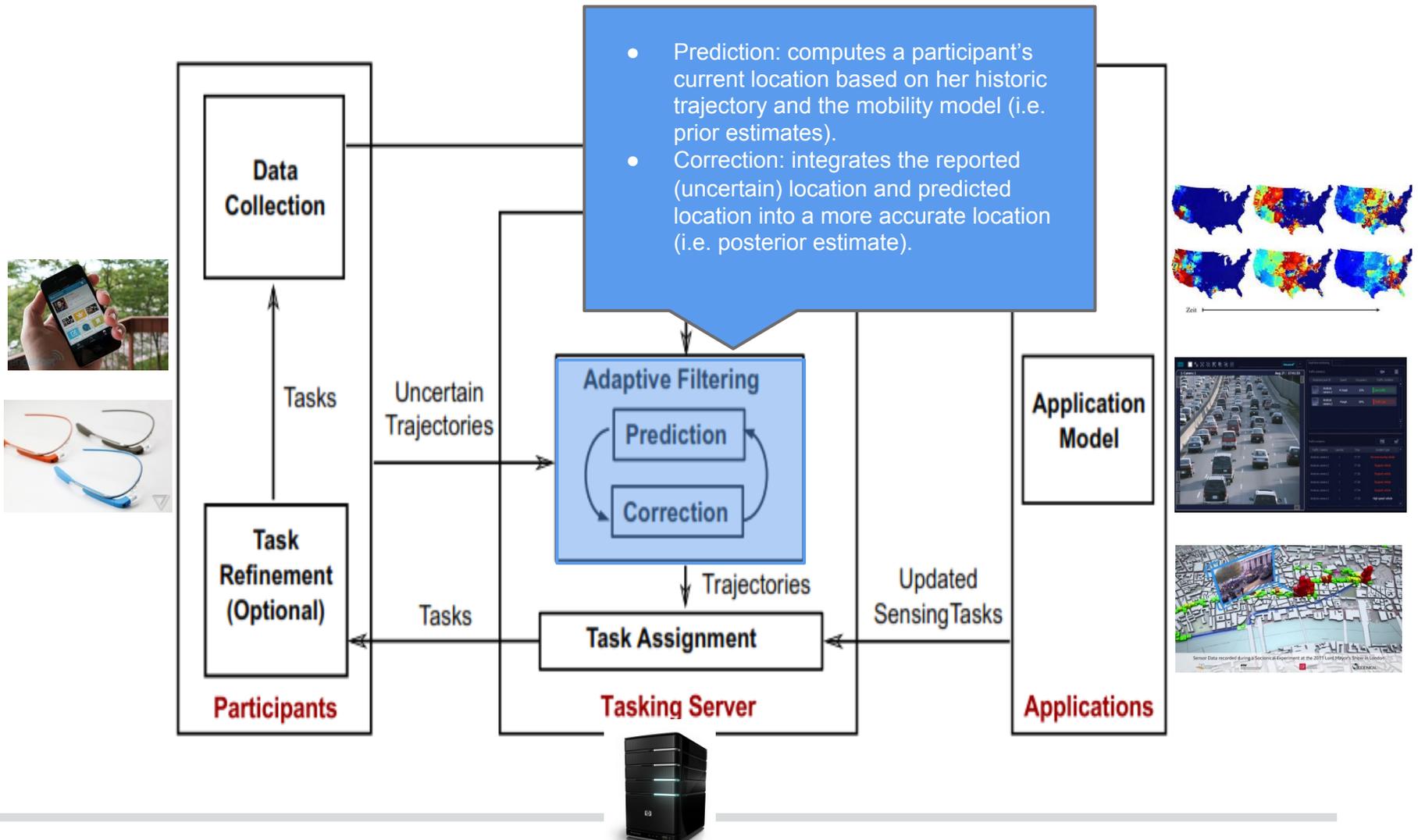
Adaptive dynamic data driven framework



Adaptive dynamic data driven framework



Adaptive dynamic data driven framework



Bayesian Filtering

- **State space model:**

$$X_t \sim p(X_t | X_{t-1})$$

