

Linking people in videos with “their” names using coreference resolution

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Video Data



Video Data

+ Text Data



TV Script

Movie Script



Closed Caption



Video Data

+ Text Data



TV Script

Use this data to label people!



Movie Script

Closed Caption



Video Data

+ Text Data



TV Script

Names

Movie Script

Closed Caption



Video Data

+ Text Data



Everingham et al. 2006
Sivic et al. 2009
Cour et al. 2009
Tapaswi et al. 2012

TV Script

Laptev et al. 2000
Marszalek et al. 2009
Bojanowski et al. 2013

Movie Script

Names
Actions
Interactions
Events

Babaguchi et al. 2002
Xu et al. 2008

Closed Caption



Video Data

+ Text Data



Everingham et al. 2006
Sivic et al. 2009
Cour et al. 2009
Tapaswi et al. 2012

TV Script

Names

Actions

Interactions

Events

Babaguchi et al. 2002
Xu et al. 2008

Closed Caption

Laptev et al. 2000
Marszalek et al. 2009
Bojanowski et al. 2013

Movie Script



Text to Video



Roland arrives as Ian waits. He stands with Mary.
She turns.

Challenging Setting

- No labelled instances
- Script is the only supervision

Unidirectional Models: **Text to Video**



Roland arrives as Ian waits. He stands with Mary.
She turns.

Everingham et al. 2006
Sivic et al. 2009
Cour et al. 2009
Tapaswi et al. 2012
Bojanowski et al. '13

Unidirectional Models: Text to Video



Roland arrives as Ian waits. He stands with Mary.
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Everingham et al. 2006
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Cour et al. 2009
Tapaswi et al. 2012
Bojanowski et al. '13

Use Limited Information

Only proper nouns

Unidirectional Models: **Text to Video**

Video = Bag of tracks



Script = Bag of names



Everingham et al. 2006

Sivic et al. 2009

Cour et al. 2009

Tapaswi et al. 2012

Bojanowski et al. '13

Use Limited Information

Only proper nouns

Treat Videos and Script as
bag of tracks and names

Unidirectional Models: **Text to Video**

Unidirectional Models

Only proper nouns

Treat Videos and Script as
bag of tracks and names

Temporal ordering of people



Roland arrives as Ian waits. He stands with Mary. She turns

Time

Temporal ordering of people

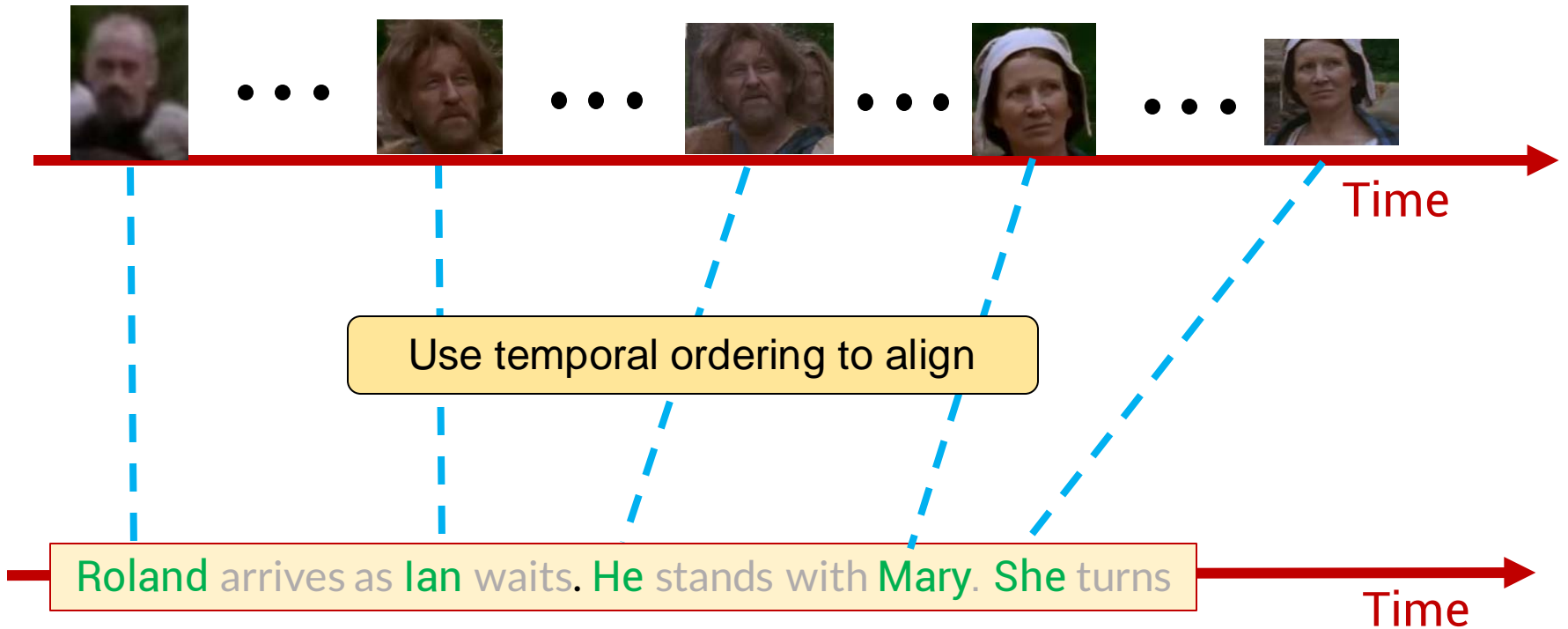


People appear in the same order in both text and videos

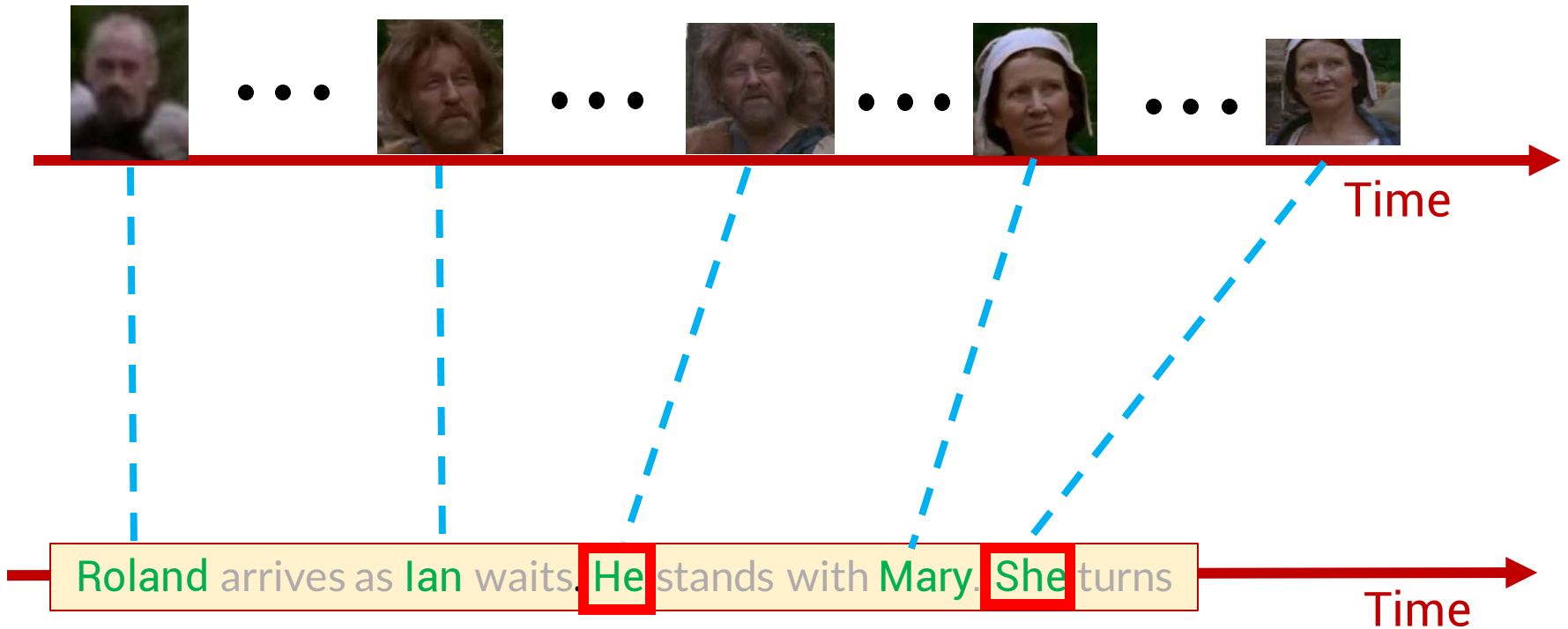
Roland arrives as Ian waits. He stands with Mary. She turns

