## Data In, Facts Out: <br> Automated Monitoring of Facts by FactWatcher

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## FactWatcher



Tuple $t$ for new real world event appended to database

| Constraint | Measure |
| :--- | :--- |
| month $=$ Feb | pts, ast, reb |
| opp_team=Nets | ast, reb |
|  <br> opp_team=Nets | ast, reb |
| $\ldots$ | $\ldots$ |

Generate factual claim
http://en.wikipedia.org/wiki/Basketball

Wesley had 12 points, 13 assists and 5 rebounds on February 25, 1996 to become the first player with a 12/13/5 (points/assists/rebounds) in February.

## Fact Finding

## Prominent streaks

Long consecutive subsequence of high values in a sequence

## One-of-the-few objects

Qualifying statements that can only be made for very few objects

## Situational facts

Comparison contexts and spaces that make a given object stand out

## FactWatcher Finds Three Types of Facts (and can be Extended)

## Domains

- sports, weather, crimes, transportation, finance, social media analytics


## Examples from Real News Media

## Prominent streaks

- "This month the Chinese capital has experienced 10 days with a maximum temperature in around 35 degrees Celsius - the most for the month of July in a decade." http://www.chinadaily.com.cn/china/2010-07/27/content_11055675.htm
- "The Nikkei 225 closed below 10000 for the 12th consecutive week, the longest such streak since June 2009."
http://www.bloomberg.com/news/articles/2010-08-06/japanese-stocks-fall-for-second-day-this-week-on-u-s-jobless-claims-yen
$A$


## FactWatcher Finds Three Types of Facts (and can be Extended)

## Examples from Real News Media

## Situational facts, One-of-the-few objects

- "Paul George had 21 points, 11 rebounds and 5 assists to become the first Pacers player with a 20/10/5 (points/rebounds/assists) game against the Bulls since Detlef Schrempf in December 1992."
- "The social world's most viral photo ever generated 3.5 million likes, 170,000 comments and 460,000 shares by Wednesday afternoon." htpp://www.cnbc.com/id/49728455


## FactWatcher Demo

 http://idir.uta.edu/factwatcher/ https://vimeo.com/user48311227
## LIVE UPDATE

[February 20, 1998] Todd Fuller had 1 assist, 3 steals and 1 block in the Golden State Warriors' defeat against the Denver Nuggets. It is one of the best performance made by him.

## Presented In

Excellent Demo Award

## COMPUTATION + JOURNALISM

## http://idir.uta.edu/factwatcher/

[April 24, 1994] David Robinson had 71 points and 14 rebounds in the San Antonio Spurs' victory against the Los Angeles Clippers. No one before had a better performance in NBA history.
[April 20, 1994] Shaquille O'neal had 53 points and 18 rebounds in the Orlando Magic's win over the Minnesota Timberwolves. No one before had a better performance in NBA history.
[February 16, 1993] Shaquille O'neal had 46 points and 21 rebounds in the Orlando Magic's defeat against the Detroit Pistons. No one before had a better performance in NBA history.
[February 27, 1992] David Robinson had 37 points and 24 rebounds in the San Antonio Spurs' victory against the Golden State Warriors. No one before had a better performance in NBA history.

Compare Similar Stories


Number of Facts

STL


## How were these Facts Discovered in Current Systems?

## Our (educated?) guess

- Experts monitor real-world events (e.g., watching an NBA game), have a gut-feeling, issue database queries, check out or not
- Prepared facts-to-be (e.g., Nowitzki only needs 477 more points to surpass O'Neal. Perhaps will happen around Christmas 2015)
- Predefined templates of facts/database queries
- Perhaps in-house systems/algorithms similar to FactWatcher
$\leftarrow \rightarrow \mathrm{C} \square$ www.esb.com $\quad$ 切


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 Sports Bureau

## Welcome to the Elias Sports Bureau

We are the world's leading source of statistics and historical data for the professional sports industry

The World's Foremost Sports Statisticians and Historians

## StatSheet

## No. 1-Seeded Louisville Clips No. 4-Seeded Michigan 82-76, Wins NCAA Championship

Filed under Game Recap on April 9th, 2013

## Share this recap

WTweet or $\mathbb{C}$ Like One person likes this. Be the first of your friends.