$$Z_t \sim p(Z_t | X_t)$$

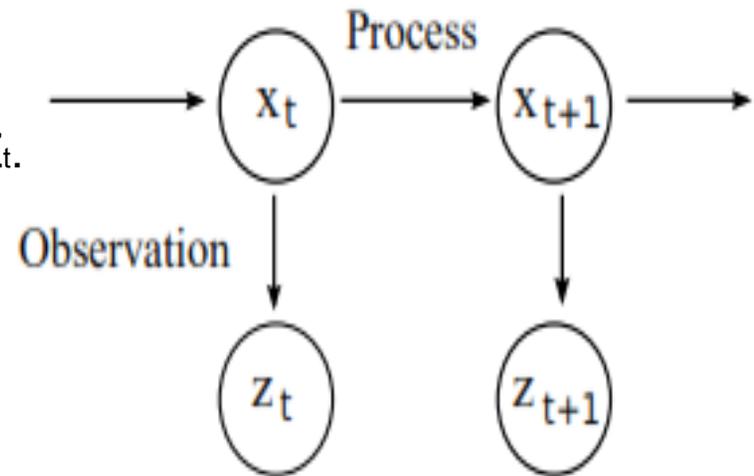
- **Measurement sequence** $Z_{1:t} = Z_1, \dots, Z_t$.

Prediction:

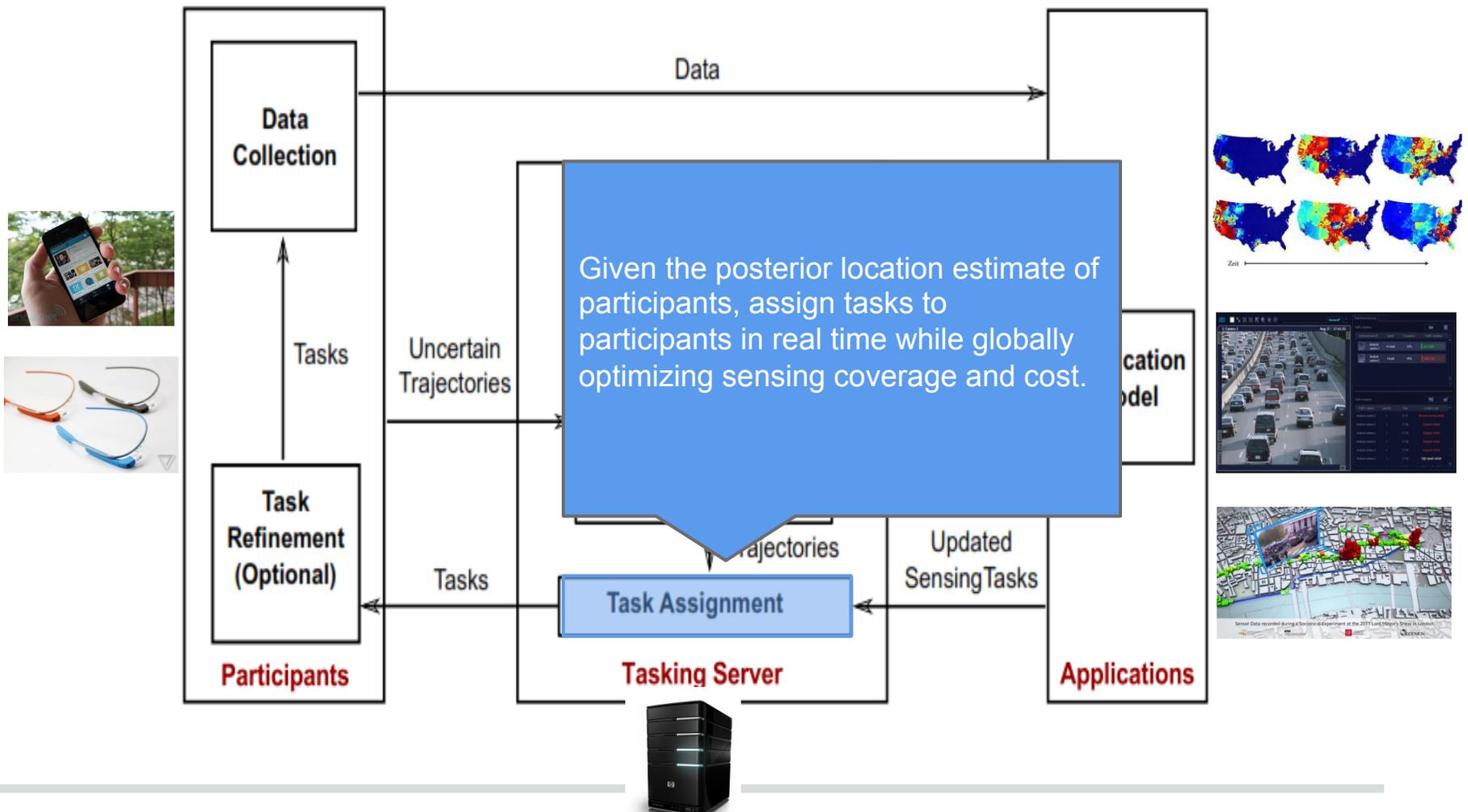
$$p(X_t | Z_{1:t-1}) = \sum p(X_t | X_{t-1}) p(X_{t-1} | Z_{1:t-1})$$