Time

Temporal ordering of people

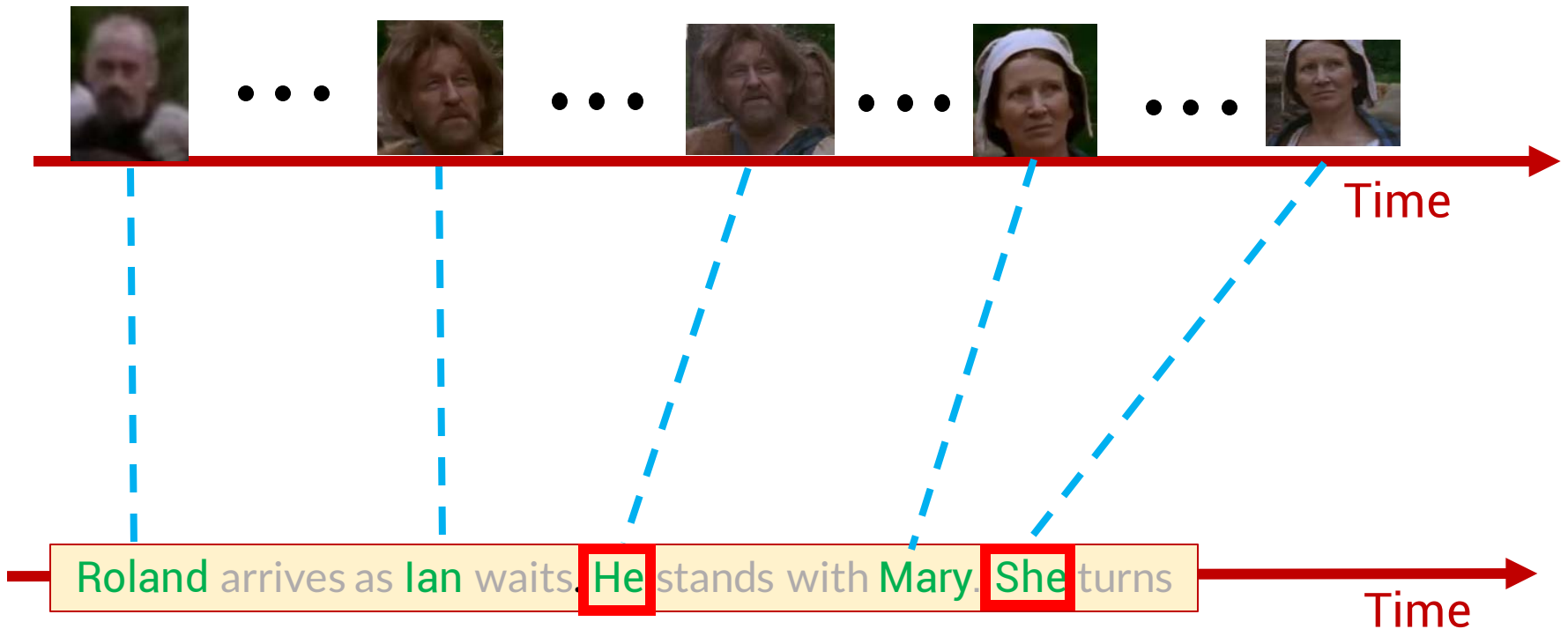


Coreference Resolution in NLP



To use the alignment we need to
identify the name of pronouns

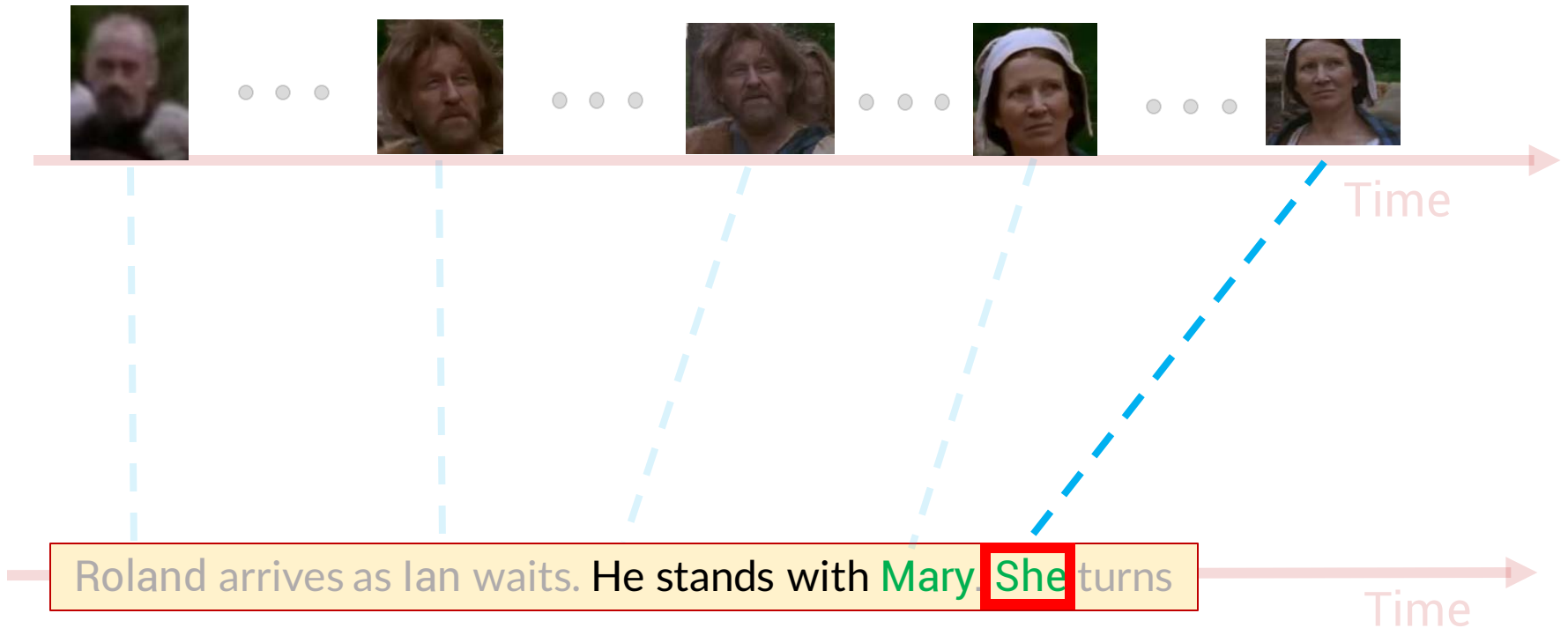
Coreference Resolution in NLP



To use the alignment we need to
identify the name of pronouns

Coreference Resolution

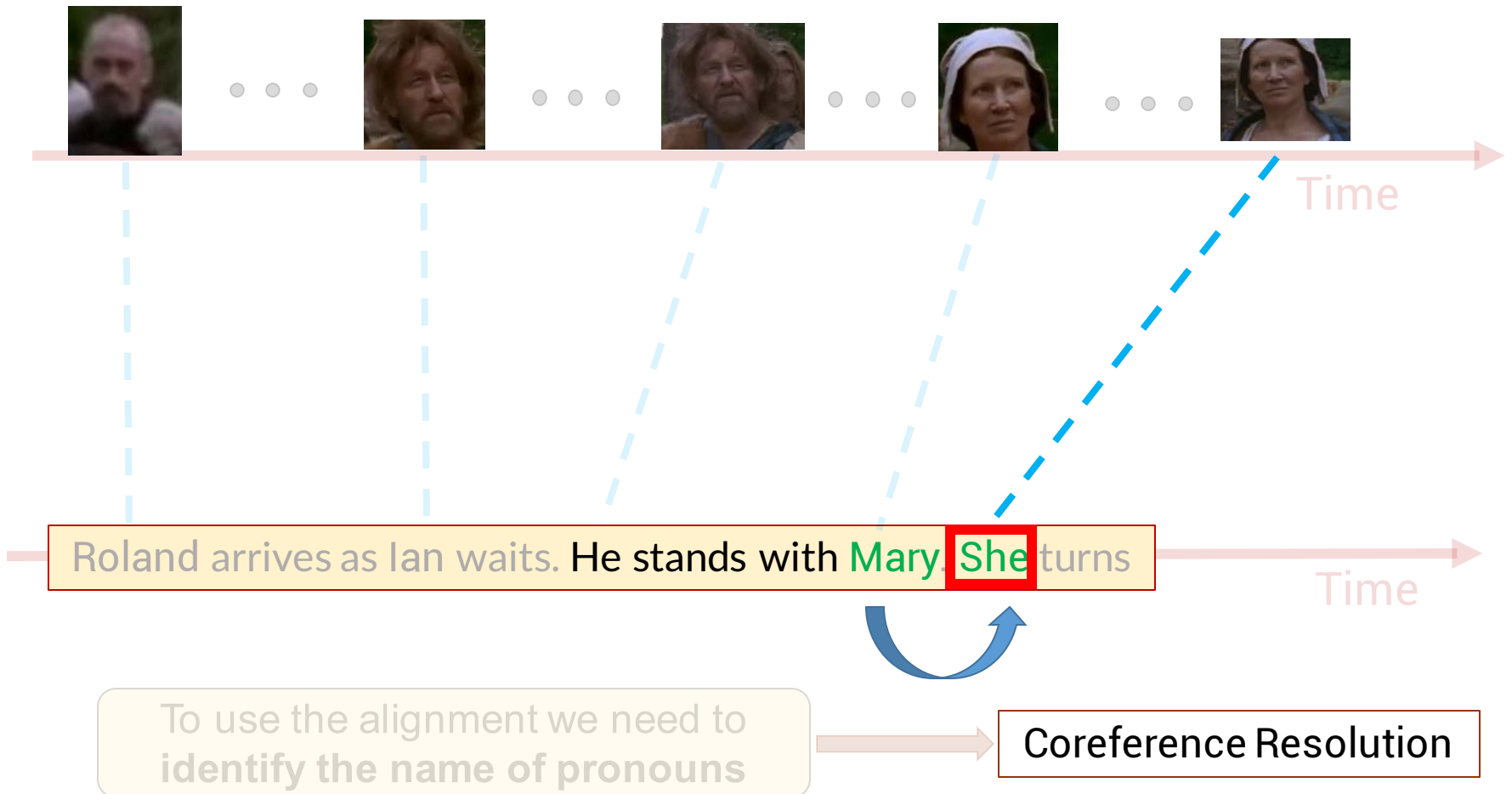
Coreference Resolution in NLP



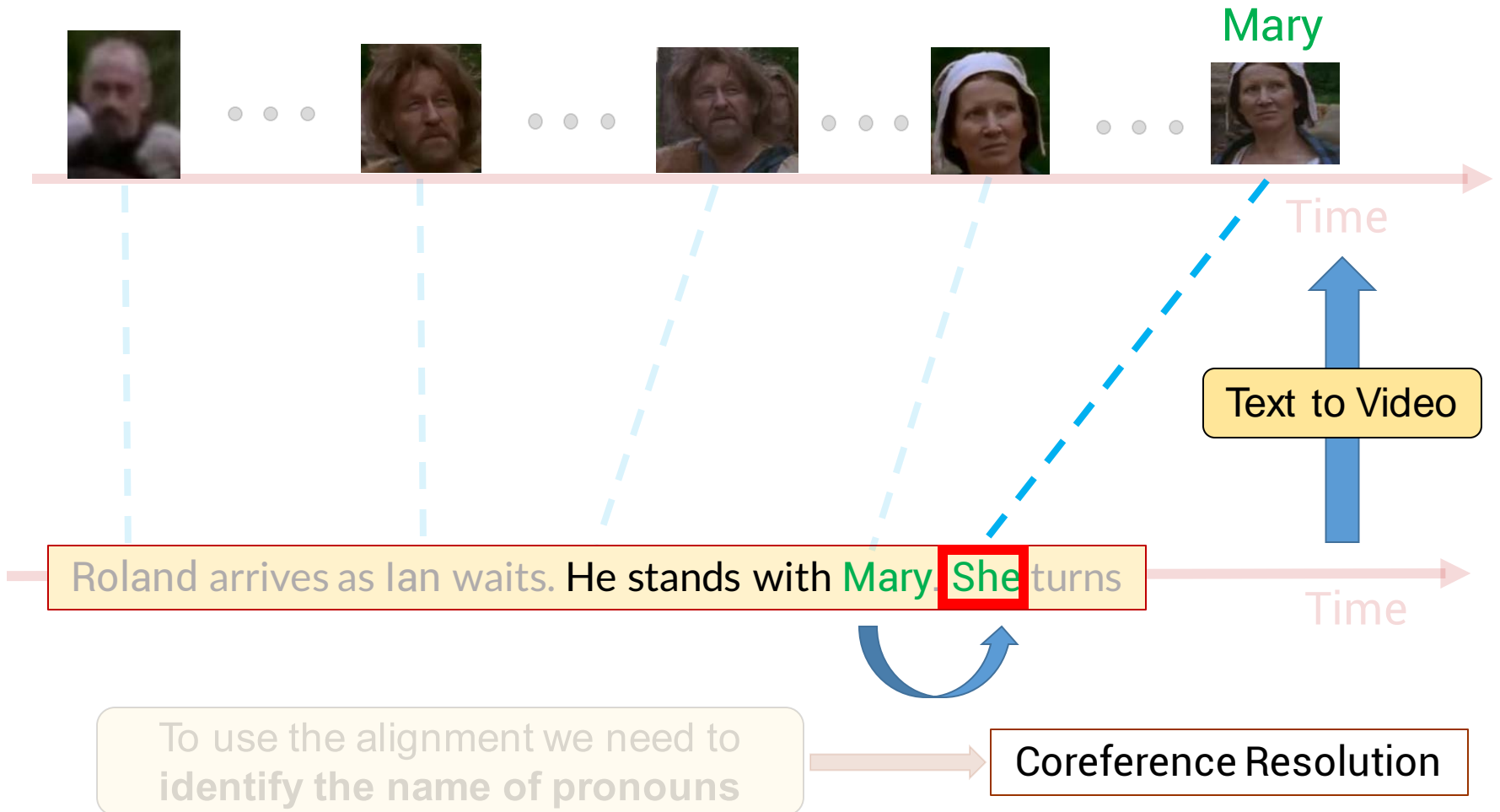
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Coreference Resolution