## NCAA Tournament 7th Round

|  | 1ST | 2ND | TOTAL | SPREAD | Mon, Apr 08 2013, 10:23 PM EDT |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| \#4 Michigan | 38 | 38 | 76 | +4.0 | Georgia Dome <br> Atlanta, Georgia |  |
| \#1 Louisville | 37 | 45 | 82 | -4.0 |  | Attendance: 74,326 <br> TV: CBS |

- StatSeed: NCAA Automatic \#1 Seed


More about Fan Satisfaction

Find another NCAA team:


| Hotbes | New Posts | Popular <br> Most Reputable Comp: | Lists <br> The Global 2000 | Video <br> Hip-Hon's Wealthiest | Search | 0 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |

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 Science
## Narrative Science

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Handling \$421 billion in accounts payables annually for companies like yours.

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in Share

## 0

(3) Submit

## Publications

- Online Frequent Episode Mining. Xiang Ao, Ping Luo, Chengkai Li, Fuzhen Zhuang, and Qing He. ICDE 2015, pages 891-902.
- Data In, Fact Out: Automated Monitoring of Facts by FactWatcher. Naeemul Hassan, Afroza Sultana, You Wu, Gensheng Zhang, Chengkai Li, Jun Yang, and Cong Yu. VLDB 2014, pages 1557-1560. Demonstration description. (excellent demonstration award)
- Finding, Monitoring, and Checking Claims Computationally Based on Structured Data. Brett Walenz, You (Will) Wu, Seokhyun (Alex) Song, Emre Sonmez, Eric Wu, Kevin Wu, Pankaj K. Agarwal, Jun Yang, Naeemul Hassan, Afroza Sultana, Gensheng Zhang, Chengkai Li, Cong Yu. 2014 Computation+Journalism Symposium.
- Incremental Discovery of Prominent Situational Facts. Afroza Sultana, Naeemul Hassan, Chengkai Li, Jun Yang, Cong Yu. ICDE 2014, pages 112-123.
- Discovering General Prominent Streaks in Sequence Data. Gensheng Zhang, Xiao Jiang, Ping Luo, Min Wang, Chengkai Li. ACM TKDD, 8(2):article 9, June 2014.
- Discovering and Learning Sensational Episodes of News Events. Xiang Ao, Ping Luo, Chengkai Li, Fuzhen Zhuang, Qing He, and Zhongzhi Shi. WWWW 2014, pages 217-218.
- On "One of the Few" Objects. You Wu, Pankaj K. Agarwal, Chengkai Li, Jun Yang, Cong Yu. KDD 2012, pages 1487-1495.
- ProminentStreak Discovery in Sequence Data. Xiao Jiang, Chengkai Li, Ping Luo, Min Wang, Yong Yu. KDD 2011, pages 1280-1288.


## Situational Facts

Incremental Discovery of Prominent Situational Facts. Afroza Sultana, Naeemul Hassan, Chengkai Li, Jun Yang, Cong Yu. ICDE 2014, pages 112-123.

## Situational Facts

"Paul George had 21 points, 11 rebounds and 5 assists to become the first Pacers player with a 20/10/5 (points/rebounds/assists) game against the Bulls since Detlef Schrempf in December 1992." (http://espn.go.com/espn/elias?date=20130205)

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"The social world's most viral photo ever generated 3.5 million likes, 170,000 comments and 460,000 shares by Wednesday afternoon."
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## Situational Facts

-Stock Data: Stock A becomes the first stock in history with price over $\$ 300$ and market cap over $\$ 400$ billion.
-Weather Data: Today's measures of wind speed and humidity are $x$ and y, respectively. City B has never encountered such high wind speed and humidity in March.
-Criminal Records: There were 50 DUI arrests and 20 collisions in city $C$ yesterday, the first time in 2013.


## A Mini-world of Basketball Gamelogs

| id | player | day | month | season | team | opp_team | pts | ast | reb |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| $t_{1}$ | Bogues | 11 | Feb. | $1991-92$ | Hornets | Hawks | 4 | 12 | 5 |
| $t_{2}$ | Seikaly | 13 | Feb. | $1991-92$ | Heat | Hawks | 24 | 5 | 15 |
| $t_{3}$ | Sherman | 7 | Dec. | $1993-94$ | Celtics | Nets | Nets | 13 | 13 |
| $t_{4}$ | Wesley | 4 | Feb. | $1994-95$ | Celtics | 5 |  |  |  |
| $t_{5}$ | Wesley | 5 | Feb. | $1994-95$ | Celtics | Timberwolves | 3 | 5 | 3 |
| $t_{6}$ | Strictland | 3 | Jan. | $1995-96$ | Blazers | Celtics | 27 | 18 | 8 |
| $t_{7}$ | Wesley | 25 | Feb. | $1995-96$ | Celtics | Nets | 12 | 13 | 5 |