Update:

$$p(X_t | Z_{1:t}) = \frac{p(Y_t | X_t) p(X_t | Z_{1:t-1})}{\sum p(Y_t | X_t) p(X_t | Z_{1:t-1})}$$



Adaptive dynamic data driven framework



Global Task Assignment

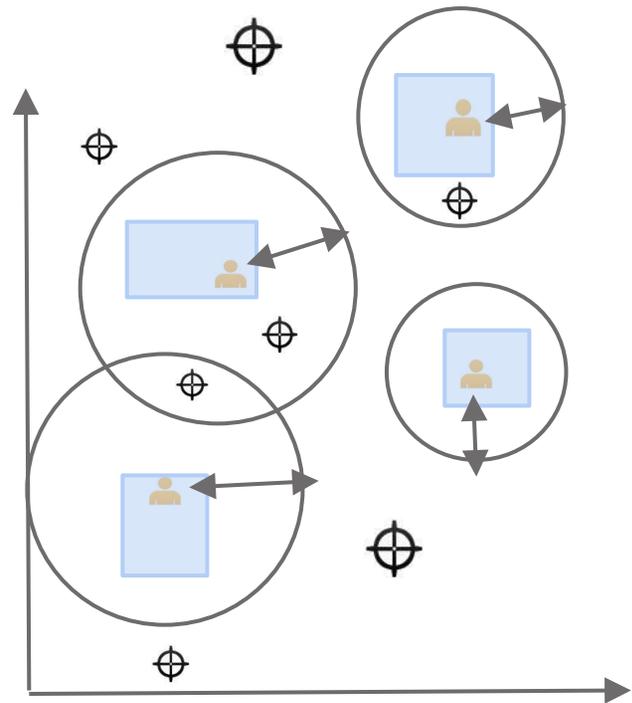
- Given a set of N participants with **uncertain locations** and **sensing range**, a set of M sensing targets, a desired task coverage goal, assign targets to qualified participants with minimum cost

- Task Cost :

$$TC = \sum_{j \in M} \sum_{i \in N} x_{i,j} d_{i,j}$$

- Task Coverage:

$$TU = \sum_{j \in M} \frac{\sum_{i \in N} x_{i,j}}{k_j}$$

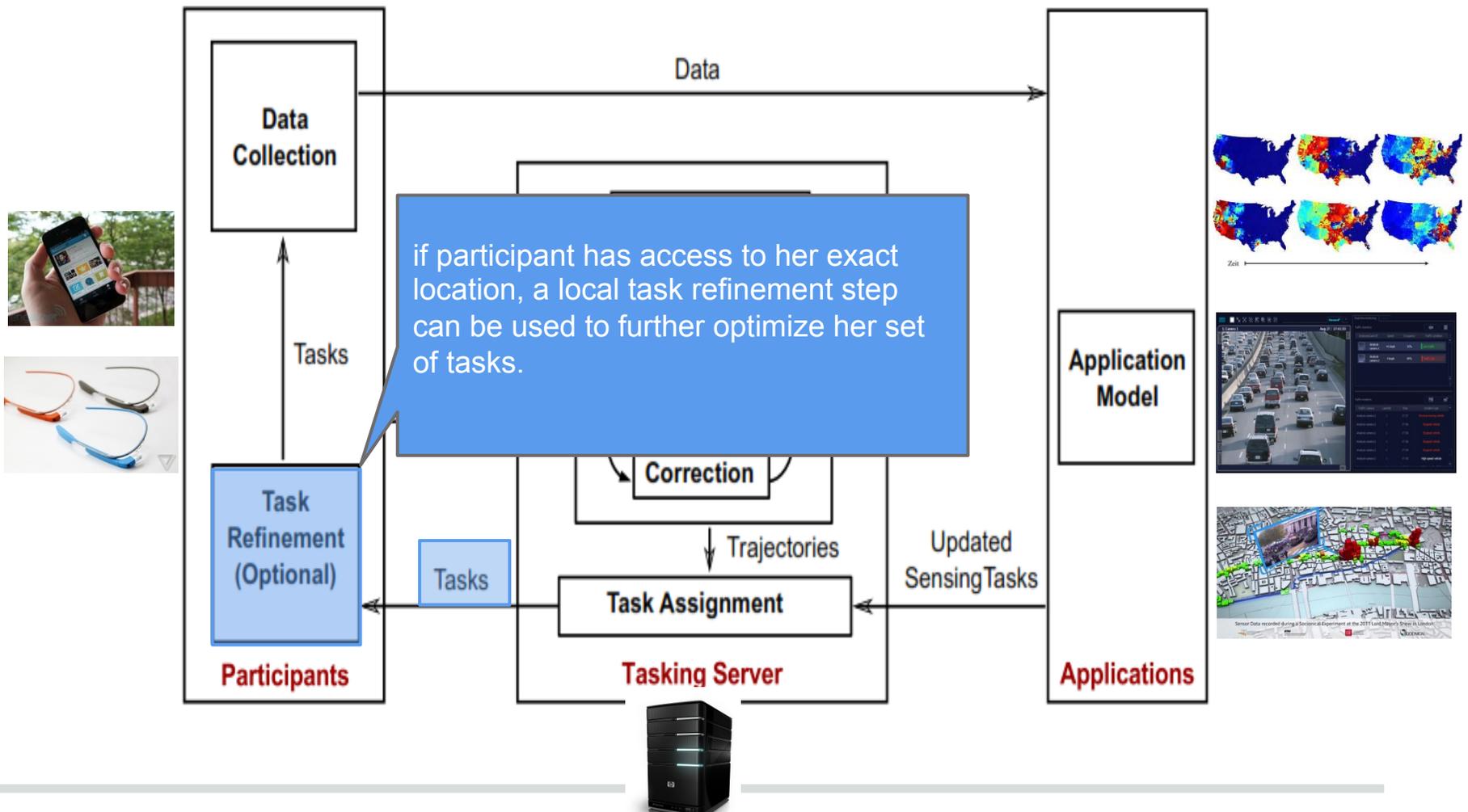


Global Task Assignment

$$\begin{aligned} & \min_x \sum_{i \in N} \sum_{j \in M} d_{i,j} x_{i,j} \\ \text{s.t.} \quad & \sum_{j \in M} \frac{\sum_{i \in N} x_{i,j}}{k_j} \geq gm \\ & \sum_{j \in M} x_{i,j} d_{i,j} \leq b_i \end{aligned}$$

- Matrix d is uncertain, hence distances should be estimated.
 - NP-Hard Problem (Minimum Set Cover Problem reduction)
 - Probabilistic greedy algorithms
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Adaptive dynamic data driven framework