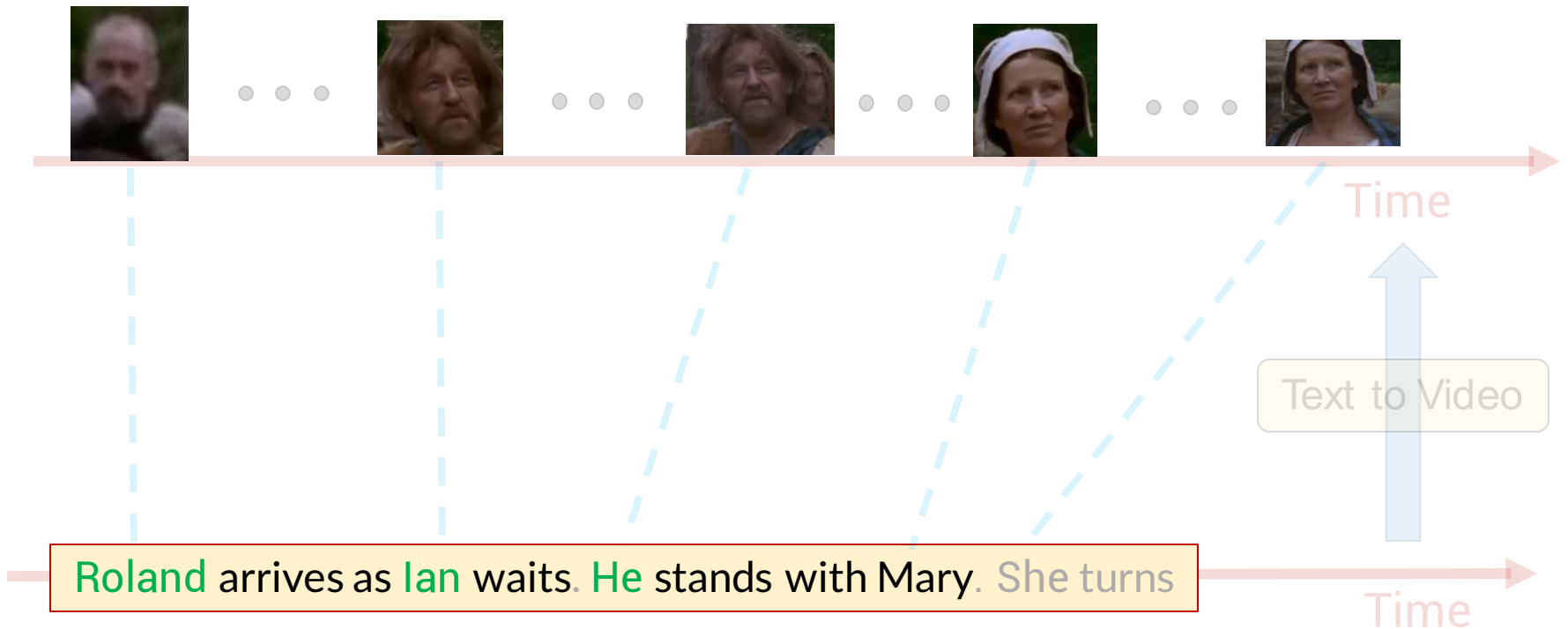
Coreference Resolution in NLP



Coreference Resolution in NLP



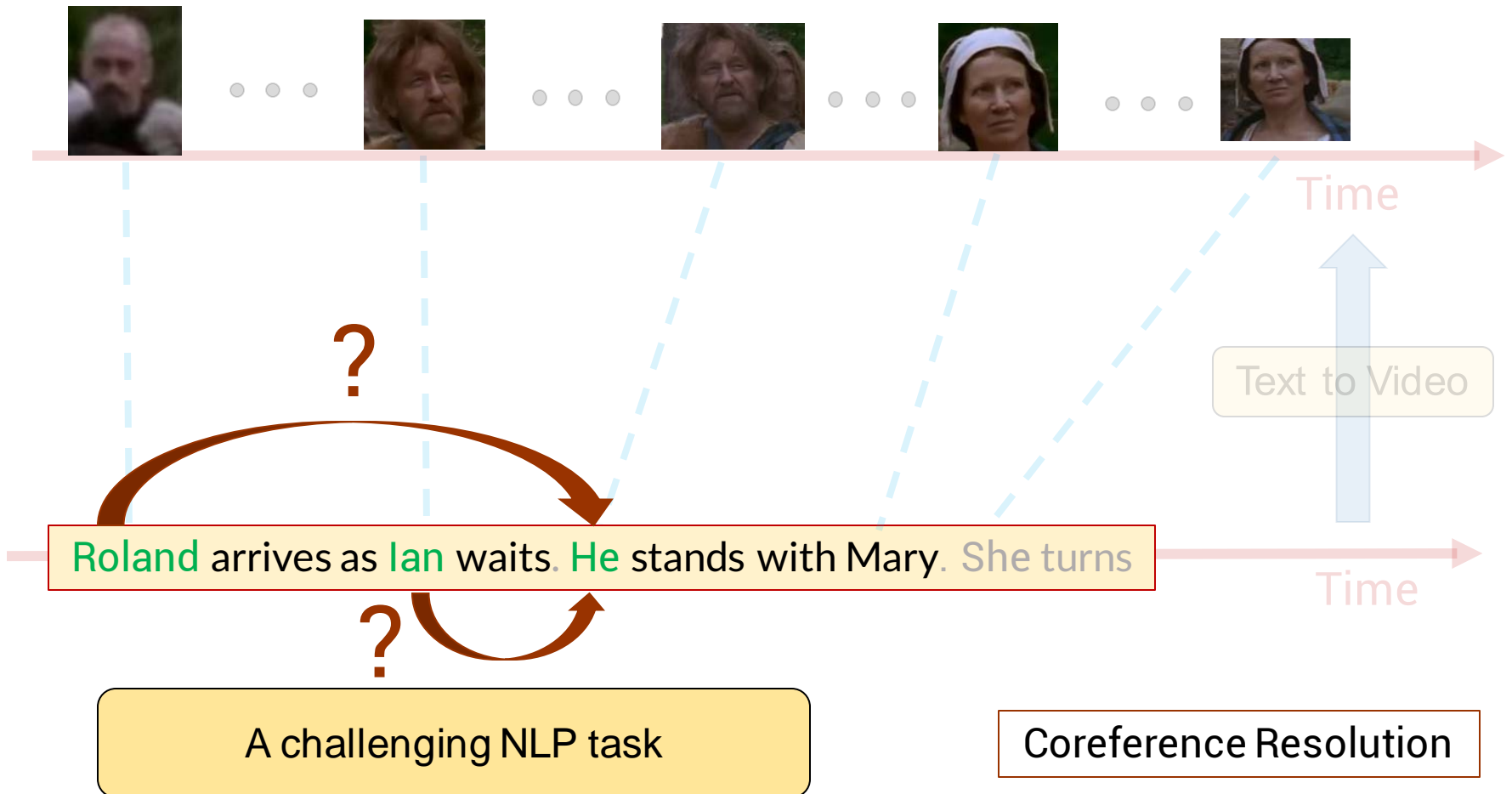
Coreference Resolution in NLP



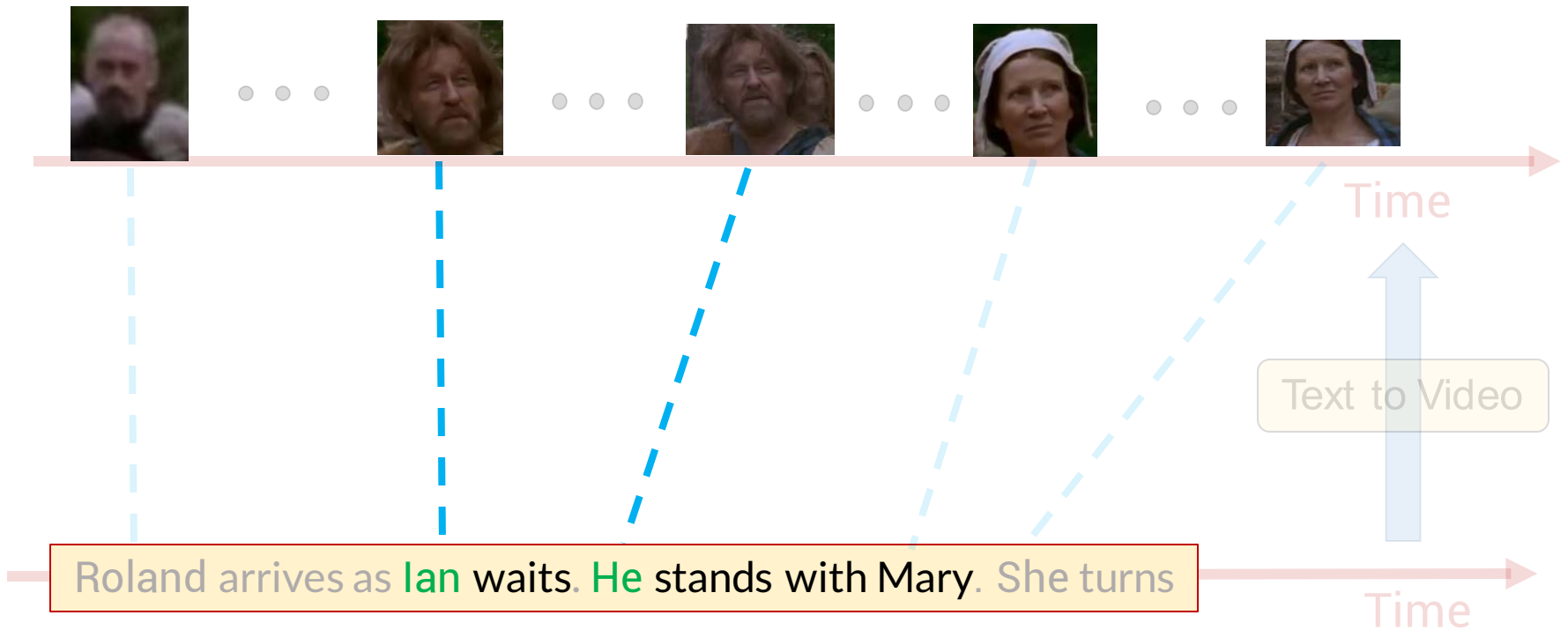
A challenging NLP task

Coreference Resolution

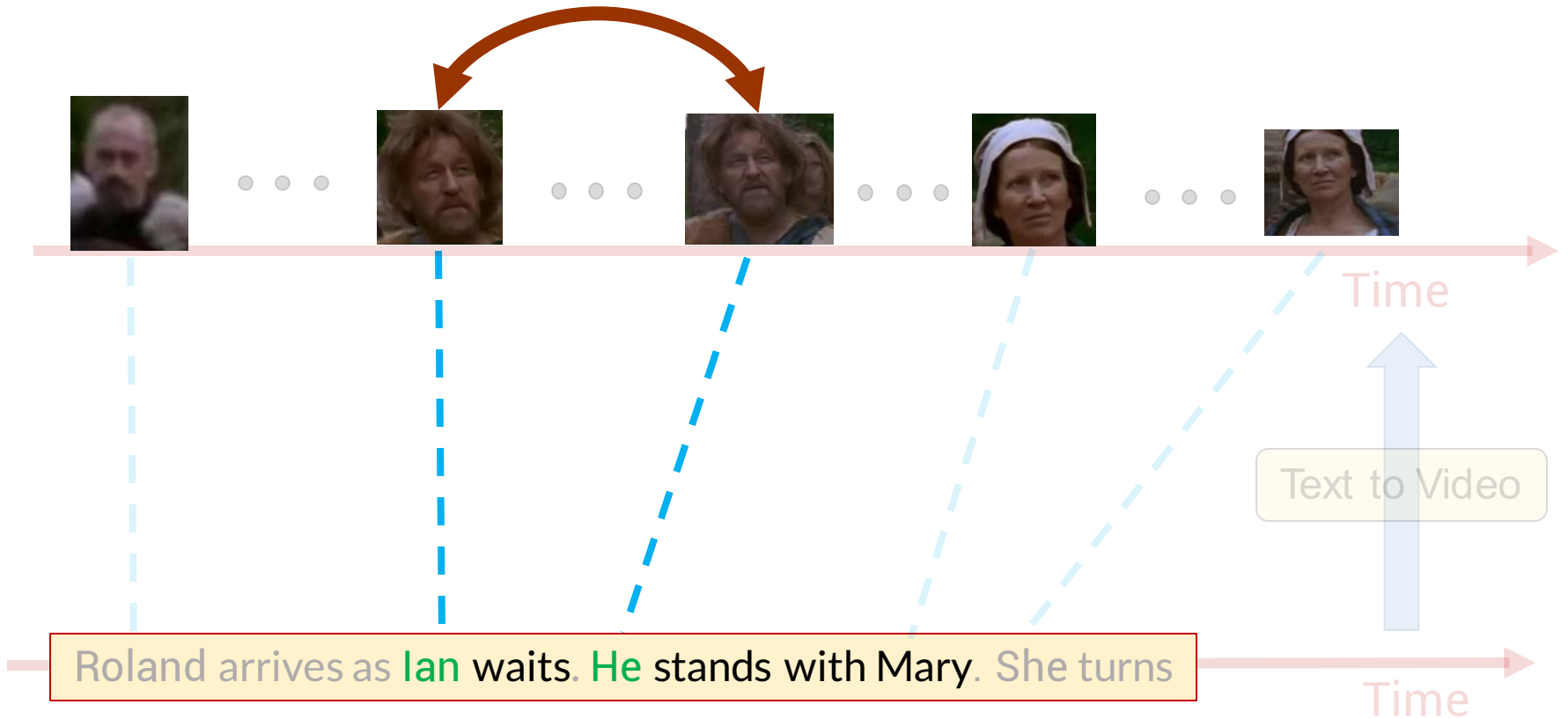
Coreference Resolution in NLP



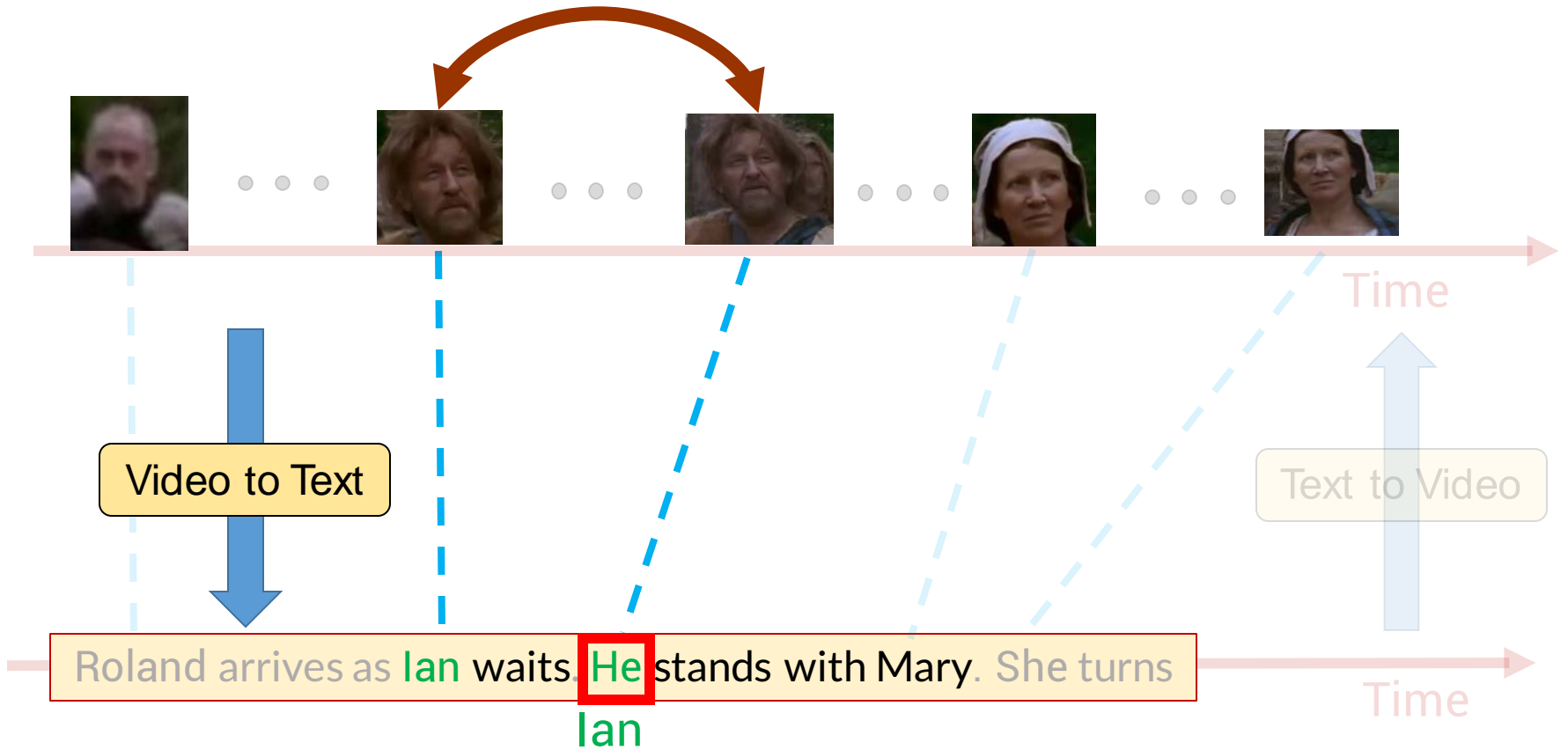
Coreference Resolution in NLP



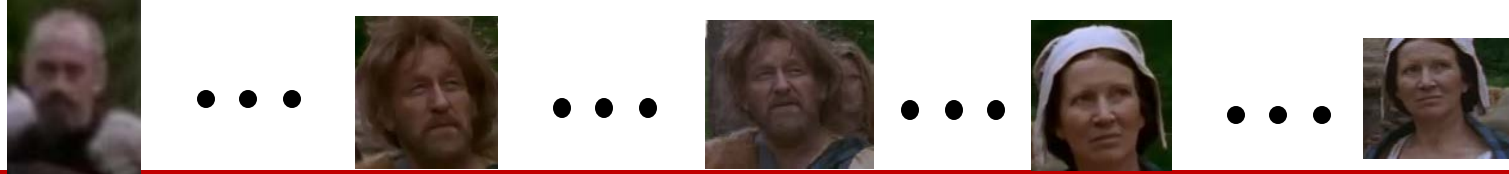
Coreference Resolution in NLP



Coreference Resolution in NLP



Bidirectional model



Video to Text

Our Bidirectional Model

Text to Video

Roland arrives as Ian waits. He stands with Mary. She turns

Linking people with coreference resolution

Everingham et al. 2006
Sivic et al. 2009
Cour et al. 2009
Tapaswi et al. 2012
Bojanowski et al. '13

Our Bidirectional Model

Joint **coreference resolution** to use pronouns and nominals

Treat as **temporal sequences** to align video and text

Unidirectional Models

Only proper nouns

Treat Videos and Script as bag of tracks and names

Linking people with “their” names

- Problem setup
- Our Bidirectional model
- Experiments
- Summary

Problem Setup (Input)

- Videos with detected **human tracks**



...



...

time



Problem Setup (Input)

- Videos with detected **human tracks**
- Descriptive script partially aligned with video segments



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Roland arrives as Ian waits. He stands with Mary

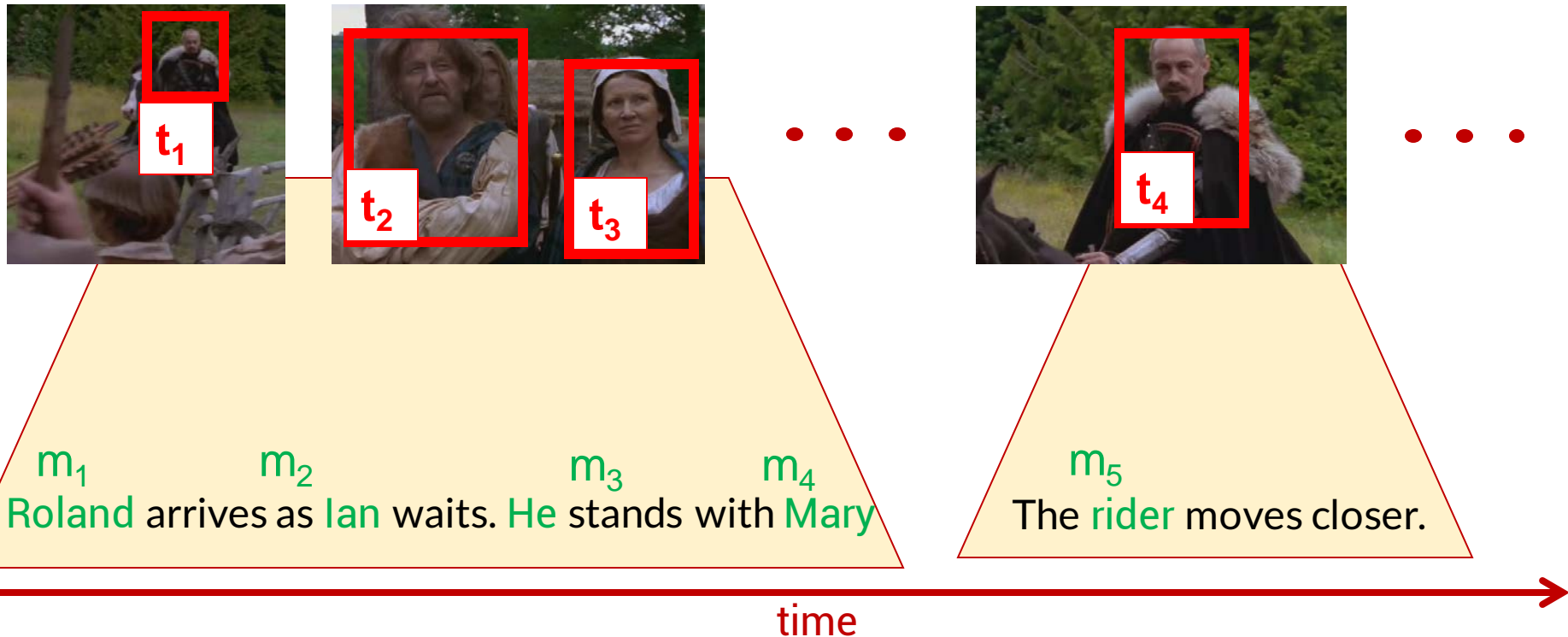
The rider moves closer.

time



Problem Setup (Input)

- Videos with detected **human tracks**
- Descriptive script partially aligned with video segments
- Human **mentions (nouns, pronouns)** identified in the script.



Problem Setup (Input)

- Videos with detected **human tracks**
- Descriptive script partially aligned with video segments
- Human **mentions (nouns, pronouns)** identified in the script.
- A list of cast names

Cast List: Roland, Ian, Mary, ...



...



...

m_1 m_2 m_3 m_4
Roland arrives as Ian waits. He stands with Mary

m_5
The rider moves closer.

time



Problem Setup (Output)



...



...

m_1

m_2

m_3

m_4

Roland arrives as Ian waits. He stands with Mary ...

m_5

The rider moves closer.

Problem Setup (Output)

Name assignment to human tracks



t_1

Roland



t_2

Ian

t_3

Mary

...



t_4

Roland

...

m_1

m_2

m_3

m_4

Roland arrives as Ian waits. He stands with Mary ...

m_5

The rider moves closer.

Problem Setup (Output)

Name assignment to human tracks



○ ○ ○



○ ○ ○

m_1

m_2

m_3

m_4

Roland arrives as Ian waits. He stands with Mary ...

m_4

The rider moves closer.

Problem Setup (Output)

Name assignment to human tracks

Name assignment to human mentions



...



...

Roland

Ian

Ian

Mary

Roland

m_1

m_2

m_3

m_4

m_4

Roland arrives as Ian waits. He stands with Mary ...

The rider moves closer.

Linking people with “their” names

- Problem setup
- **Our Bidirectional model**
- Experiments
- Summary

Our Bidirectional Model

Minimize joint cost of name assignment to tracks and mentions with alignment



Our Bidirectional Model

Minimize joint cost of name assignment to tracks and mentions with alignment



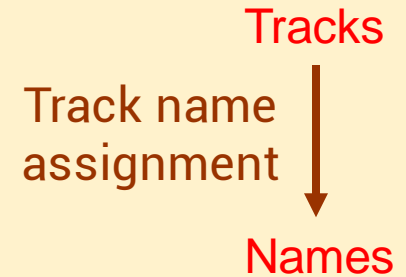
Roland



Ian

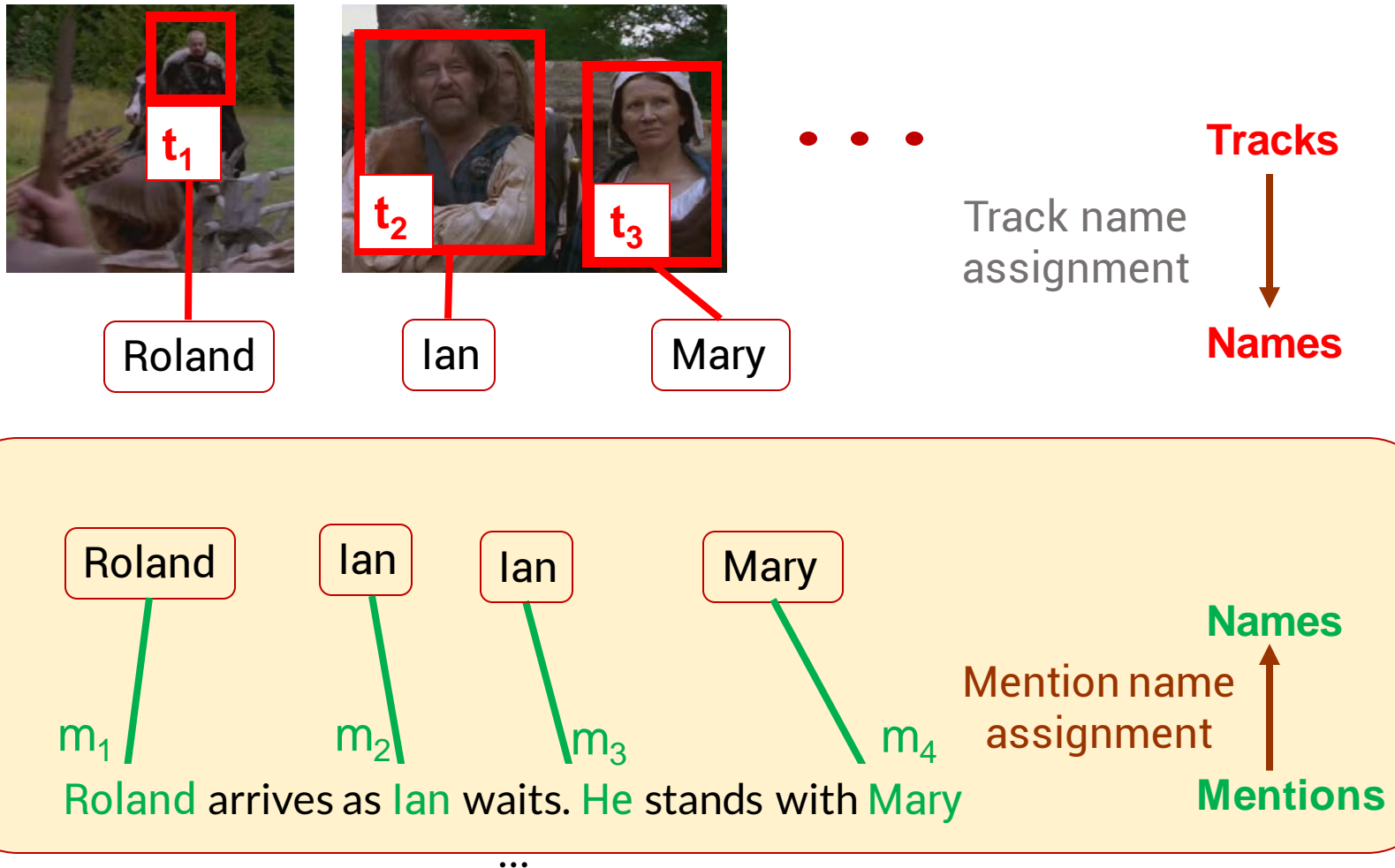
Mary

...



