

LEGENDARY OF 18 WEAPONS - MOTION CAPTURE DATA ANALYSIS FOR WII REMOTE

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OUTLINE

○ Introduction

- Background information
- Motivation & Objective

○ The Wii Remote

- How to get data from Wii Remote ?
- What can be obtained from Wii Remote ?

○ Method of Classification

- Possible Classification Methods
- Working Principle of Dynamic Time Warping

○ How do we apply Dynamic Time Warping ?

○ Tuning

○ Future Work

○ Q & A

INTRODUCTION

BACKGROUND INFORMATION

◎ Legendary of 18 Weapons

- 18 major weapons used in traditional martial art



* Images are extracted from <http://www.kenchan.com.hk>

BACKGROUND INFORMATION

○ Motion capture

- Recording movements in a digital manner
- With the use of accelerometers

○ Wii Remote

- Has built-in accelerometers
- Capture movements of 1 hand
- Can be extended to both hands with the use of Nunchuk



MOTIVATION

- ◉ Wii is a revolutionary console
 - A new method of playing games
 - Costly to buy a console
- ◉ Wii Remote can pair with personal computers
 - Capture data using Wii Remote and send to computer
 - Players can still enjoy with specially-designed games



OBJECTIVES

- Construct a motion classifier
- Build an interface
 - A game, or an educational software
 - Perform motions in front of computer
 - Recognize correctly and be displayed in the computer

THE WII REMOTE

HOW TO GET DATA FROM WII REMOTE ?

- Bluetooth as the communication channel
- Pair up with Bluetooth USB adapter to communicate
- Use Wii Remote API get the data from Wii Remote
 - *WiiYourSelf!* as the API
 - Be integrated to our classifier



WHAT CAN BE OBTAINED FROM WII REMOTE?

- 3D Acceleration data

- x axis, y axis, z axis

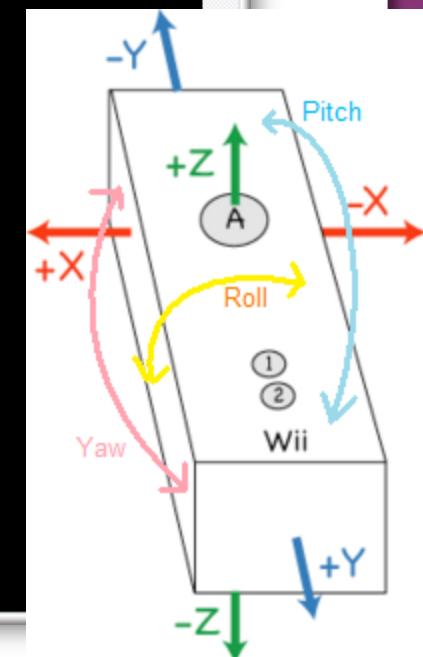
- Infra-red data

- Used with sensor bar
 - Determine pointing position



WHAT CAN BE OBTAINED FROM WII REMOTE?

```
ca - WiiYourself! - Demo:  
-WiiYourself!- library Demo: | <c> gl.tter 2007  
v0.96b | http://wiiyourself.gl.tter.org  
  
-- TRY: B = rumble, A = square, 1 = sine, 2 = daisy, Home = Exit --  
  
Battery: 47% LEDs: [--*--] RUMBLE using WriteFile()  
  
Buttons: [ ] [ ] [ ] [ ] [ ] [ ] [B|A] [ ] [ ]  
Accel: X -0.240 Y +0.074 Z +1.080  
Orient: UpdateAge 3 Pitch -3deg Roll 2deg  
<X +0.037 Y +0.068 Z +0.997>  
IR: Mode EXT. 0: Not found Size 0 X 0.000 Y 0.000  
1: Not found Size 0 X 0.000 Y 0.000  
2: Not found Size 0 X 0.000 Y 0.000  
3: Not found Size 0 X 0.000 Y 0.000  
Speaker: On | Frequency 3130 Hz Volume 0x20  
  
Extnsn.: None
```



METHOD OF CLASSIFICATION

POSSIBLE CLASSIFICATION METHODS

○ Tree-based & rule-based

- Compute attributes based on motion data
 - Relative Max / Min Acceleration Time
 - Initial / Average / End Acceleration
 - Rotation angle
- WEKA
 - Open source data mining software
 - Generate decision trees / rule with different tree-based / rule-based algorithms based on values of attributes
 - C4.5 decision tree learner