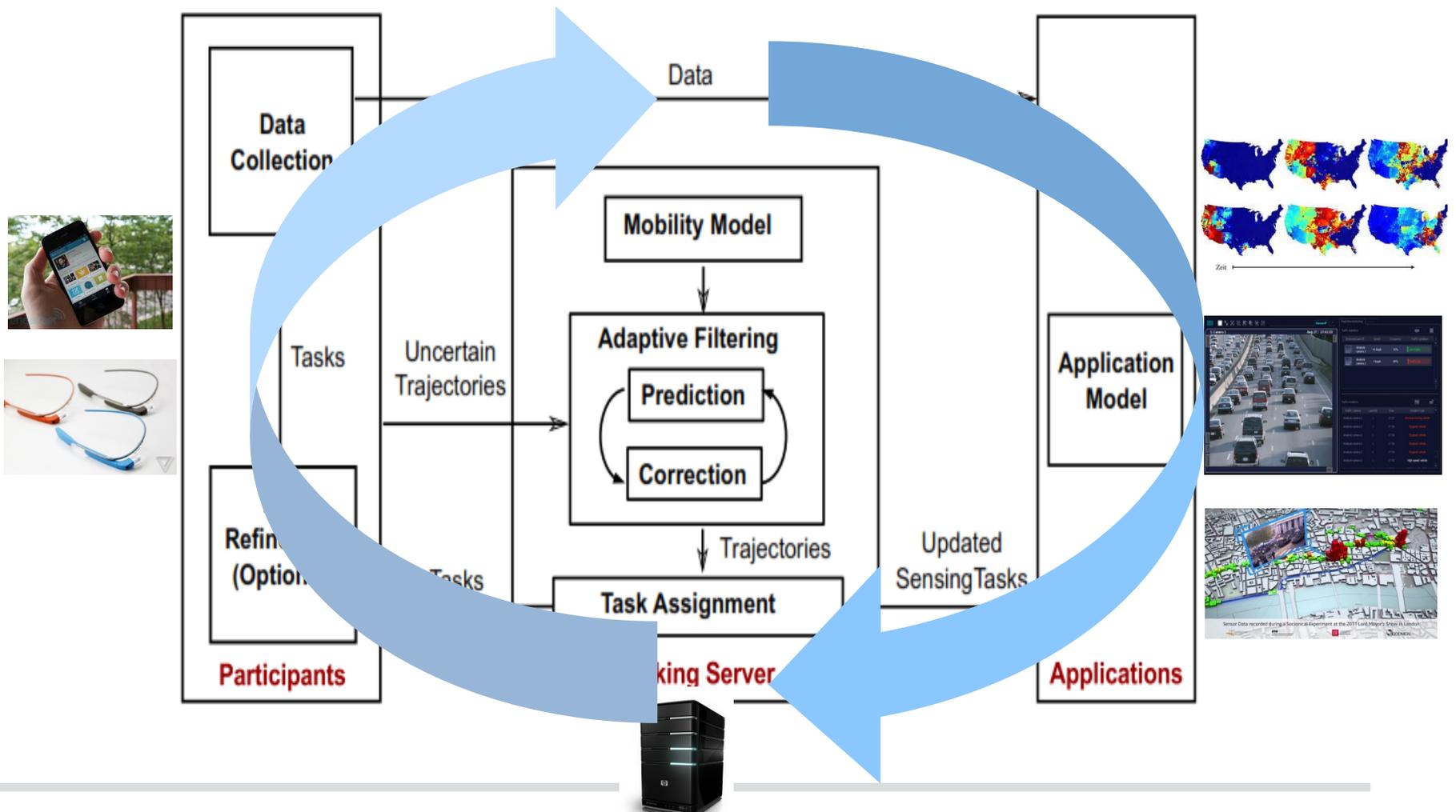


Local Task Refinement

$$\begin{aligned} & \min_y \sum_{j \in M} d_{i,j} y_{i,j} \\ \text{s.t. } & |\mathbf{y}_i - \mathbf{x}_i| < \epsilon \\ & \sum_{j \in M} \frac{y_{i,j}}{k_j} \geq \sum_{j \in M} \frac{x_{i,j}}{k_j} \\ & \sum_{j \in M} y_{i,j} d_{i,j} \leq b_i \end{aligned}$$