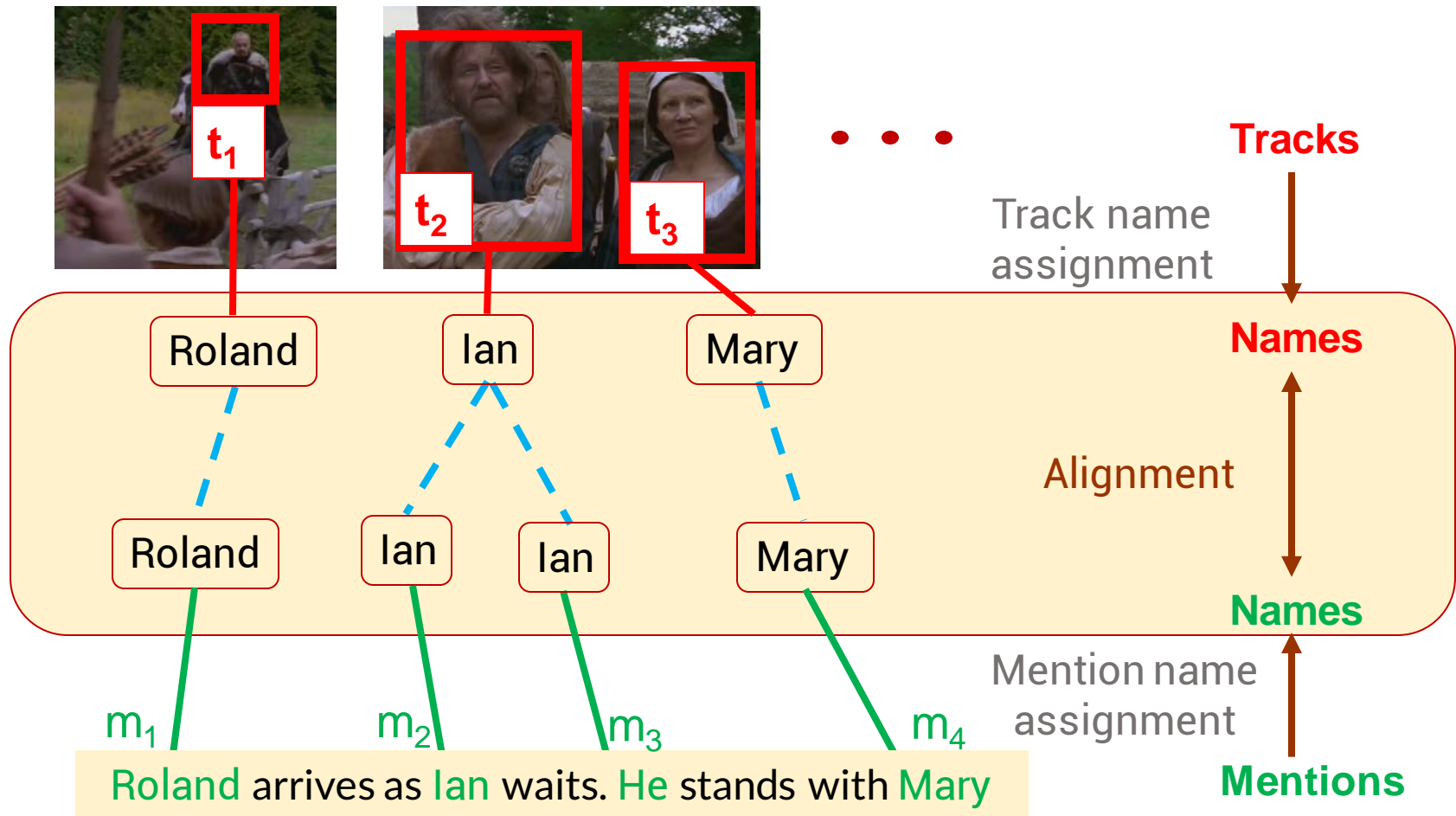
Our Bidirectional Model

Minimize joint cost of name assignment to tracks and mentions with alignment



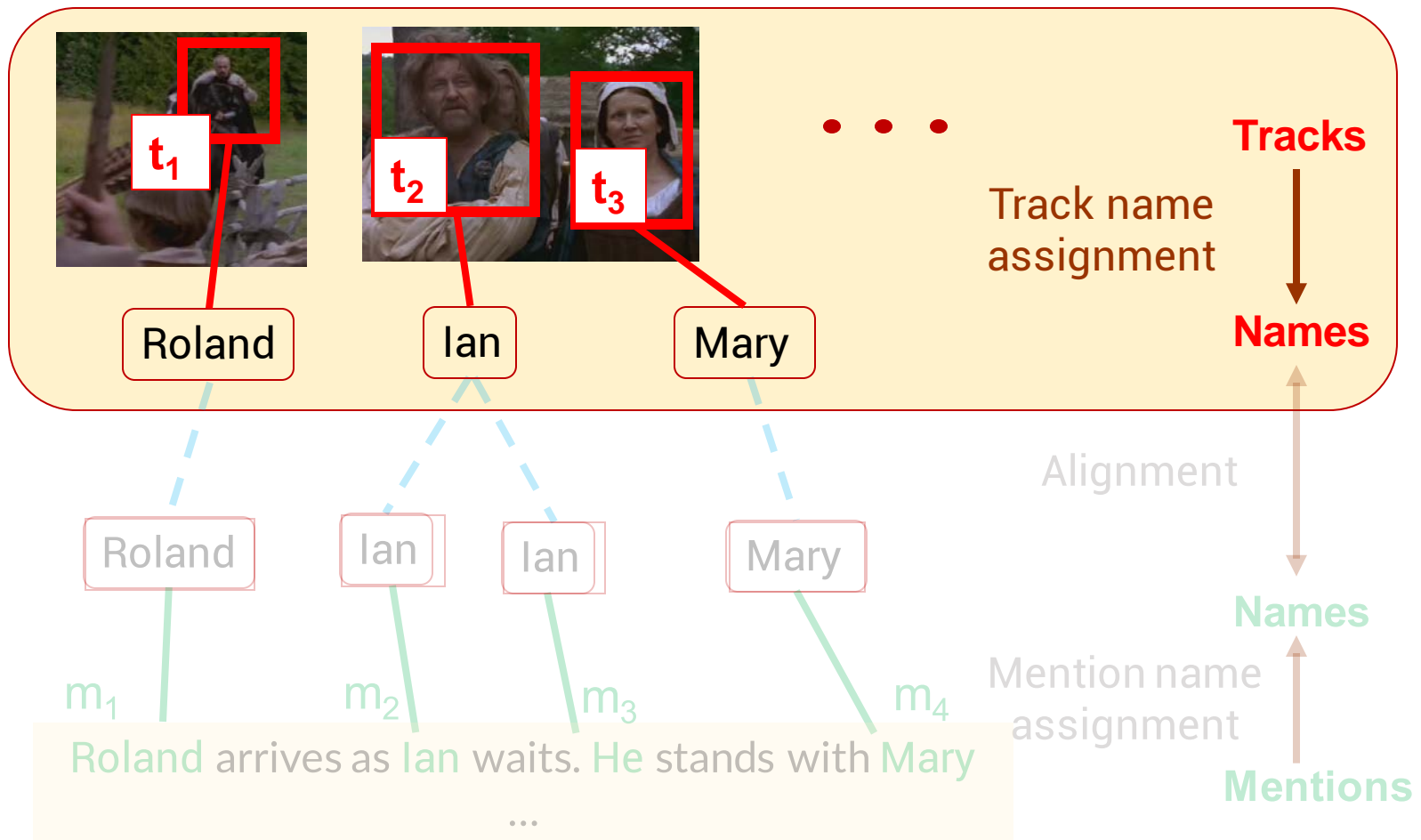
Our Bidirectional Model

Minimize joint cost of name assignment to tracks and mentions with alignment



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Our Bidirectional Model: Track name assignment



Our Bidirectional Model: Track name assignment

$\mathbf{Y} \in \{0,1\}^{T \times P}$ Name assignment for tracks



...



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Our Bidirectional Model: Track name assignment

$\mathbf{Y} \in \{0,1\}^{T \times P}$ Name assignment for tracks

Roland	Ian	Mary
1	0	0



Roland



...

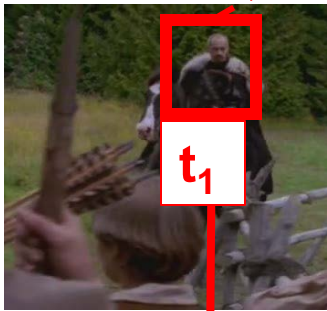


...

Our Bidirectional Model: Track name assignment

$\mathbf{Y} \in \{0,1\}^{T \times P}$ Name assignment for tracks

	Roland	Ian	Mary
t_1	1	0	0
t_2	0	1	0



Roland



Ian

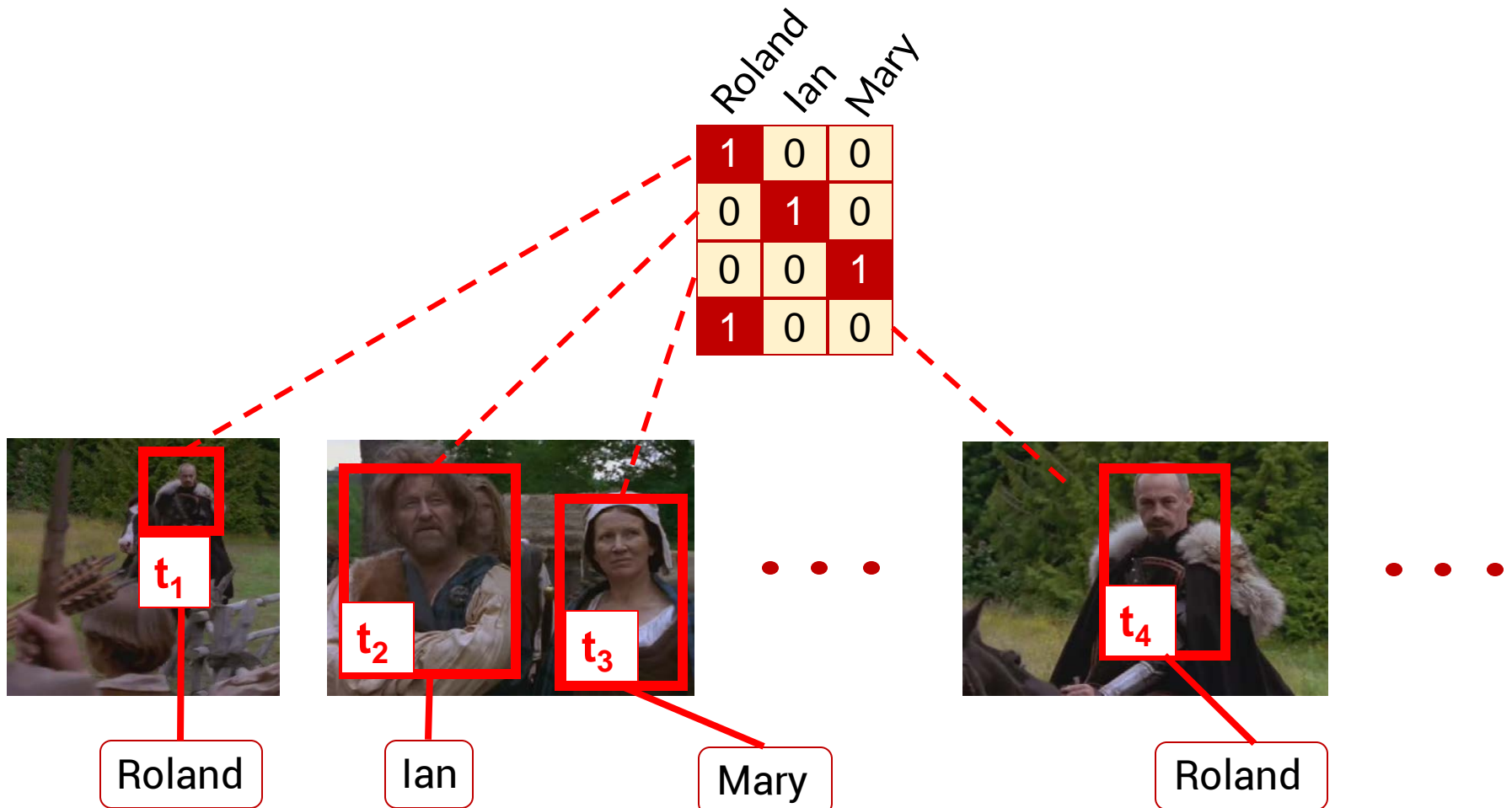
...



...

Our Bidirectional Model: Track name assignment

$\mathbf{Y} \in \{0,1\}^{T \times P}$ Name assignment for tracks



Our Bidirectional Model: Track name assignment

$\mathbf{Y} \in \{0,1\}^{T \times P}$ Name assignment for tracks



...



...

Our Bidirectional Model: Track name assignment

$\mathbf{Y} \in \{0,1\}^{T \times P}$ Name assignment for tracks

Clustering cost*

$$\text{tr}\left(\mathbf{Y}^T \Pi_{\text{track}} \mathbf{Y}\right)$$

Convex quadratic in \mathbf{Y}

- Discriminative clustering.
- Tracks with **similar features** should have **same name**.

*Bojanowski et al. ICCV'13



...



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Our Bidirectional Model: Track name assignment

$\mathbf{Y} \in \{0,1\}^{T \times P}$ Name assignment for tracks

Clustering cost*

$$\text{tr}(\mathbf{Y}^T \mathbf{\Pi}_{\text{track}} \mathbf{Y})$$

Convex quadratic in \mathbf{Y}

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...



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Our Bidirectional Model: Track name assignment

$\mathbf{Y} \in \{0,1\}^{T \times P}$ Name assignment for tracks

Clustering cost

$$\text{tr}(\mathbf{Y}^T \mathbf{\Pi}_{\text{track}} \mathbf{Y})$$

Convex quadratic in \mathbf{Y}

Scene constraint

Rol. Ian Mary

Scene (Ian,
Mary)



...



...

Our Bidirectional Model: Track name assignment

$\mathbf{Y} \in \{0,1\}^{T \times P}$ Name assignment for tracks

Clustering cost

$$\text{tr}(\mathbf{Y}^T \mathbf{\Pi}_{\text{track}} \mathbf{Y})$$

Convex quadratic in \mathbf{Y}

Scene constraint

Rol.	Ian	Mary
X		
X		

The tracks are not “Roland”

Scene (Ian,
Mary)



...



...

Our Bidirectional Model: Track name assignment

$\mathbf{Y} \in \{0,1\}^{T \times P}$ Name assignment for tracks

Clustering cost

$$\text{tr}(\mathbf{Y}^T \Pi_{\text{track}} \mathbf{Y})$$

Convex quadratic in \mathbf{Y}

Scene constraint

$$\sum_{\substack{t \in \text{scene} \\ p \notin \text{scene}}} \mathbf{Y}_{t,p} = 0$$

Linear constraint



...

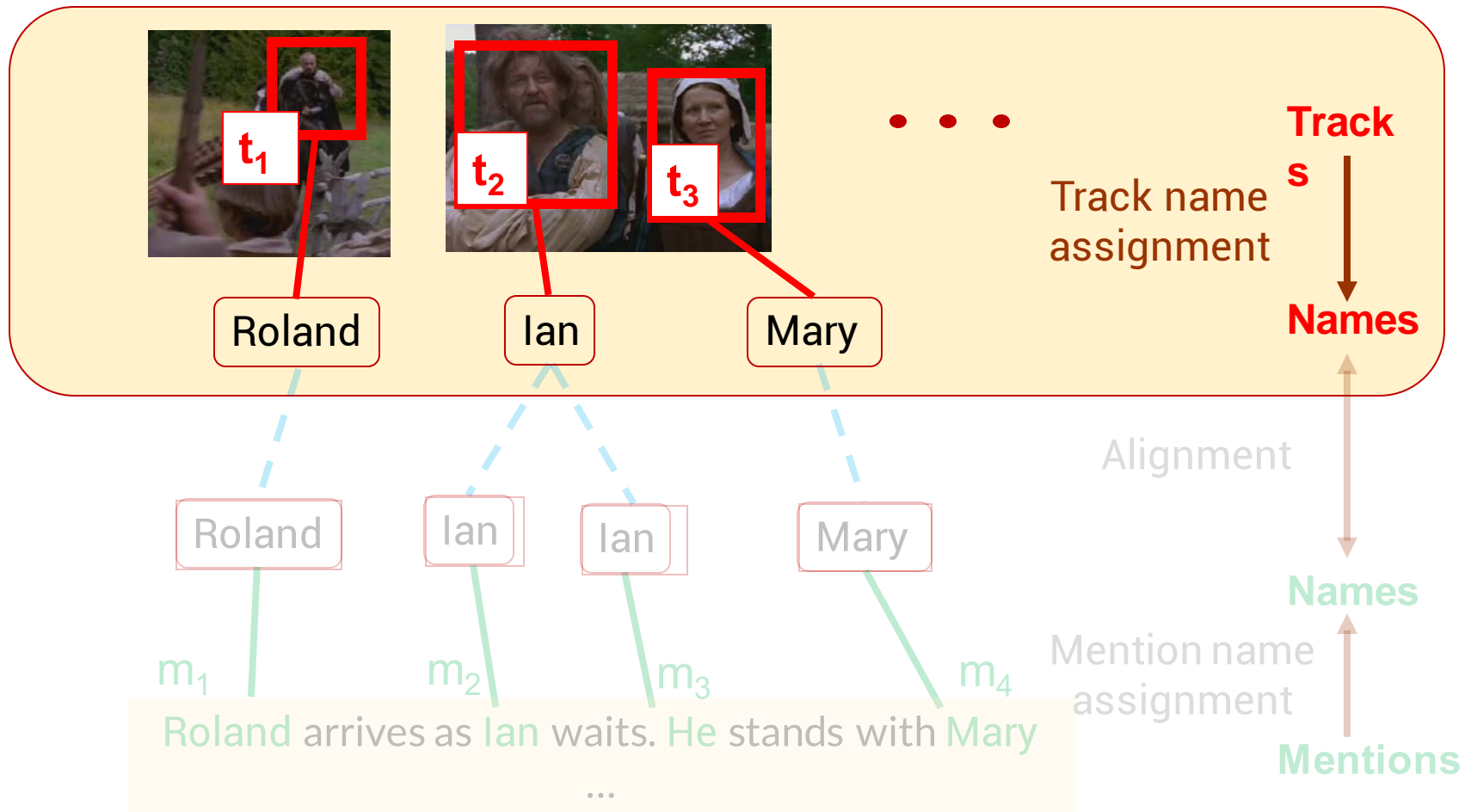


...

Our Bidirectional Model

$$\min \operatorname{tr}(\mathbf{Y}^T \Pi_{\text{track}} \mathbf{Y})$$

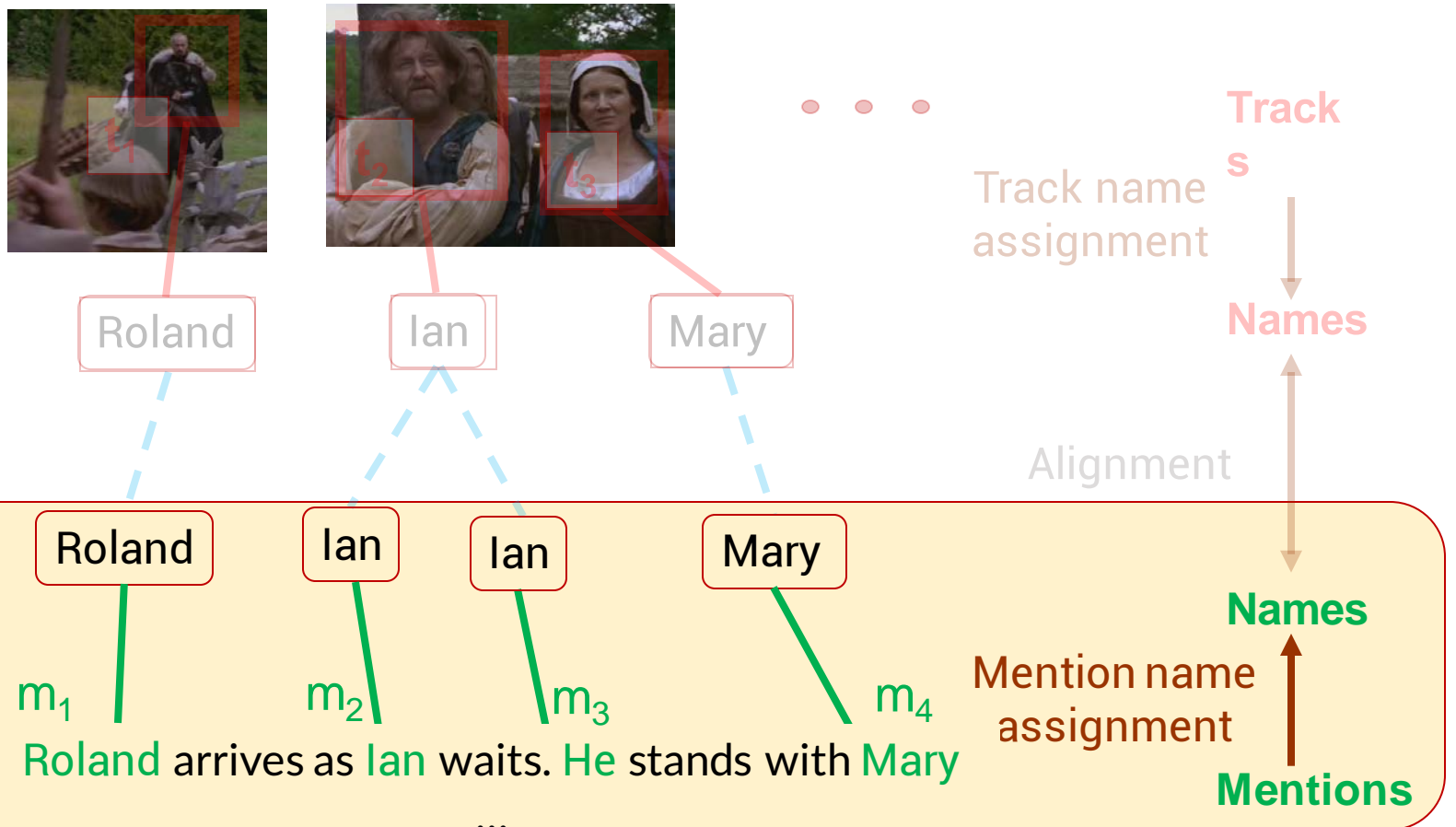
$$\text{s.t. } \mathbf{Y} \in \mathcal{C}_Y$$



Our Bidirectional Model

$$\min \operatorname{tr}(\mathbf{Y}^T \Pi_{\text{track}} \mathbf{Y})$$

$$\text{s.t. } \mathbf{Y} \in \mathcal{C}_Y$$



Our Bidirectional Model: **Mention name assignment**

$\mathbf{Z} \in \{0,1\}^{M \times P}$ Name assignment for mentions

m_1

m_2

m_3

m_4

Roland arrives as Ian waits. He stands with Mary ...

m_5

The rider moves closer.

Our Bidirectional Model: **Mention name assignment**

$\mathbf{Z} \in \{0,1\}^{M \times P}$ Name assignment for mentions

	Roland	Ian	Mary
m_1	1	0	0
m_2	0	1	0
m_3	0	1	0
m_4	0	0	1
m_5	1	0	0

m_1 Roland
 m_2 Ian
 m_3 Ian
 m_4 Mary

Roland arrives as Ian waits. He stands with Mary ...

m_5 Roland

The rider moves closer.