## Last tuple appenaed to tade

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| $t_{3}$ | Sherman | 7 | Dec. | $1993-94$ | Celtics | Nets | 13 | 13 | 5 |
| $t_{4}$ | Wesley | 4 | Feb. | $1994-95$ | Celtics | Nets | 2 | 5 | 2 |
| $t_{5}$ | Wesley | 5 | Feb. | $1994-95$ | Celtics | Timberwolves | 3 | 5 | 3 |
| $t_{6}$ | Strictland | 3 | Jan. | $1995-96$ | Blazers | Celtics | 27 | 18 | 8 |
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## A Mini-world of Basketball Gamelogs

| id |  |  | month |  |  |  | pts | ast | reb |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| $t_{1}$ |  |  | Feb. |  |  |  |  |  |  |
| $t_{2}$ |  |  | Feb. |  |  |  |  | 12 |  |
|  |  |  |  |  |  |  | 5 |  |  |
| $t_{4}$ |  |  | Feb. |  |  |  | 5 | 15 |  |
| $t_{5}$ |  |  | Feb. |  |  |  |  |  |  |
|  |  |  |  |  |  | 2 | 5 | 2 |  |
| $t_{7}$ |  |  |  |  |  |  |  |  |  |

## A Mini-world of Basketball Gamelogs

| id |  |  | month |  |  |  | pts | ast | reb |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| $t_{1}$ |  |  | Feb. |  |  |  |  | 4 | 12 |
| $t_{2}$ |  |  | Feb. |  |  |  | 5 |  |  |
|  |  |  |  |  |  |  | 24 | 5 | 15 |
| $t_{4}$ |  |  | Feb. |  |  |  |  |  |  |
| $t_{5}$ |  |  | Feb. |  |  |  |  |  |  |
|  |  |  |  |  |  | 2 | 5 | 2 |  |
| $t_{7}$ |  |  |  |  |  |  | 5 | 3 |  |

- Wesley had 12 points, 13 assists and 5 rebounds on February 25, 1996 to become the first player with a 12/13/5 (points/assists/rebounds) in February.


## A Mini-world of Basketball Gamelogs

| id |  |  |  | season |  |  | pts | ast | reb |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |
| $t_{6}$ |  |  |  |  |  |  |  |  |  |
| $t_{7}$ |  |  |  |  |  |  |  |  |  |

## A Mini-world of Basketball Gamelogs

| id |  |  |  |  |  | team | opp_team |  | ast | reb |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |  |
| $t$ |  |  |  |  |  | Celties | Nets |  | 13 | 5 |
| $t_{\text {t }}$ |  |  |  |  |  | Celtics | Nets |  | 5 | 2 |
|  |  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |  |
| $\square$ |  |  |  |  |  | Celties | Nets |  | 13 | 5 |

-Wesley had 13 assists and 5 rebounds on February 25, 1996 to become the second Celtics player with a $13 / 5$ (assists/rebounds) game against the Nets.

## Problem Definition

## Dimension space: $\mathscr{D}=\left\{d_{1}, \ldots, d_{n}\right\}$

Measure space: $\mathscr{M}=\left\{m_{1}, \ldots, m_{s}\right\}$

| id | player | day | month | season | team | opp_team | pts | ast | reb |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| $t_{1}$ | Bogues | 11 | Feb. | $1991-92$ | Hornets | Hawks | 4 | 12 | 5 |
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| $t_{6}$ | Strictland | 3 | Jan. | $1995-96$ | Blazers | Celtics | 27 | 18 | 8 |

## append-only table

## Problem Definition

$\square$ Constraint $(C): d_{1}=v_{1} \wedge d_{2}=v_{2} \wedge \ldots \wedge d_{n}=v_{n}, v_{i} \in \operatorname{dom}\left(d_{i}\right) \cup\{*\}$

- team=Celtics $\wedge$ opp_team=Nets

| id |  |  |  |  | team | opp_team |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |
| $t_{3}$ |  |  |  |  | Celtics | Nets |  |  |  |
| $t_{4}$ |  |  |  |  | Celtics | Nets |  |  |  |
|  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |

## Problem Definition

$\square$ Constraint-Measure Pair ( $C, M$ ): Combination of a constraint and measure subspace

- (team=Celtics $\wedge$ opp_team=Nets, \{assists,rebounds\})

| id |  |  |  |  | team | opp_team |  | ast | reb |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |
| $t_{3}$ |  |  |  |  |  | Celtics | Nets |  |  |
| $t_{4}$ |  |  |  |  | Celtics | Nets |  | 5 | 5 |
|  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |

## Problem Definition

$\square$ Contextual skyline: skyline regarding ( $C, M$ )

- $\sigma_{\text {team }}=$ Celtics $\wedge$ opp_team=Nets $(R), M=\{$ assists,rebounds $\}$ $>\left\{t_{3}\right\}$

| id |  |  |  |  | team | opp_team |  | ast | reb |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |
| $t_{3}$ |  |  |  |  | Celtics | Nets |  | 13 | 5 |
| $t_{4}$ |  |  |  |  | Celtics | Nets |  | 5 | 2 |
|  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |

## FactWatcher



Tuple $t$ for new real world event appended to database

| Constraint | Measure |
| :--- | :--- |
| month $=$ Feb | pts, ast, reb |
| opp_team=Nets | ast, reb |
|  <br> opp_team=Nets | ast, reb |
| $\ldots$ | $\ldots$ |

Generate factual claim
http://en.wikipedia.org/wiki/Basketball

Wesley had 12 points, 13 assists and 5 rebounds on February 25, 1996 to become the first player with a 12/13/5 (points/assists/rebounds) in February.

## Modeling

| id | $\boldsymbol{d}_{\boldsymbol{1}}$ | $\boldsymbol{d}_{\boldsymbol{2}}$ | $\boldsymbol{d}_{\mathbf{3}}$ | $\boldsymbol{m}_{\boldsymbol{1}}$ | $\boldsymbol{m}_{\boldsymbol{2}}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $t_{1}$ | $a_{1}$ | $b_{2}$ | $c_{2}$ | 10 | 15 |
| $t_{2}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | 15 | 10 |
| $t_{3}$ | $a_{2}$ | $b_{1}$ | $c_{2}$ | 17 | 17 |
| $t_{4}$ | $a_{2}$ | $b_{1}$ | $c_{1}$ | 20 | 20 |
| $t_{5}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | 11 | 15 |



Tuple Satisfied Constraint $C^{t}$ : If $\forall d_{i} \in \mathscr{O}, C . d_{i}=*$ or $C . d_{i}=t . d_{i}, t$ satisfies C.
(

| $\boldsymbol{i d}$ | $\boldsymbol{d}_{\boldsymbol{1}}$ | $\boldsymbol{d}_{2}$ | $\boldsymbol{d}_{3}$ | $\boldsymbol{m}_{\boldsymbol{1}}$ | $\boldsymbol{m}_{2}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $t_{1}$ | $a_{1}$ | $b_{2}$ | $c_{2}$ | 10 | 15 |
| $t_{2}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | 15 | 10 |
| $t_{3}$ | $a_{2}$ | $b_{1}$ | $c_{2}$ | 17 | 17 |
| $\boldsymbol{t}_{4}$ | $\boldsymbol{a}_{2}$ | $b_{1}$ | $c_{1}$ | 20 | 20 |
| $t_{5}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | $\mathbf{1 1}$ | $\mathbf{1 5}$ |

## Modeling

## Lattice of $C^{4}$



Lattice of $C^{s}$

( | $\boldsymbol{i d}$ | $\boldsymbol{d}_{\boldsymbol{1}}$ | $\boldsymbol{d}_{2}$ | $\boldsymbol{d}_{3}$ | $\boldsymbol{m}_{\boldsymbol{1}}$ | $\boldsymbol{m}_{2}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $t_{1}$ | $a_{1}$ | $b_{2}$ | $c_{2}$ | 10 | 15 |
| $t_{2}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | 15 | 10 |
| $t_{3}$ | $a_{2}$ | $b_{1}$ | $c_{2}$ | 17 | 17 |
| $\boldsymbol{t}_{4}$ | $\boldsymbol{a}_{2}$ | $\boldsymbol{b}_{1}$ | $c_{1}$ | $\mathbf{2 0}$ | $\mathbf{2 0}$ |
| $\boldsymbol{t}_{5}$ | $\boldsymbol{a}_{1}$ | $\boldsymbol{b}_{1}$ | $c_{1}$ | $\mathbf{1 1}$ | $\mathbf{1 5}$ |



Lattice of $C$.