- Minimize distance to travel
 - To avoid over/under coverage, the difference between refined assignments (y) and global assignments (x) should be small
 - NP-Hard Problem (Minimum Set Cover Problem reduction)
 - Probabilistic greedy algorithms
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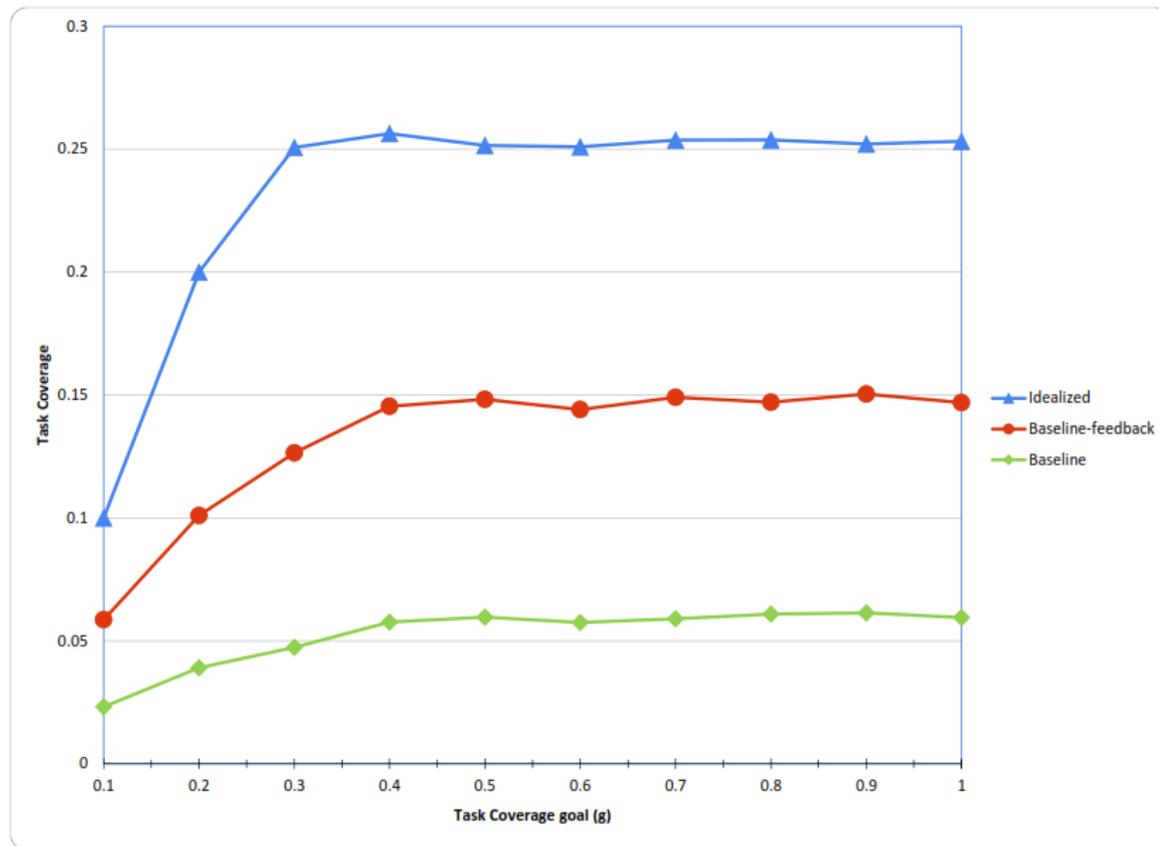
Adaptive dynamic data driven framework