Our Bidirectional Model: **Mention name assignment**

$\mathbf{Z} \in \{0,1\}^{M \times P}$ Name assignment for mentions

Mentions independently are not informative!

m_3

Roland arrives as Ian waits. **He** stands with Mary ...

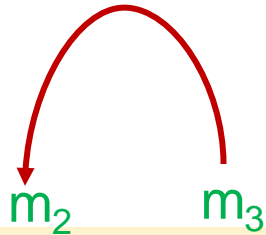
The rider moves closer.

Our Bidirectional Model: **Mention name assignment**

$\mathbf{Z} \in \{0,1\}^{M \times P}$ Name assignment for mentions

We rely on context to identify the mention-name.

Name of an ambiguous mention depends on an antecedent mention



Roland arrives as **lan** waits. **He** stands with Mary ...

The rider moves closer.

Our Bidirectional Model: **Mention name assignment**

$\mathbf{Z} \in \{0,1\}^{M \times P}$ Name assignment for mentions

$\mathbf{R} \in \{0,1\}^{M^2}$ Antecedence variable

m_1

m_2

m_3

m_4

m_5

Roland arrives as Ian waits. He stands with Mary ...

The rider moves closer.

Our Bidirectional Model: **Mention name assignment**

$\mathbf{Z} \in \{0,1\}^{M \times P}$ Name assignment for mentions

$\mathbf{R} \in \{0,1\}^{M^2}$ Antecedence variable

Every pair of mentions has a **text feature vector**, which determines antecedence.

m_1

m_2

m_3

m_4

m_5

Roland arrives as Ian waits. He stands with Mary ...

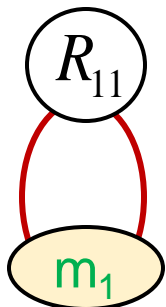
The rider moves closer.

Our Bidirectional Model: **Mention name assignment**

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Roland arrives as Ian waits. He stands with Mary ...

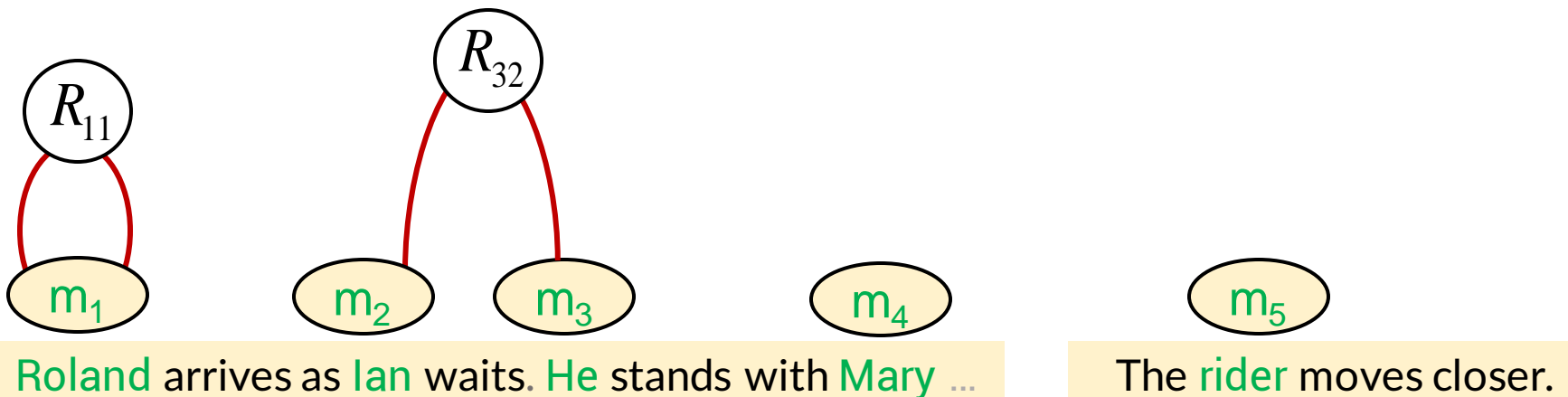
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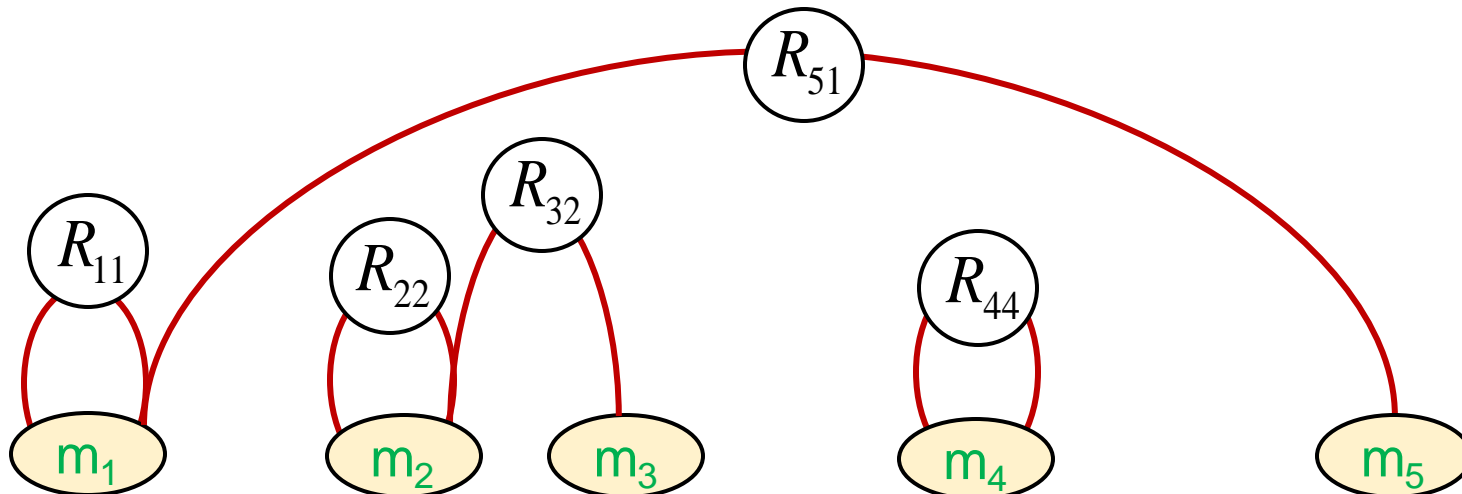


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Every pair of mentions has a **text feature vector**, which determines antecedence.



Roland arrives as Ian waits. He stands with Mary ...

The rider moves closer.

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m_1

m_2

m_3

m_4

m_5

Roland arrives as Ian waits. He stands with Mary ...

The rider moves closer.

Our Bidirectional Model: **Mention name assignment**

$\mathbf{Z} \in \{0,1\}^{M \times P}$ Name assignment for mentions

$\mathbf{R} \in \{0,1\}^{M^2}$ Antecedence variable

Antecedent clustering cost

$$\mathbf{R}^T \Pi_{men} \mathbf{R}$$

Convex quadratic in \mathbf{R}

- Discriminative clustering
- Mention-pairs with **similar features** have **similar antecedence**

m_1

m_2

m_3

m_4

m_5

Roland arrives as Ian waits. He stands with Mary ...

The rider moves closer.

Our Bidirectional Model: **Mention name assignment**

$\mathbf{Z} \in \{0,1\}^{M \times P}$ Name assignment for mentions

$\mathbf{R} \in \{0,1\}^{M^2}$ Antecedence variable

Antecedent clustering cost

$$\mathbf{R}^T \Pi_{men} \mathbf{R}$$

Convex quadratic in \mathbf{R}

m_2 m_3

Roland arrives as **lan** waits. **He** stands with Mary ...

The rider moves closer.

Our Bidirectional Model: **Mention name assignment**

$\mathbf{Z} \in \{0,1\}^{M \times P}$ Name assignment for mentions

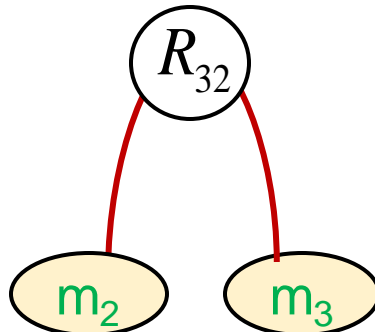
$\mathbf{R} \in \{0,1\}^{M^2}$ Antecedence variable

Antecedent clustering cost

$$\mathbf{R}^T \Pi_{men} \mathbf{R}$$

Convex quadratic in \mathbf{R}

\mathbf{R} constrains \mathbf{Z}



Antecedent mention shares the same name.

Roland arrives as **lan** waits. **He** stands with Mary ...

The rider moves closer.

Our Bidirectional Model: **Mention name assignment**

$\mathbf{Z} \in \{0,1\}^{M \times P}$ Name assignment for mentions

$\mathbf{R} \in \{0,1\}^{M^2}$ Antecedence variable

Antecedent clustering cost

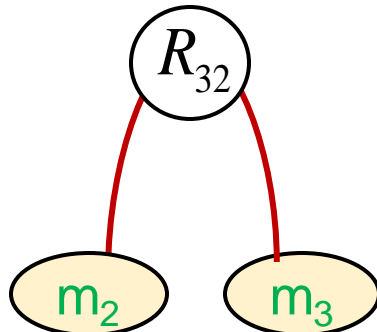
$$\mathbf{R}^T \Pi_{men} \mathbf{R}$$

Convex quadratic in \mathbf{R}

\mathbf{R} constrains \mathbf{Z}

$$\|\mathbf{z}_i - \mathbf{z}_j\|_{\infty} \leq (1 - \mathbf{R}_{ij})$$

Convex constraint



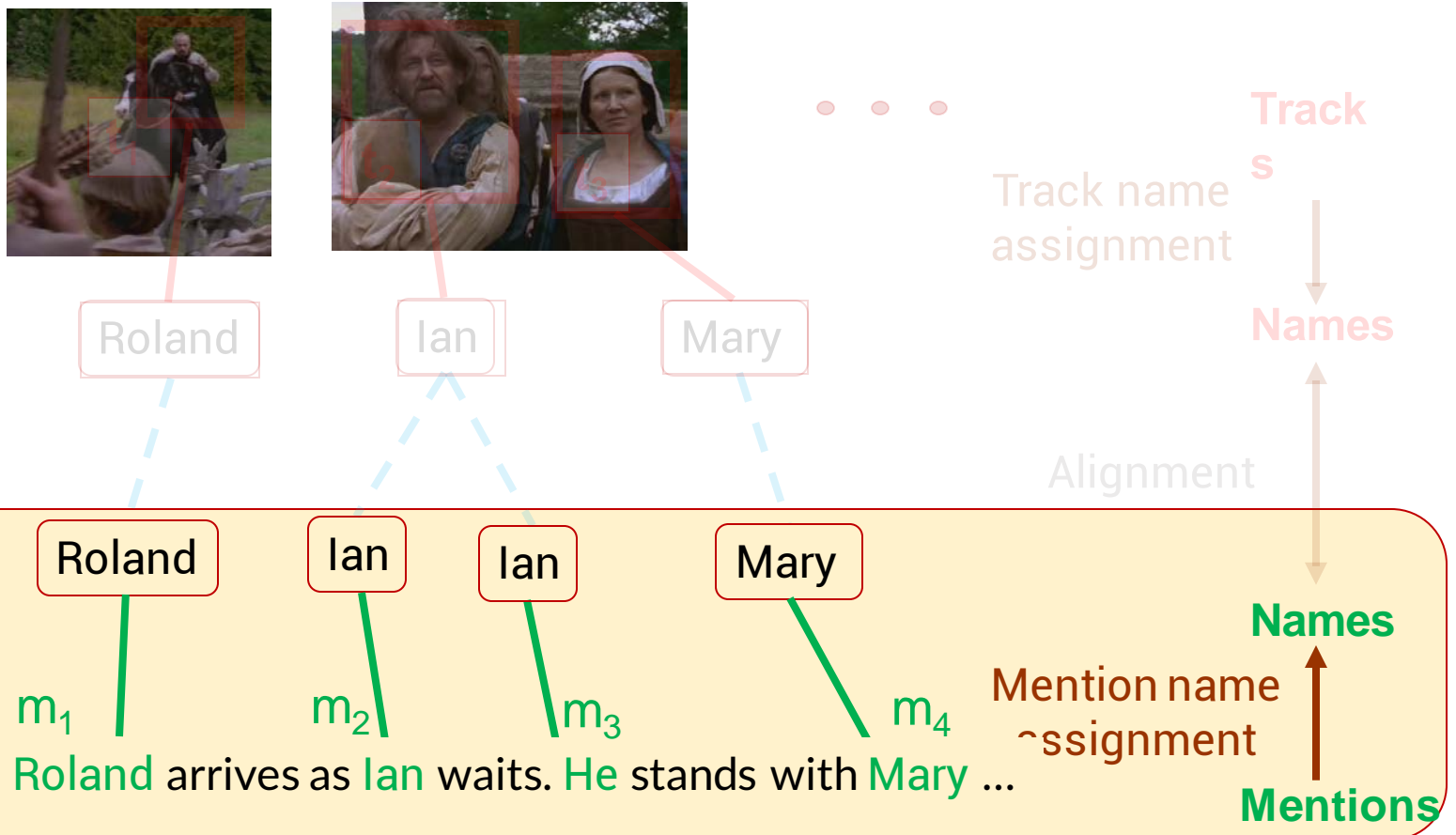
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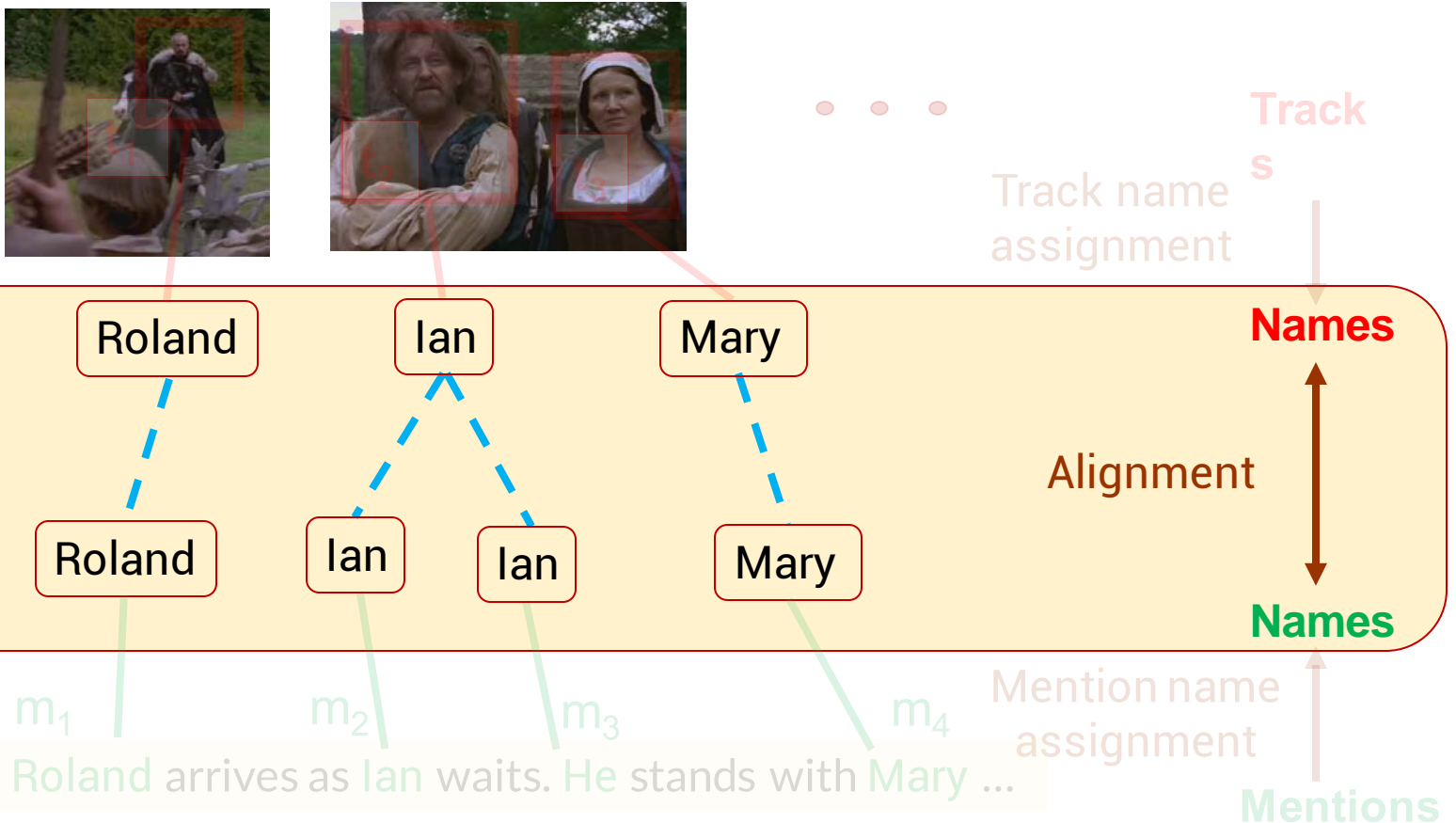
Our Bidirectional Model

$$\begin{aligned} \min \quad & \text{tr}(\mathbf{Y}^T \Pi_{track} \mathbf{Y}) + \mathbf{R}^T \Pi_{men} \mathbf{R} \\ \text{s.t} \quad & \mathbf{Y} \in C_Y, \quad \mathbf{Z}, \mathbf{R} \in C_{Z,R} \end{aligned}$$