## Lattice of $C^{s}$

Lattice Intersection: $C^{t_{4}} t_{s}=C * \cap C_{s}$

## Brute-Force Approach

| $\boldsymbol{i d}$ | $\boldsymbol{d}_{\boldsymbol{1}}$ | $\boldsymbol{d}_{\boldsymbol{2}}$ | $\boldsymbol{d}_{\mathbf{3}}$ | $\boldsymbol{m}_{\boldsymbol{1}}$ | $\boldsymbol{m}_{\boldsymbol{2}}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
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| $t_{2}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | 15 | 10 |
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| :---: | :---: | :---: | :---: | :---: | :---: |
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| $t_{2}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | 15 | 10 |
| $t_{3}$ | $a_{2}$ | $b_{1}$ | $c_{2}$ | 17 | 17 |
| $t_{4}$ | $a_{2}$ | $b_{1}$ | $c_{1}$ | 20 | 20 |
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| $t_{3}$ | $a_{2}$ | $b_{1}$ | $c_{2}$ | 17 | 17 |
| $t_{4}$ | $a_{2}$ | $b_{1}$ | $c_{1}$ | 20 | 20 |
| $t_{5}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | 11 | 15 |



## Brute-Force Approach

| $\boldsymbol{i d}$ | $\boldsymbol{d}_{\boldsymbol{1}}$ | $\boldsymbol{d}_{\mathbf{2}}$ | $\boldsymbol{d}_{\mathbf{3}}$ | $\boldsymbol{m}_{\boldsymbol{1}}$ | $\boldsymbol{m}_{\boldsymbol{2}}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $t_{1}$ | $a_{1}$ | $b_{2}$ | $c_{2}$ | 10 | 15 |
| $t_{2}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | 15 | 10 |
| $t_{3}$ | $a_{2}$ | $b_{1}$ | $c_{2}$ | 17 | 17 |
| $t_{4}$ | $a_{2}$ | $b_{1}$ | $c_{1}$ | 20 | 20 |
| $t_{5}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | 11 | 15 |



## Brute-Force Approach

| $\boldsymbol{i d}$ | $\boldsymbol{d}_{\boldsymbol{1}}$ | $\boldsymbol{d}_{\mathbf{2}}$ | $\boldsymbol{d}_{\mathbf{3}}$ | $\boldsymbol{m}_{\boldsymbol{1}}$ | $\boldsymbol{m}_{\boldsymbol{2}}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
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| $t_{3}$ | $a_{2}$ | $b_{1}$ | $c_{2}$ | 17 | 17 |
| $t_{4}$ | $a_{2}$ | $b_{1}$ | $c_{1}$ | 20 | 20 |
| $t_{5}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | 11 | 15 |



## Brute-Force Approach

| $\boldsymbol{i d}$ | $\boldsymbol{d}_{\boldsymbol{1}}$ | $\boldsymbol{d}_{\mathbf{2}}$ | $\boldsymbol{d}_{\mathbf{3}}$ | $\boldsymbol{m}_{\boldsymbol{1}}$ | $\boldsymbol{m}_{\boldsymbol{2}}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $t_{1}$ | $a_{1}$ | $b_{2}$ | $c_{2}$ | 10 | 15 |
| $t_{2}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | 15 | 10 |
| $t_{3}$ | $a_{2}$ | $b_{1}$ | $c_{2}$ | 17 | 17 |
| $t_{4}$ | $a_{2}$ | $b_{1}$ | $c_{1}$ | 20 | 20 |
| $t_{5}$ | $a_{1}$ | $b_{1}$ | $c_{1}$ | 11 | 15 |

Total $|R|^{*}\left(2^{|\mathscr{D}+||\mathcal{M}|}-1\right)$ comparisons!
Total 16 comparisons in this case!