POSSIBLE CLASSIFICATION METHODS

○ Similarity search

- No further computations needed
- Compare patterns of motion data
- Euclidean Distance Algorithm
- Dynamic Time Warping

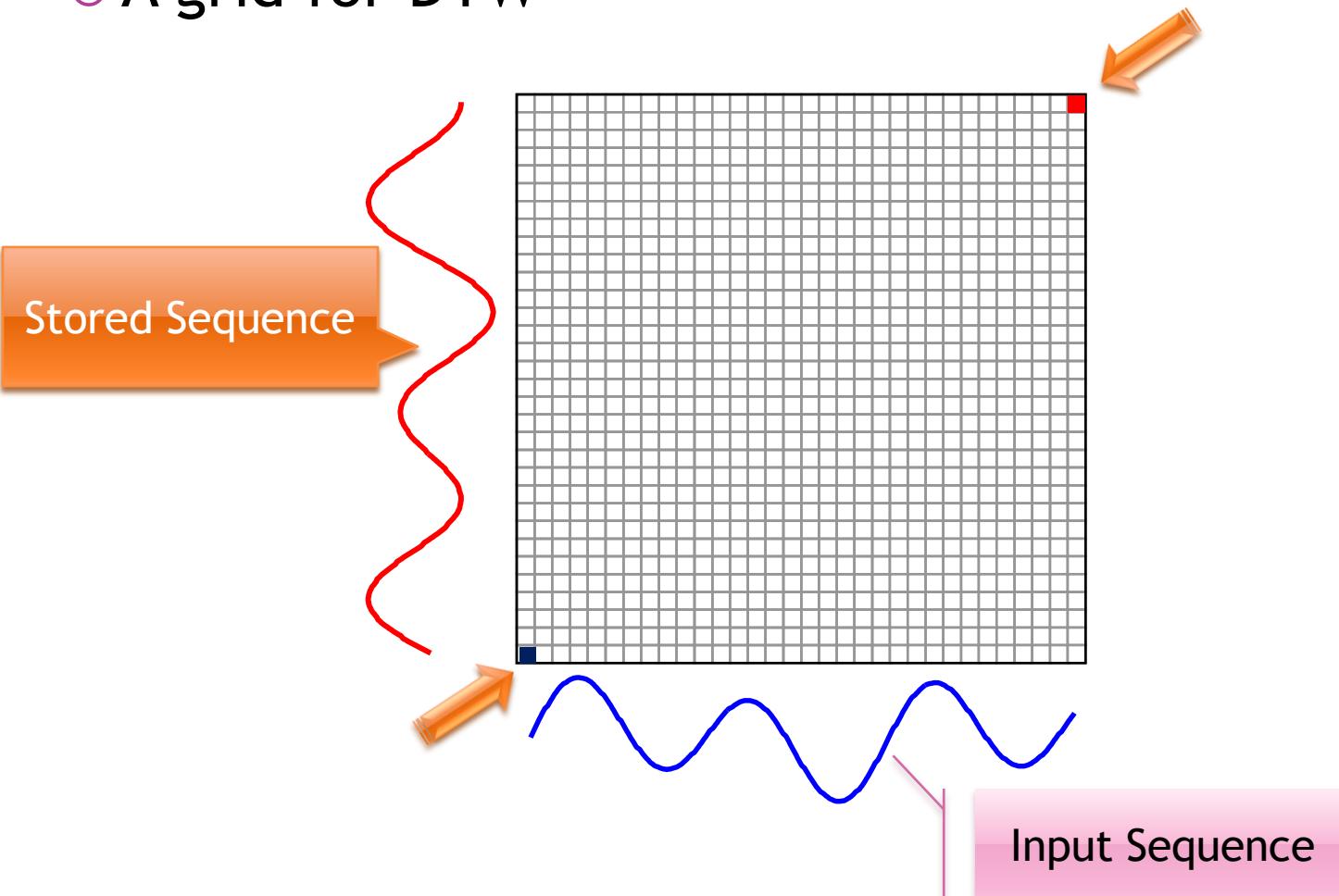


WORKING PRINCIPLE OF DYNAMIC TIME WARPING

- Find the best match between two given sequences.
- Sequences mapped non-linearly with the time dimension.
 - To detect similarity regardless of variations in time and speed
- To produce a matching
 - First model sequences into a 2-dimensional grid