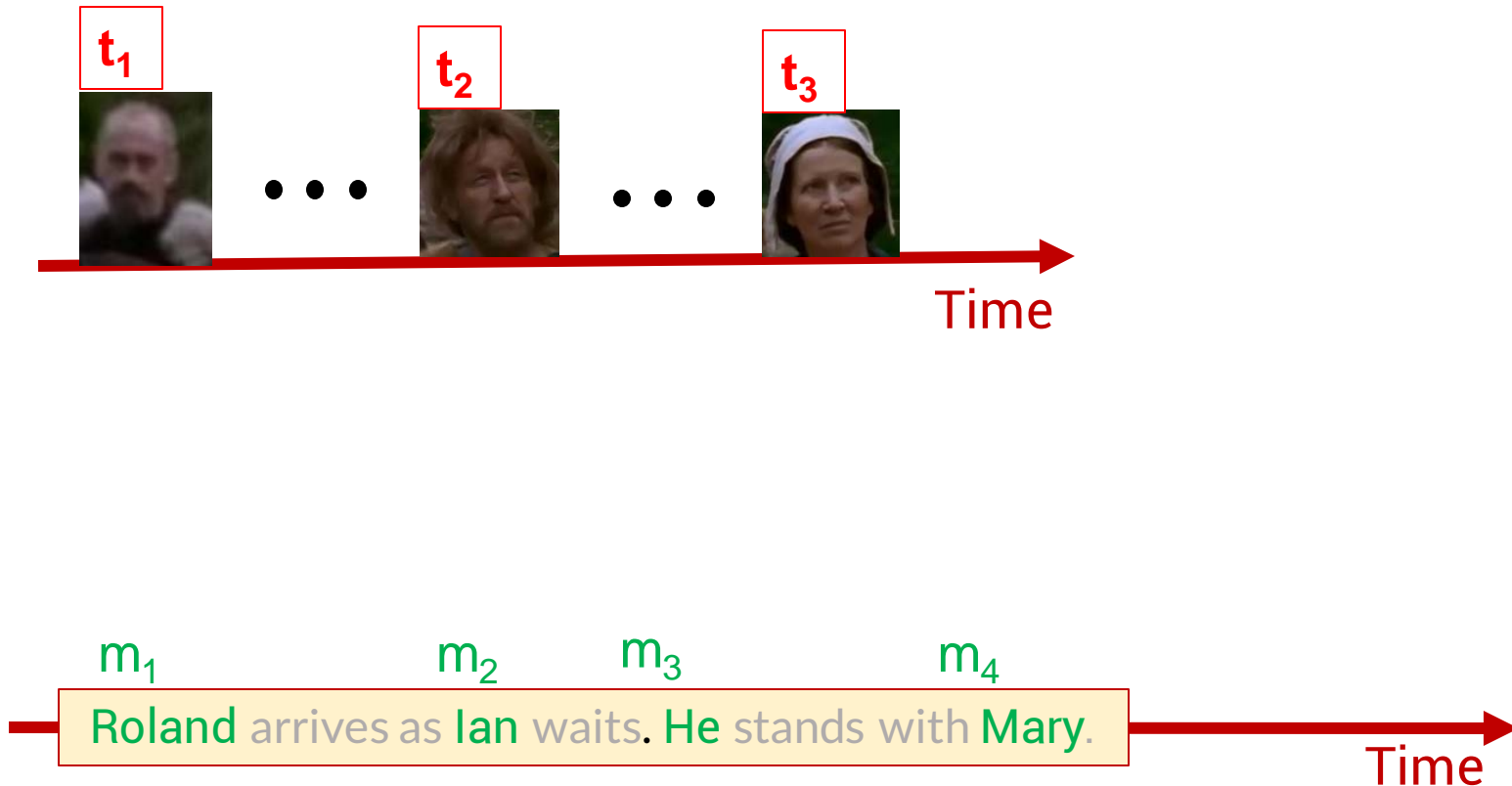
Our Bidirectional Model

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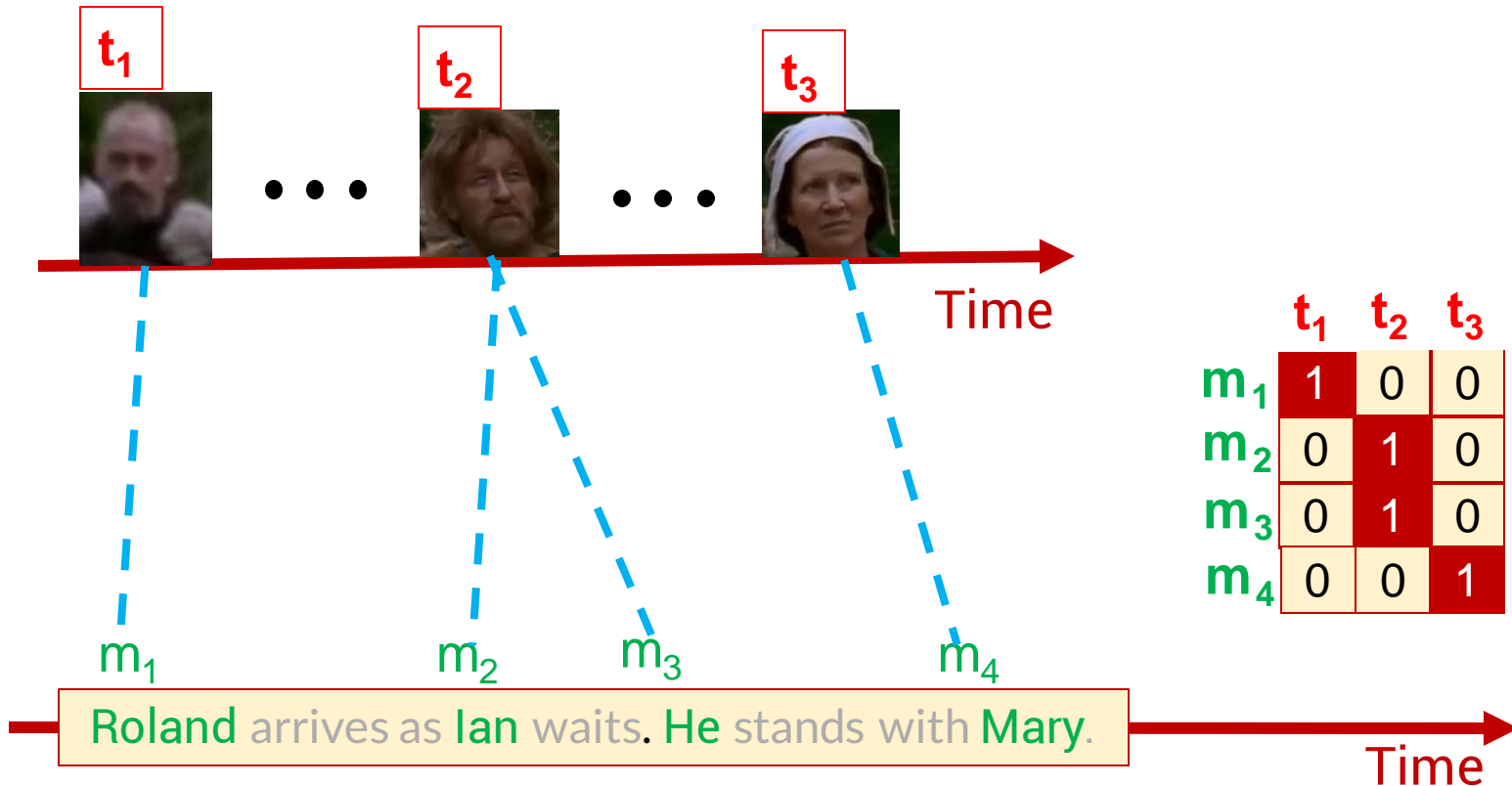
Our Bidirectional Model: **Alignment**

$\mathbf{A} \in \{0,1\}^{T \times M}$ Alignment matrix between tracks and mentions



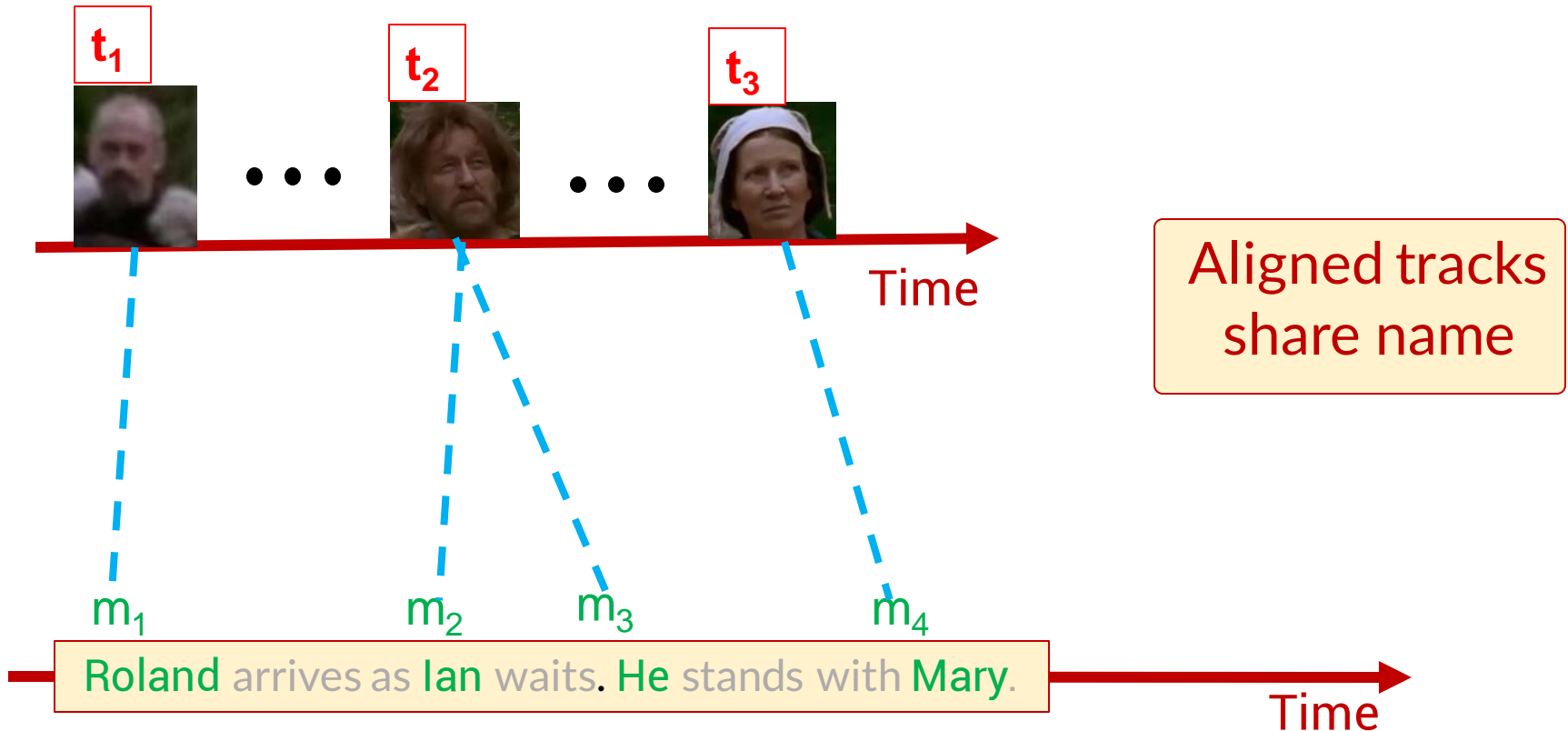
Our Bidirectional Model: **Alignment**

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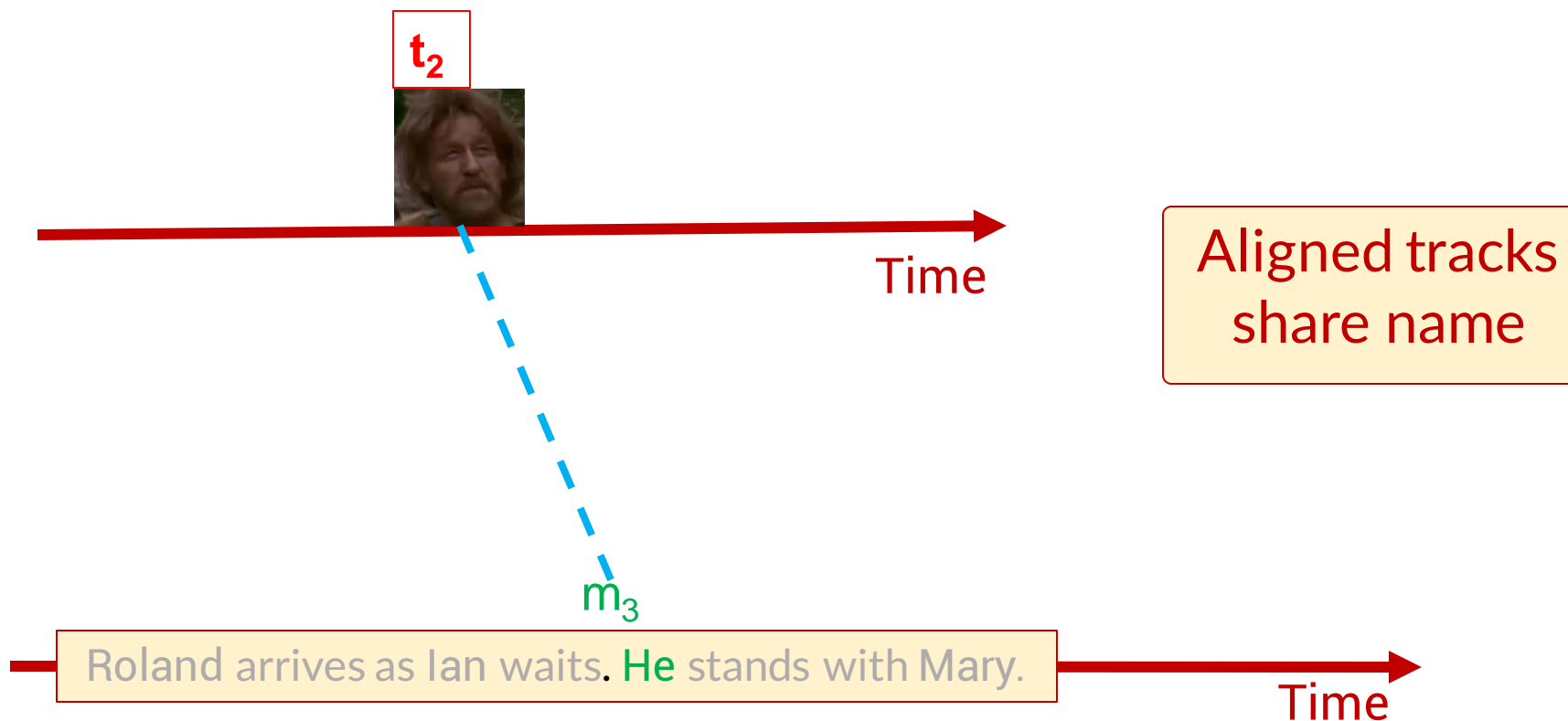
Our Bidirectional Model: **Alignment**

$\mathbf{A} \in \{0,1\}^{T \times M}$ Alignment matrix between tracks and mentions



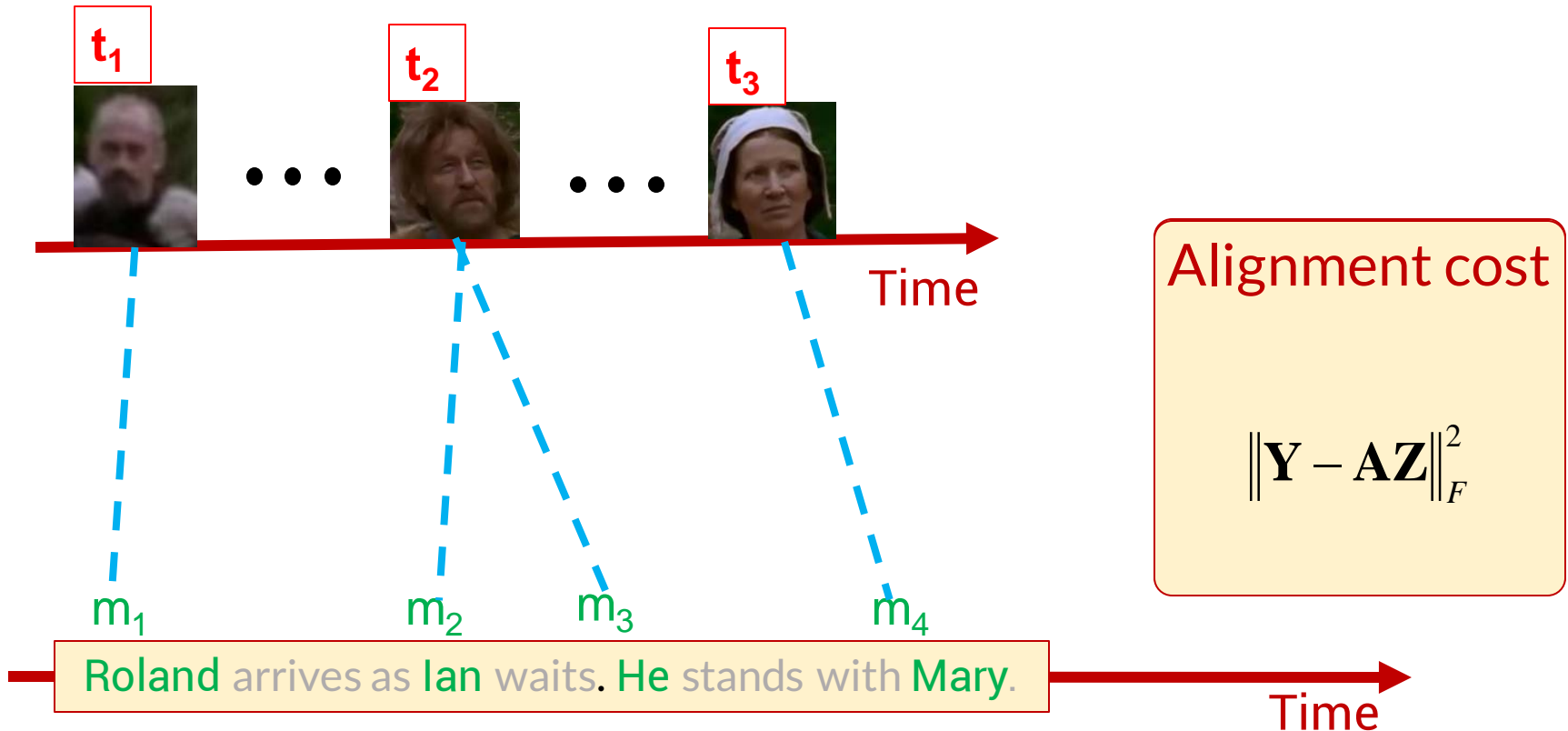
Our Bidirectional Model: **Alignment**

$\mathbf{A} \in \{0,1\}^{T \times M}$ Alignment matrix between tracks and mentions



Our Bidirectional Model: **Alignment**

$\mathbf{A} \in \{0,1\}^{T \times M}$ Alignment matrix between tracks and mentions



Our Bidirectional Model: **Alignment**

$\mathbf{A} \in \{0,1\}^{T \times M}$ Alignment matrix between tracks and mentions

Dynamic Time Warping

Dynamic program to optimize in \mathbf{A}
(enforces monotonic alignment)

Alignment cost

$$\|\mathbf{Y} - \mathbf{AZ}\|_F^2$$

Our Bidirectional Model

$$\begin{aligned} \min \quad & \text{tr}(\mathbf{Y}^T \Pi_{track} \mathbf{Y}) + \mathbf{R}^T \Pi_{men} \mathbf{R} + \|\mathbf{Y} - \mathbf{AZ}\|_F^2 \\ \text{s.t} \quad & \mathbf{Y} \in C_Y, \quad \mathbf{Z}, \mathbf{R} \in C_{Z,R}, \quad \mathbf{A} \in C_A \end{aligned}$$



...