## Challenges

$>$ Exhaustive comparison with every tuple > Under every constraint
$>$ Over every measure subspace

## Experiment Setup

## $\square$ NBA Dataset

- 317,371 tuples of NBA box scores from 1991-2004 seasons
- 8 dimension attributes
- 7 measure attributes
$\square$ Weather Dataset
- 7.8 million tuples of weather forecast from different locations of six countries \& regions of UK
- 7 dimension attributes
- 7 measure attributes


## Discovered Facts

> Lamar Odom had 30 points, 19 rebounds and 11 assists on March 6, 2004. No one before had a better or equal performance in NBA history.
$>$ Allen Iverson had 38 points and 16 assists on April 14, 2004 to become the first player with a $38 / 16$ (points/assists) game in the 2004-2005 season.
$>$ Damon Stoudamire scored 54 points on January 14, 2005. It is the highest score in history made by any Trail Blazers.
$A$

## Prominent Streaks

Prominent Streak Discovery in Sequence Data. Xiao Jiang, Chengkai Li, Ping Luo, Min Wang, Yong Yu. KDD 2011, pages 1280-1288.

Discovering General Prominent Streaks in Sequence Data. Gensheng Zhang, Xiao Jiang, Ping Luo, Min Wang, Chengkai Li. ACM TKDD, 8(2):article 9, June 2014.

## Prominent Streaks

## Prominent streaks stated in news articles:

"This month the Chinese capital has experienced 10 days with a maximum temperature in around 35 degrees Celsius - the most for the month of July in a decade."
"The Nikkei 225 closed below 10000 for the 12th consecutive week, the longest such streak since June 2009."
"He (LeBron James) scored 35 or more points in nine consecutive games and joined Michael Jordan and Kobe Bryant as the only players since 1970 to accomplish the feat."

## Concepts <br> Streak

Input: a sequence of values
Streak $\langle[I, r], v>$ is a triple: left-end ( $I$ ), right-end ( $r$ ), minimum value in interval $[I, r]$

$$
\begin{array}{lllllllll} 
& 3 & 7 & 7 & 2 & 5 & 6 & 7 & 3 \\
& <[6,8], 4> & & & &
\end{array}
$$

## Streak dominance relation

$$
\begin{gathered}
s 1=<[\mid 1, r 1], v 1>\text { dominates } s 2=<[\mid 2, r 2], v 2>\text { if and only if } \\
r 1-|1>r 2-| 2, v 1>=v 2 \text { or } r 1-|1>=r 2-| 2, v 1>v 2
\end{gathered}
$$

## Prominent streaks (PS)

A streak is prominent if it is not dominated by any other streaks.

## Example

## 3177254673



## Prominent Streaks are Skyline Points in 2-d Space

 3177254673

Each streak can be viewed as a point in a 2-dimension space (length, minimal_value). Prominent streaks are skyline points in this space.

## Tasks

## Task 1: discovery

Find all prominent streaks in a sequence

## Task 2: monitoring

Always keep prominent streaks up-to-date, when sequence grows (real-world sequences often grow)

## Solution Framework



## Candidate Generation: Number Of Candidate streaks

## Brute-force

Quadratic (How many streaks are in a sequence of $n$ values? These are all candidate streaks.)

Superlinear


Linear

Given a sequence of $n$ values, how many streaks are there? 'That's the number of candidate streaks the bruteforce algorithm will produce.
$n *(n+1) / 2$

- Streaks starting at the $1^{\text {st }}$ position: n
- Streaks start at the $2^{\text {nd }}$ position: $\mathrm{n}-1$
- Streaks start at the nth position: 1


## Local Prominent Streak

## Local dominance relation

$s 1=<[11, r 1], v 1>$ locally dominates $s 2=<[12, r 2], v 2>$ if and only if a) s1 dominates s2 and b) $[11, r 1] \supset[l 2, r 2]$ (i.e., s1 subsumes/contains $s 2$ and $s 2$ is thus a sub-sequence of $s 1$ ).

## Local prominent streak (LPS)

A streak is locally prominent if it is not locally dominated by any other streaks.


## Important Properties <br> (1) LPS is sufficient

A prominent streak must be an LPS.

## (2) LPS is small

The number of LPSs is less than or equal to the sequence length.
(Hint: The number of LPSs getting min value at position k is at most 1.)

## Conclusion

LPS is an excellent set of candidate streaks, of linear size.
The candidate generation problem becomes finding local prominent streaks.

Local prominent streaks (candidates)

prominent streaks (skyline points)


## Linear LPS (LLPS) Method

Given a sequence $p_{1}, p_{2}, \ldots, p_{n}, \ldots$, maintain a list of growing streaks when scanning the sequence rightward.