Experiments : Data Sets

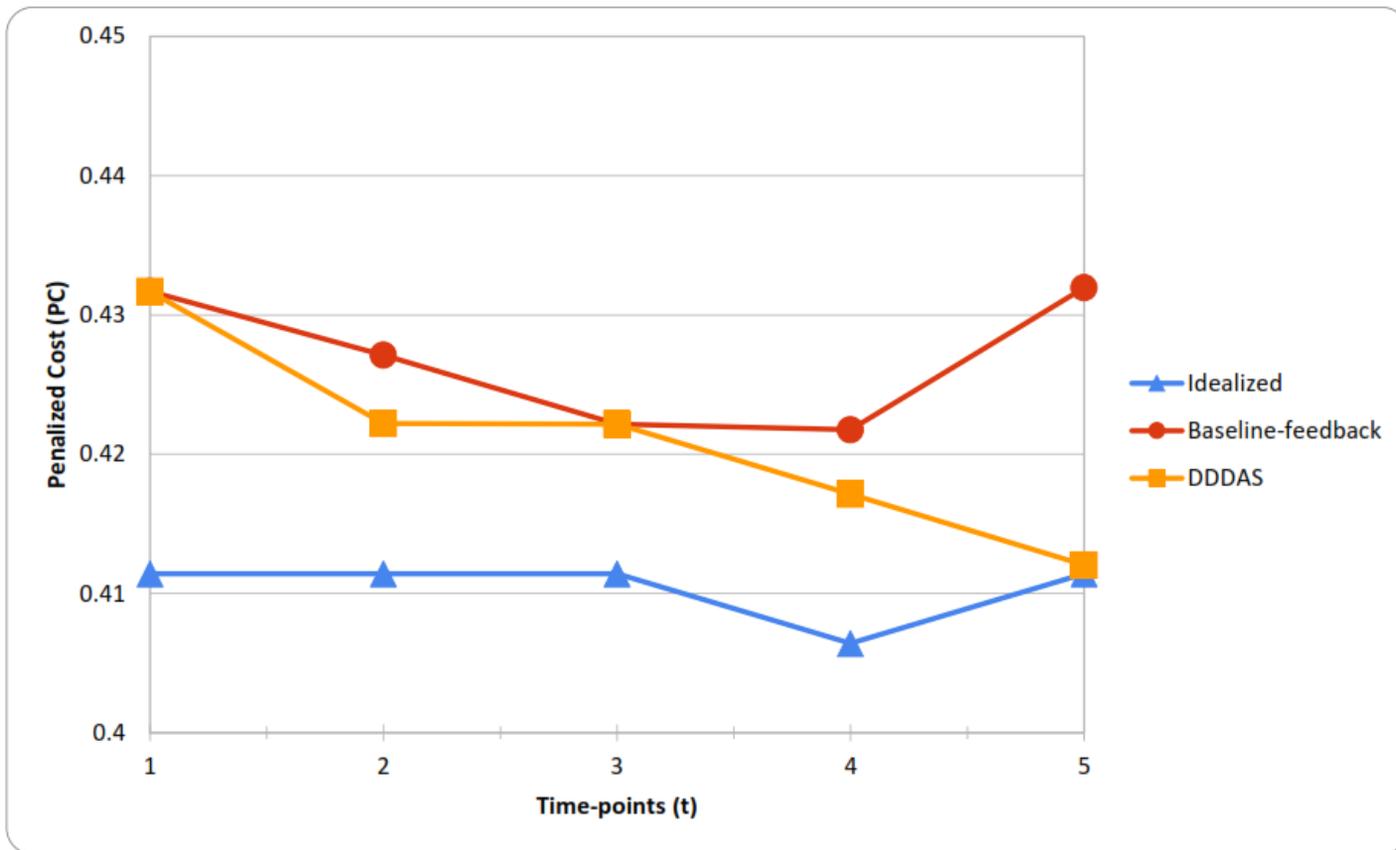
- **Gowalla:** A real-world data set which contains check-in information of users in a location-based social network
 - **Geolife GPS Trajectories:** Collected in (Microsoft Research Asia) Geolife project by 182 users in a period of over three years (from April 2007 to August 2012).
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Results



Relative task coverage for different coverage goal, $n = 200$, and $m = 200$, using datasets Gowalla

Results



Penalized cost for different timepoints, $n = 100$, and $m = 100$, using dataset Geolife

Ongoing and Future Work

- Enhanced mobility modeling and trajectory prediction algorithms for more accurate participant tracking
 - Implementing mobile app and server software for crowd sensing with dynamic task assignment
 - Formal location perturbation mechanisms
 - Trustworthiness of participants in terms of the quality of the data contributed by them
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