Track
 S

Track name
assignment

Names

Alignment

Names

Mention name
assignment

Mentions

Roland

Ian

Mary

Roland

Ian

Ian

Mary

m_1

m_2

m_3

m_4

Roland arrives as Ian waits. He stands with Mary ...

Optimization

$$\begin{aligned} \min \quad & \text{tr}(\mathbf{Y}^T \Pi_{\text{track}} \mathbf{Y}) + \mathbf{R}^T \Pi_{\text{men}} \mathbf{R} + \|\mathbf{Y} - \mathbf{AZ}\|_F^2 \\ \text{s.t} \quad & \mathbf{Y} \in C_Y, \quad \mathbf{Z}, \mathbf{R} \in C_{Z,R}, \quad \mathbf{A} \in C_A \end{aligned}$$

- Relax $\mathbf{Y}, \mathbf{Z}, \mathbf{R}$ to be in $[0,1]$
- Use Block Coordinate Descent

Optimization

$$\begin{aligned} \min \quad & \text{tr}(\mathbf{Y}^T \Pi_{\text{track}} \mathbf{Y}) + \mathbf{R}^T \Pi_{\text{men}} \mathbf{R} + \|\mathbf{Y} - \mathbf{AZ}\|_F^2 \\ \text{s.t} \quad & \mathbf{Y} \in C_Y, \quad \mathbf{Z}, \mathbf{R} \in C_{Z,R}, \quad \mathbf{A} \in C_A \end{aligned}$$

- Relax $\mathbf{Y}, \mathbf{Z}, \mathbf{R}$ to be in $[0,1]$
- Use Block Coordinate Descent
- Quadratic Program to optimize \mathbf{Y}

Optimization

$$\begin{aligned} \min \quad & \text{tr}(\mathbf{Y}^T \Pi_{\text{track}} \mathbf{Y}) + \mathbf{R}^T \Pi_{\text{men}} \mathbf{R} + \|\mathbf{Y} - \mathbf{AZ}\|_F^2 \\ \text{s.t} \quad & \mathbf{Y} \in C_Y, \quad \mathbf{Z}, \mathbf{R} \in C_{Z,R}, \quad \mathbf{A} \in C_A \end{aligned}$$

- Relax $\mathbf{Y}, \mathbf{Z}, \mathbf{R}$ to be in $[0,1]$
- Use Block Coordinate Descent
- Quadratic Program to optimize \mathbf{Y}
- Quadratic Program to optimize \mathbf{Z}, \mathbf{R}

Optimization

$$\begin{aligned} \min \quad & \text{tr}(\mathbf{Y}^T \Pi_{track} \mathbf{Y}) + \mathbf{R}^T \Pi_{men} \mathbf{R} + \|\mathbf{Y} - \mathbf{AZ}\|_F^2 \\ \text{s.t} \quad & \mathbf{Y} \in C_Y, \quad \mathbf{Z}, \mathbf{R} \in C_{Z,R}, \quad \mathbf{A} \in C_A \end{aligned}$$

- Relax $\mathbf{Y}, \mathbf{Z}, \mathbf{R}$ to be in $[0,1]$
- Use Block Coordinate Descent
- Quadratic Program to optimize \mathbf{Y}
- Quadratic Program to optimize \mathbf{Z}, \mathbf{R}
- Dynamic Program to optimize in terms of \mathbf{A}

Linking people with “their” names

- Problem setup
- Our Bidirectional model
- **Experiments**
- Summary

Dataset



We reveal Lynette holding Porter by his feet, while he clings to Preston's desk.



Missy points to the larger kid. The big kid walks off. Other kids jeer.



Cary eyes the siblings, as Alicia looks across the bullpen

Dataset



We reveal Lynette holding Porter by his feet, while he clings to Preston's desk.



Missy points to the larger kid. The big kid walks off. Other kids jeer.



Cary eyes the siblings, as Alicia looks across the bullpen

pronoun/nominal

Dev. Set (14 episodes)

3329 tracks (3 eps.)

811 mentions

Dataset



We reveal Lynette holding Porter by his feet, while he clings to Preston's desk.



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Cary eyes the siblings, as Alicia looks across the bullpen

pronoun/nominal

Dev. Set (14 episodes)

3329 tracks (3 eps.)

811 mentions

- **No training set** since we have no labelled examples.
- The model **parameters** such as regularization constant are tuned on the **dev. set**.

Dataset



We reveal Lynette holding Porter by his feet, while he clings to Preston's desk.



Missy points to the larger kid. The big kid walks off. Other kids jeer.



Cary eyes the siblings, as Alicia looks across the bullpen

pronoun/nominal

Dev. Set (14 episodes)	3329 tracks (3 eps.)	811 mentions
Test Set (5 episodes)	4757 tracks	300 mentions

Dataset



We reveal Lynette holding Porter by his feet, while he clings to Preston's desk.



Missy points to the larger kid. The big kid walks off. Other kids jeer.



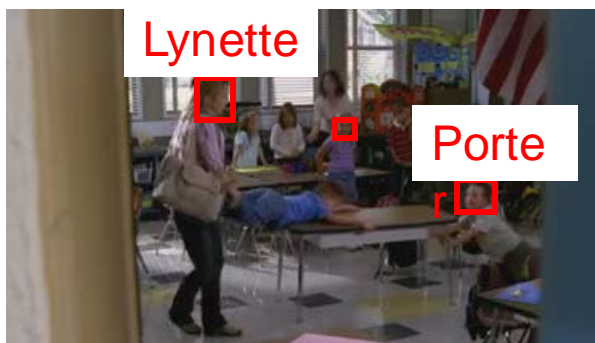
Cary eyes the siblings, as Alicia looks across the bullpen

		pronoun/nominal
Dev. Set (14 episodes)	3329 tracks (3 eps.)	811 mentions
Test Set (5 episodes)	4757 tracks	300 mentions

Tasks:

- Name assignment to tracks;
- Name assignment to mentions;

Dataset



We reveal Lynette holding Porter by his feet, while he clings to Preston's desk.



Missy points to the larger kid. The big kid walks off. Other kids jeer.



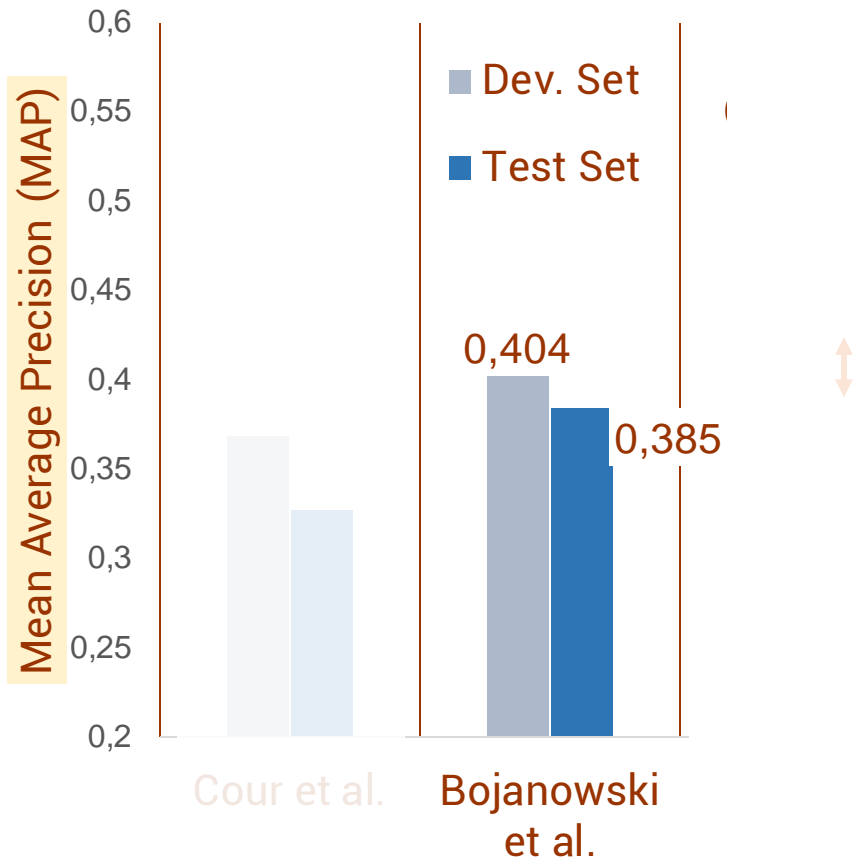
Cary eyes the siblings, as Alicia looks across the bullpen

Dev. Set (14 episodes)	3329 tracks (3 eps.)	pronoun/nominal 811 mentions
Test Set (5 episodes)	4757 tracks	300 mentions

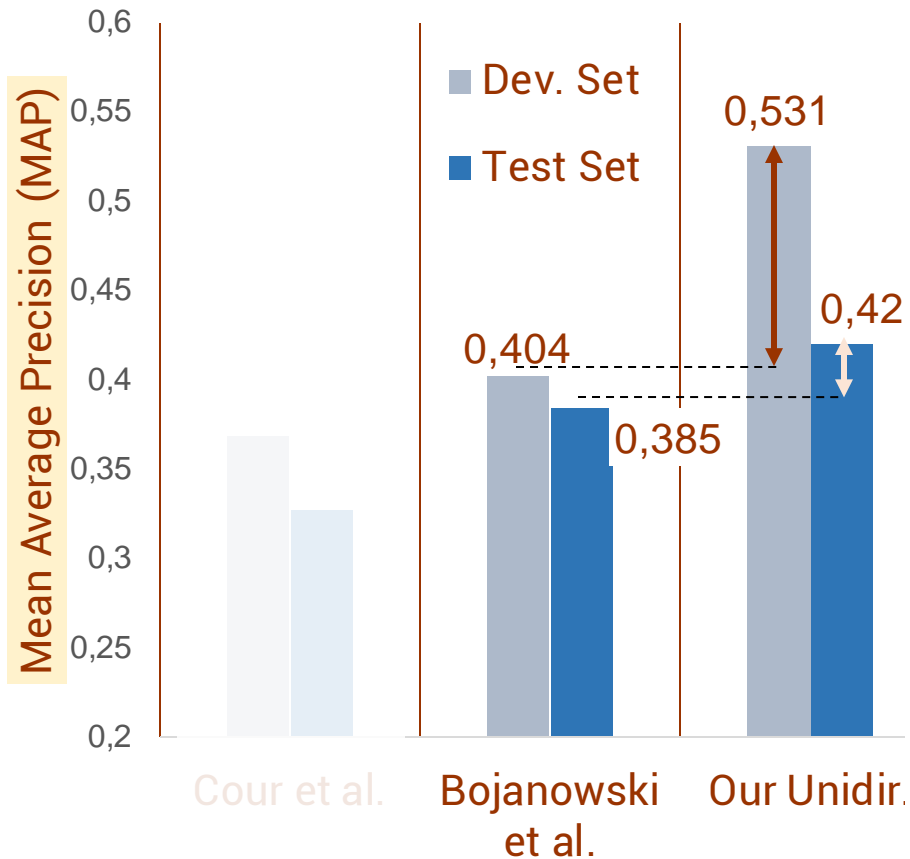
Tasks:

- **Name assignment to tracks;**
- Name assignment to mentions;

Experiments: Name Assignment to Tracks



Experiments: Name Assignment to Tracks



Our Unidirectional Model

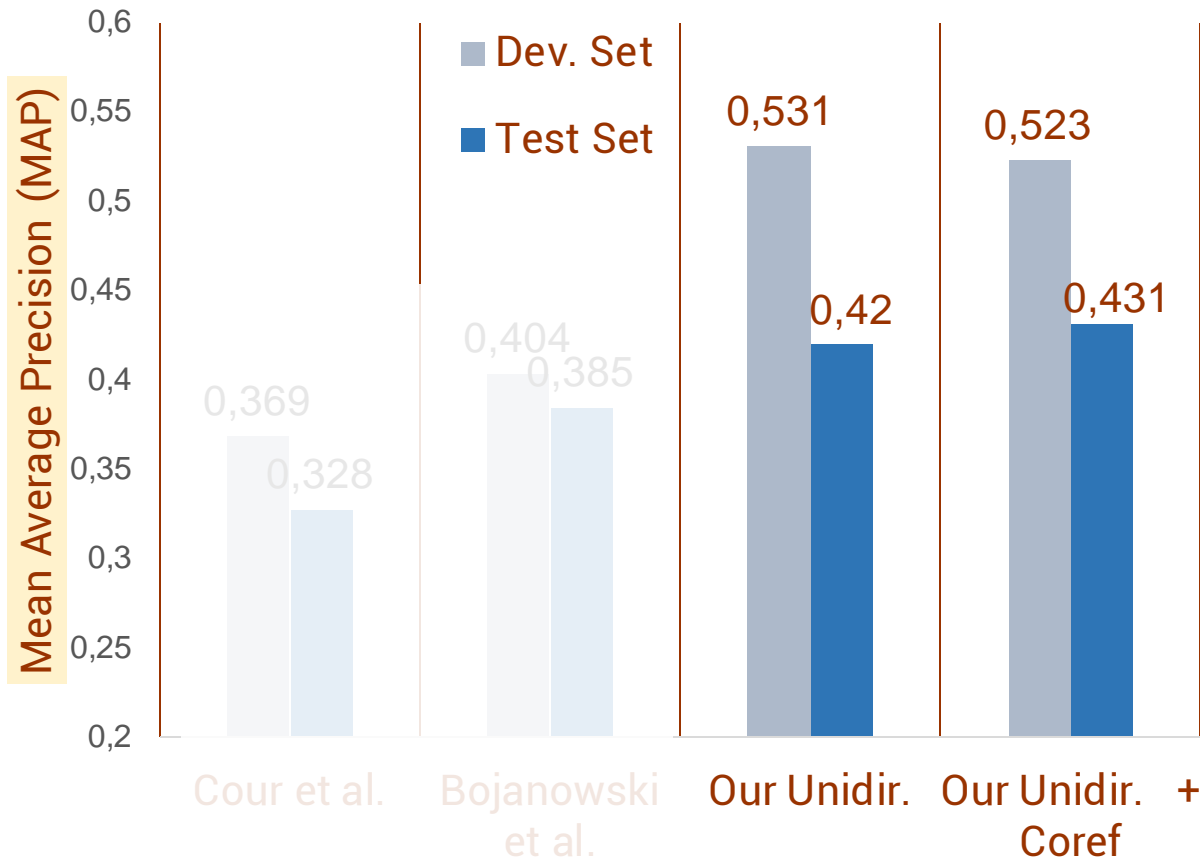
Bojanowski et al
(ICCV'13)
with

Additional Constraints

$$\min \operatorname{tr}(\mathbf{Y}^T \Pi_{track} \mathbf{Y})$$

$$\text{s.t. } \mathbf{Y} \in C_Y$$

Experiments: Name Assignment to Tracks

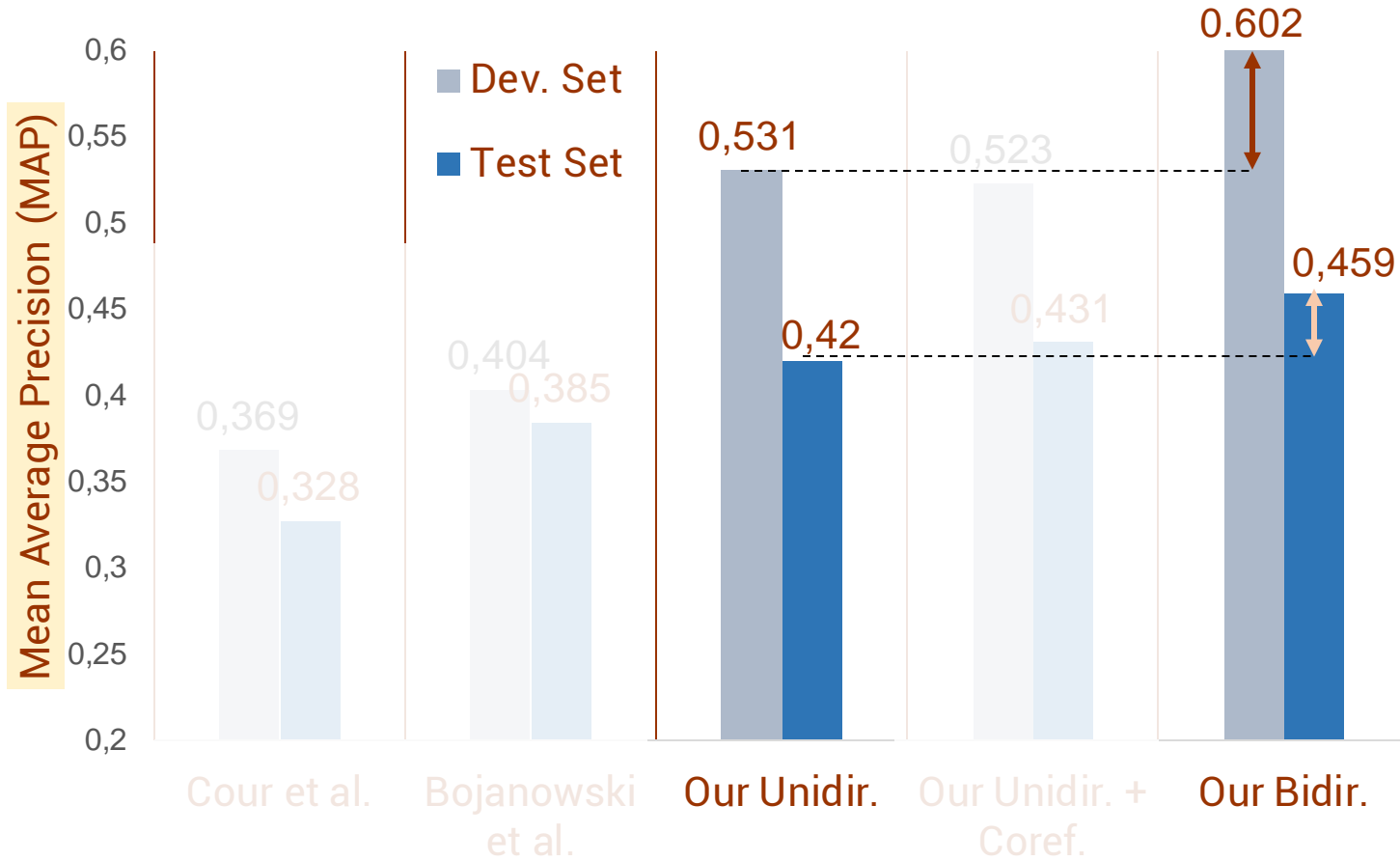


No Alignment

Unidirectional Track Name
+
Unidirectional Mention
Name

**without
Alignment**

Experiments: Name Assignment to Tracks



Bidirectional Model

Full Model with alignment

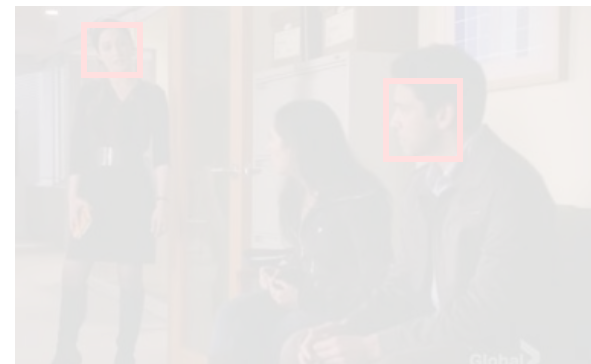
Dataset



We reveal Lynette **Porter** Porter by his feet, while **he** clings to Preston's desk.