1. At $p_{1}$, create a growing streak to include the $1^{\text {st }}$ position.
2. After $p_{k}$, right-ends of growing streaks are all at position $k$.
3. At $p_{k+1}$, try to extend the growing streaks rightward. The growing streaks are partitioned by $p_{k+1}$ into two groups:
(3.a) Those growing steaks with $\mathrm{s.v}<=\mathrm{p}_{\mathrm{k}+1}$ : extend s to include position $\mathrm{k}+1$.
(3.b) Those growing streaks with s.v $>p_{k+1}$ : include $s$ into LPS; remove $s$ from growing streaks as it won't grow anymore.
4. At $p_{k+1}$, do one of the following, based on which condition is satisfied:
(4.a) There was a growing streak with s.v $=p_{k+1}$ : nothing more needs to be done for $p_{k+1}$.
(4.b) There were one or more growing streaks with s.v $>\mathrm{p}_{\mathrm{k}+1}$ : create a new growing streak by extending the leftmost (longest) such growing streak $s$ that satisfies s.v $>p_{k+1}$ to include position $k+1$.
(4.c) There was no growing streak with s.v $>=p_{k+1}$ : create a new growing streak to include just position $\mathrm{k}+1$.

## An illustration of the method at $\mathrm{p}_{10}$, with all remaining growing streaks highlighted



## Linear LPS (LLPS) Method

Growing streaks share the same right-end. Their minimum values monotonically increase, if they are listed in the increasing order of left-ends.

Two different ways of illustrating the method at $\mathrm{p}_{10}$, with all remaining growing streaks highlighted





## Linear LPS (LLPS) Method

Monitoring (keeping prominent streaks up-to-date) is simple:

- Keeps applying the algorithm when new values of the sequence are received.
- After the latest value at $\mathrm{p}_{\mathrm{n}}$, it has found all LPSs ending before position n.
- Growing streaks ending at n either will eventually be LPSs or can be grown into LPSs ending at a future position after $n$.
- If prominent streaks till n are requested, compare all found LPSs and all growing streaks.


## Datasets In Experiments


(a) Data Sequence
(b) Prominent Streaks

## Sample Prominent Streaks

 Melbourne daily min/max temperature between 1981 and 1990 (Melb1 \& Melb2)More than 2000 days with min temperature above zero 6 days: the longest streak above 35 degrees Celsius


## Traffic count of Wikipedia page of Lady Gaga (Wiki2)

More than half of the prominent streaks are around Sep. 12th (VMA 2010) at least 2000 hourly visits lasting for almost 4 days


## General Prominent Streaks

## Top-k, multi-dimensional and multi-sequence PS

"He (LeBron James) scored 35 or more points in nine consecutive games and joined Michael Jordan and Kobe Bryant as the only players since 1970 to accomplish the feat."
"Only player in NBA history to average at least 20 points, 10 rebounds and 5 assists per game for 6 consecutive seasons." (http://en.wikipedia.org/wiki/Kevin Garnett)

## NLPS/LLPS extended to such general PSs

## Experiments On Multi-Sequence PSs

Tat Table IX. Multi-sequence Prominent Streaks in Datast NBA1

| n11 |
| ---: |
| $\square$ |
| $\square$ |


| length | minimal value | players |
| :---: | :---: | :---: |
| 1 | 71 | David Robinson |
| 2 | 51 | Allen Iverson; Antawn Jamison |
| 4 | 42 | Kobe Bryant |
| 9 | 40 | Kobe Bryant |
| 13 | 35 | Kobe Bryant |
| 14 | 32 | Kobe Bryant |
| 16 | 30 | Kobe Bryant |
| 17 | 27 | Michael Jordan. |
| 27 | 26 | Allen Iversom |
| 34 | 24 | Tracy Mrecrady |
| 45 | 21 | Allen Iversom |
| 57 | 20 | Allen Iversom |
| 74 | 19 | Shaquille O'Neal |
| 94 | 18 | Shaquille O'Neal |
| 96 | 17 | Karl Malone |
| 119 | 16 | Karl Malone |
| 149 | 15 | Karl Malone |
| 159 | 14 | Karl Malone |
| 263 | 13 | Karl Malone |
| 357 | 12 | Karl Malone |
| 527 | 11 | Karl Malone |
| 575 | 10 | Karl Malone |
| 758 | 7 | Karl Malone |
| 858 | 6 | Shaquille O'Neal |
| 866 | 2 | Karl Malone |
| 932 | 1 | John Stockton |
| 1185 | 0 | Jim Jackson |



1 A

## Experiments On Multi-Dim PSs

Table X. Data Sequences Used in Experiments on Multi-dimensional Prominent Streak Discovery.

| name | length | \# prominent streaks | \# dimensions | description |
| :---: | :---: | :---: | :---: | :---: |
| Malone | 986 | 640 | 6 | $1991-2004$ game log of Karl Malone (minutes, points, <br> rebounds, assists, steals, blocks) |


(a) Number of Prominent Streaks

(b) Execution Time of LLPS

Fig. 13. Experiments on Increasing Dimensionality.

## Experiments On General PSs

Table XIII. Data Sequences Used in Experiments on Top-5 Multi-sequence Multi-dimensional Prominent Streak Discovery

| name | \# sequences | average length | \# dimensions | \# prominent streaks | description |
| :---: | :---: | :---: | :---: | :---: | :---: |
| NBA2 | 1185 | 290 | 6 | 10867 | 1991-2004 game log of all |
|  |  |  |  | N- |  |
|  |  |  |  | blayers (minutes, points, re- |  |
|  |  |  |  | bounds, assists, steals, blocks) |  |

Table XIV. Number of Candidate Streaks, Top-5 Multi-sequence Multi-dimensional Prominent Streak Discovery.

| name | Baseline | NLPS | LLPS |
| :---: | :---: | :---: | :---: |
| NBA2 | $9.41 \times 10^{7}$ | $2.98 \times 10^{6}$ | $8.76 \times 10^{5}$ |

Table XV. Execution Time (in Milliseconds), Top-5 Multi-sequence Multi-dimensional Prominent Streak Discovery.

| name | Baseline | NLPS | LLPS |
| :---: | :---: | :---: | :---: |
| NBA2 | $1.39 \times 10^{7}$ | $4.33 \times 10^{5}$ | $1.14 \times 10^{5}$ |



Fig. 14. Distribution of Prominent Streaks by Length.

## Exercise

Apply the LLPS algorithm to find prominent streaks in 4135473 . Show local prominent streaks and growing streaks at every position.

## One-of-the-few Objects

On "One of the Few" Objects. You Wu, Pankaj K. Agarwal, Chengkai Li, Jun Yang, Cong Yu. KDD 2012, pages 1487-1495

## One-Of-The-Few Claims

## Do these claims really hold water?

Karl Malone is ONE OF THE ONLY TWO players in NBA history with 25,000 points, 12,000 rebounds, and 5,000 assists in one's career.

He is ONE OF THE ONLY THREE candidates who have raised more than $25 \%$ from PAC contributions and $25 \%$ from self-financing.

## How do we find truly interesting claims or individuals?

## X Is One-Of-K $\rightarrow \mathrm{X}$ Is In K-Skyband

## Claim

Karl Malone is ONE OF THE ONLY TWO players in NBA history with 25,000 points, 12,000 rebounds, and 5,000 assists in one's career.

## General claim

Fewer than $k$ objects dominate $X$ in subspace of attributes $S \subseteq\left\{A \_1, A \_2, \ldots, A \_d\right\}$ $k$-skyband [Papadias et al. 2005] in $S$ is the set of points each dominated by fewer than $k$ other points in $S$
1-skyband: skyline


A

## Small K $\neq$ Interesting

## Subspaces are different

E.g., 2-skyand in \{rebounds\} vs. in \{rebounds, assists\}


îDiR

## Small K $\neq$ Interesting

## Data distribution matters

E.g., 2 -skyand in \{points, rebounds\} vs. in \{rebounds, assists\}



## Top- $\tau$ Skyband

## k-Skyband

Using the same $k$ for all subspaces doesn't work Asking user pick $k$ for each subspace is infeasible Top- $\tau$ Skyband


- User specifies a single parameter $\tau$ to cap \# skyband objects.
- For each subspace $S$, find its top- $\tau$ skyband, i.e., the largest $k$-skyband containing no more than $\tau$ objects
- E.g., in \{points, rebounds\}:
$\tau=\mathbf{2} \rightarrow 1$-skyband (size 2)
$\tau=\mathbf{6} \rightarrow 2$-skyband (size 5; 3-skyband would be too big)