Bryan points to the larger kid. The **big kid** walks off. Other kids jeer.



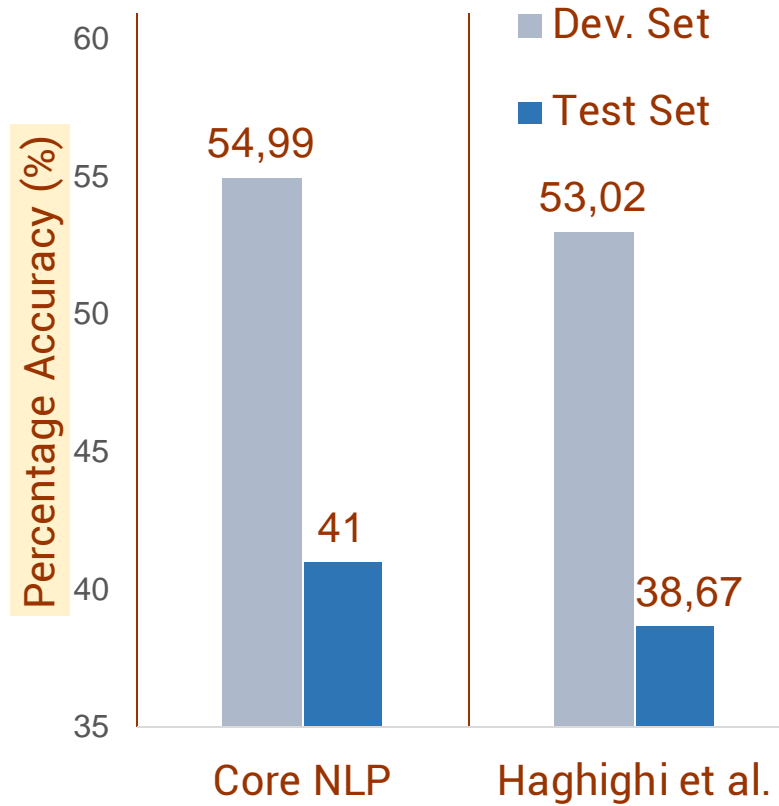
Cary eyes the siblings, as Alicia looks across the bullpen

		pronoun/nominal
Dev. Set (14 episodes)	3329 tracks (3 eps.)	811 mentions
Test Set (5 episodes)	4757 tracks	300 mentions

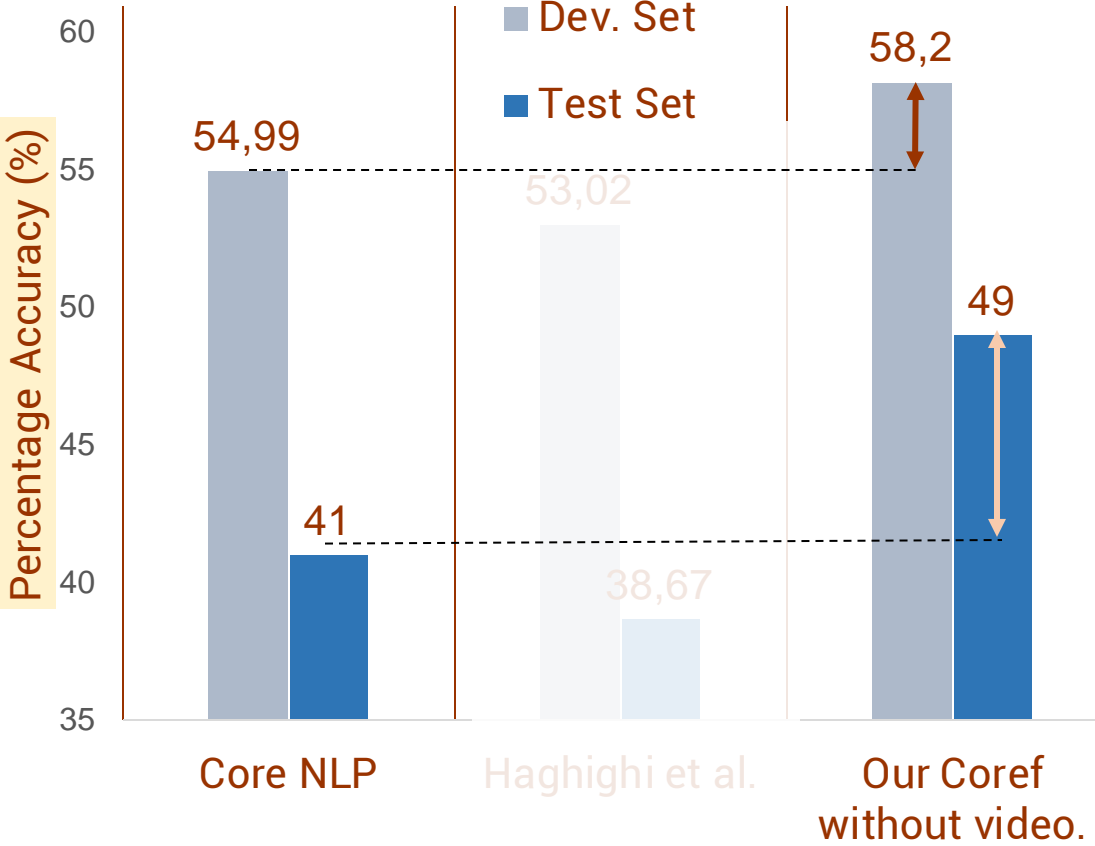
Tasks:

- Name assignment to tracks;
- **Name assignment to mentions;**

Experiments: Name Assignment to Mentions



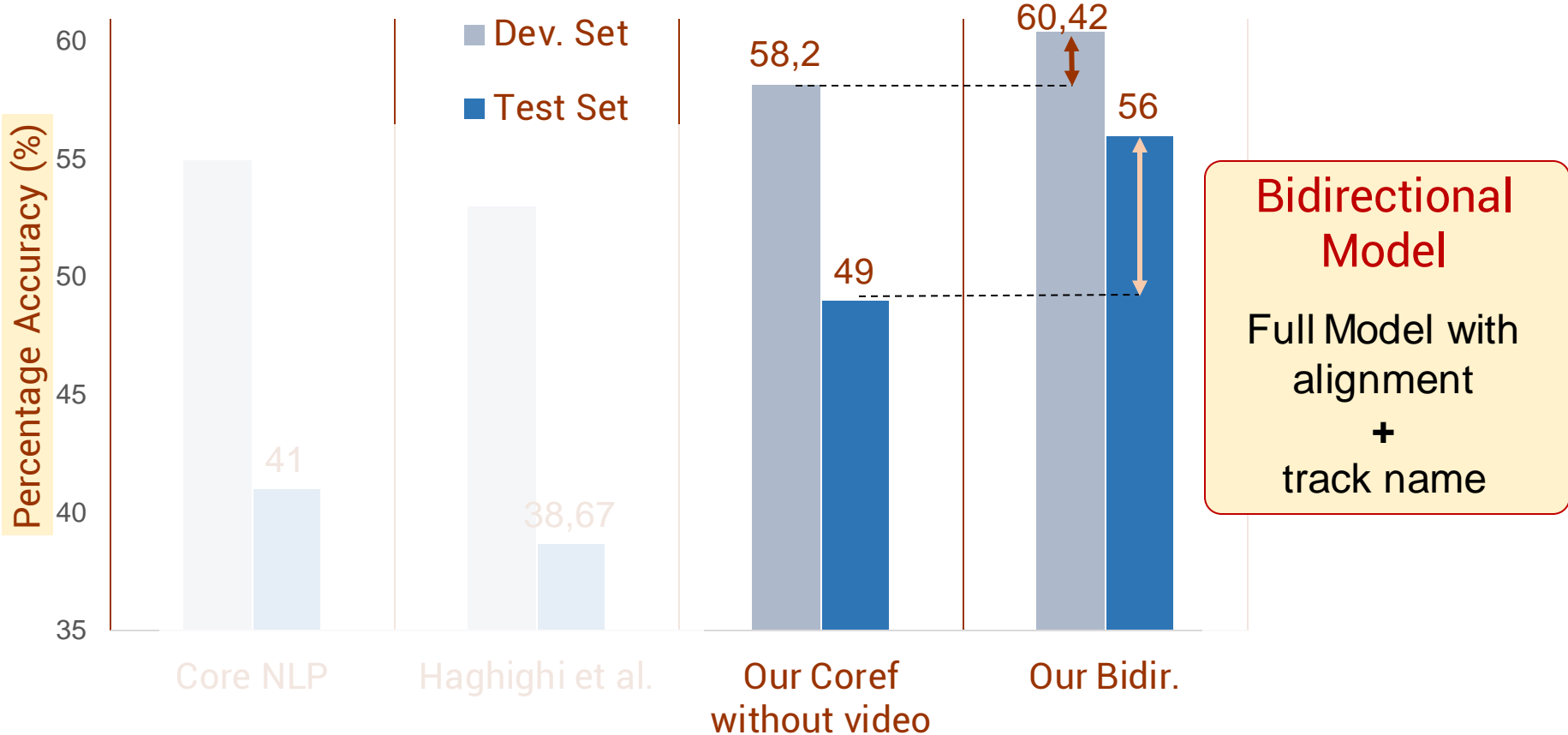
Experiments: Name Assignment to Mentions



No Video information used

Additional constraints specific to script data

Experiments: Name Assignment to Mentions



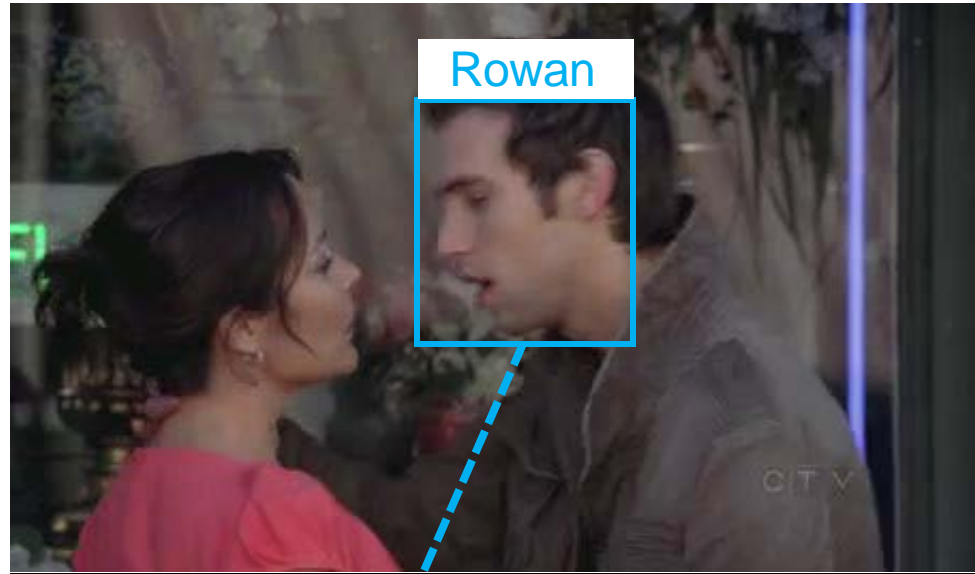
Experiments: Benefit of bidirectional model

Experiments: Benefit of bidirectional model



Gabriel cues the entry of Rowan. Rose doesn't notice him. He takes her in his arms.

Experiments: Benefit of bidirectional model



Rowan

Gabriel cues the entry of Rowan. Rose doesn't notice **him**. **He** takes her in his arms.

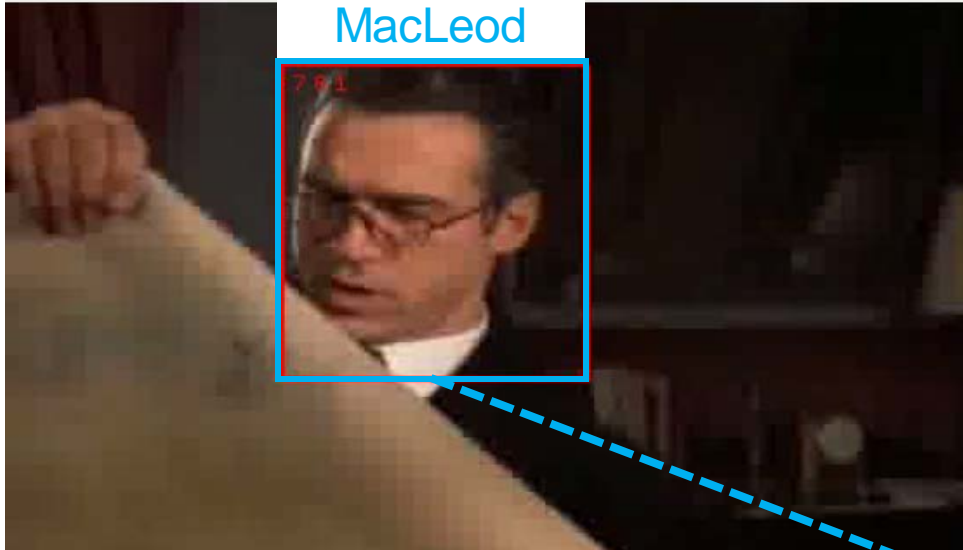
Gabriel(unidir), Rowan(bidir)

Experiments: Benefit of bidirectional model



Edouard & MacLeod unfurl the canvas, searching for the name. He then peers at it. She turns

Experiments: Benefit of bidirectional model



MacLeod

Edouard & MacLeod unfurl the canvas. **He** then peers at it. She turns.

Edouard(unidir), MacLeod(bidir)

Error Analysis: Name Assignment to Mentions



Beckett turns... **She** bites
her lips and shakes her
head

Beckett(unidir),
Castle(bidir)

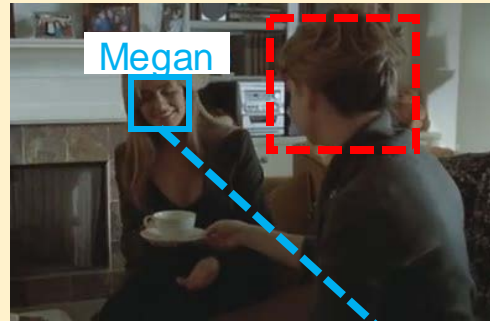
Low resolution faces lead
to bad name assignment!

Error Analysis: Name Assignment to Mentions



Beckett turns... **She** bites her lips and shakes her head

Beckett(unidir),
Castle(bidir)




Elaine Tillman, fragile but with inner strength. **She** looks to Megan.

Elaine(unidir), Megan(bidir)

Faces not detected or face not visible!


Error Analysis: Name Assignment to Mentions



Castle

Beckett turns... **She** bites her lips and shakes her head

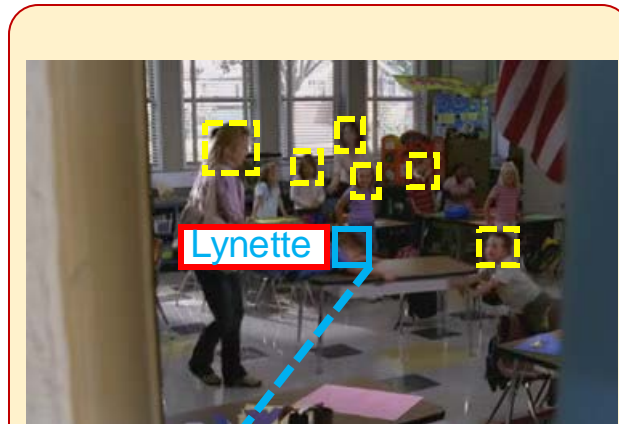
Beckett(unidir),
Castle(bidir)



Megan

Elaine Tillman, fragile but with inner strength. **She** looks to Megan.

Elaine(unidir), Megan(bidir)



Lynette

Lynette holds Porter by feet, while **he** clings to the desk.

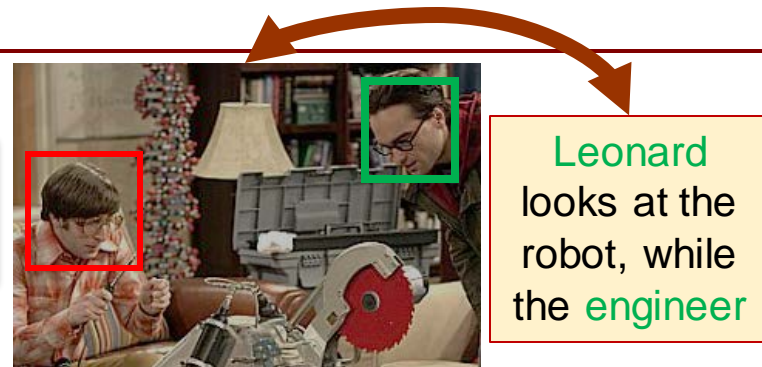
Long clip with too many people, leads to wrong alignment

Linking people with “their” names

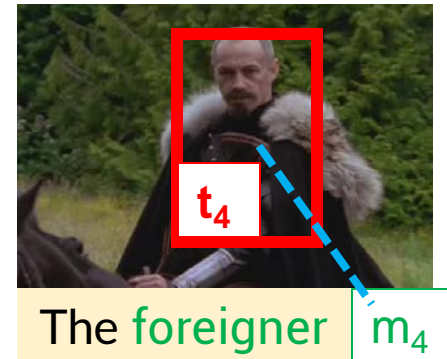
- Problem setup
- Our Bidirectional model
- Experiments
- **Summary**

Summary

Bidirectional model to handle challenging problems in Vision and NLP



Temporal ordering based alignment



Model evaluated on new TV episode Dataset



Cary eyes the siblings, as Alicia looks across the bullpen

Thank You



Microsoft®
Research

Mind's Eye

NIST
National Institute of
Standards and Technology

Google
Research