WORKING PRINCIPLE OF DYNAMIC TIME WARPING

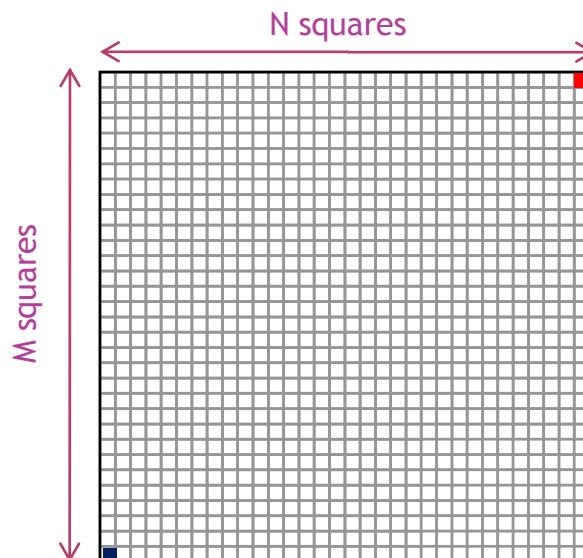
- A grid for DTW



WORKING PRINCIPLE OF DYNAMIC TIME WARPING

○ Dimension of grid

- Determined by the total number of data points of the sequence
- One data point is related with one row / column of squares



WORKING PRINCIPLE OF DYNAMIC TIME WARPING

- The next step is to compute the values for all the squares
- Each square stores a distance ...

- Defined by sum of 2 parts

$$\gamma(i,j) = d(s_i, p_j) + \min [\gamma(i-1, j-1), \gamma(i-1, j), \gamma(i, j-1)]$$

- The Euclidean distance between the corresponding elements of the two sequences
 - Minimum distances among the 3 neighboring squares

WORKING PRINCIPLE OF DYNAMIC TIME WARPING

- Euclidean distance is defined as the ordinary distance that could be measured with a ruler
 - For two points P and Q in general n-dimension space, it is defined as:

$$\sqrt{(p_1 - q_1)^2 + (p_2 - q_2)^2 + \dots + (p_n - q_n)^2} = \sqrt{\sum_{i=1}^n (p_i - q_i)^2}$$

- In particular for 3D space
 - For two 3D points, $P = (p_x, p_y, p_z)$ and $Q = (q_x, q_y, q_z)$, the Euclidean is defined as:

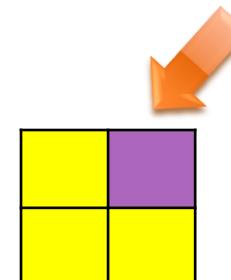
$$\sqrt{(p_1 - q_1)^2 + (p_2 - q_2)^2 + (p_3 - q_3)^2}$$

WORKING PRINCIPLE OF DYNAMIC TIME WARPING

- The next step is to compute the values of all the squares
- Each square stores a distance ...
 - Defined by sum of 2 parts

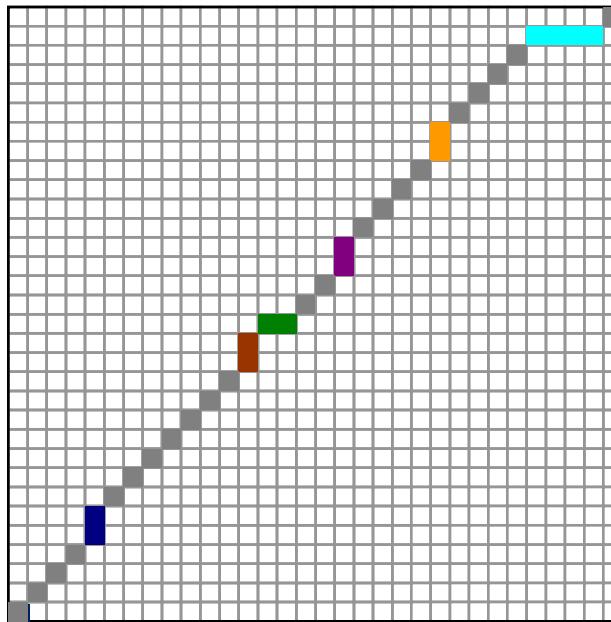
$$\gamma(i,j) = d(s_i, p_j) + \min [\gamma(i-1, j-1), \gamma(i-1, j), \gamma(i, j-1)]$$

- The Euclidean distance between the corresponding elements of the two sequences
- Minimum distances among the 3 neighboring squares
 - Values computed already



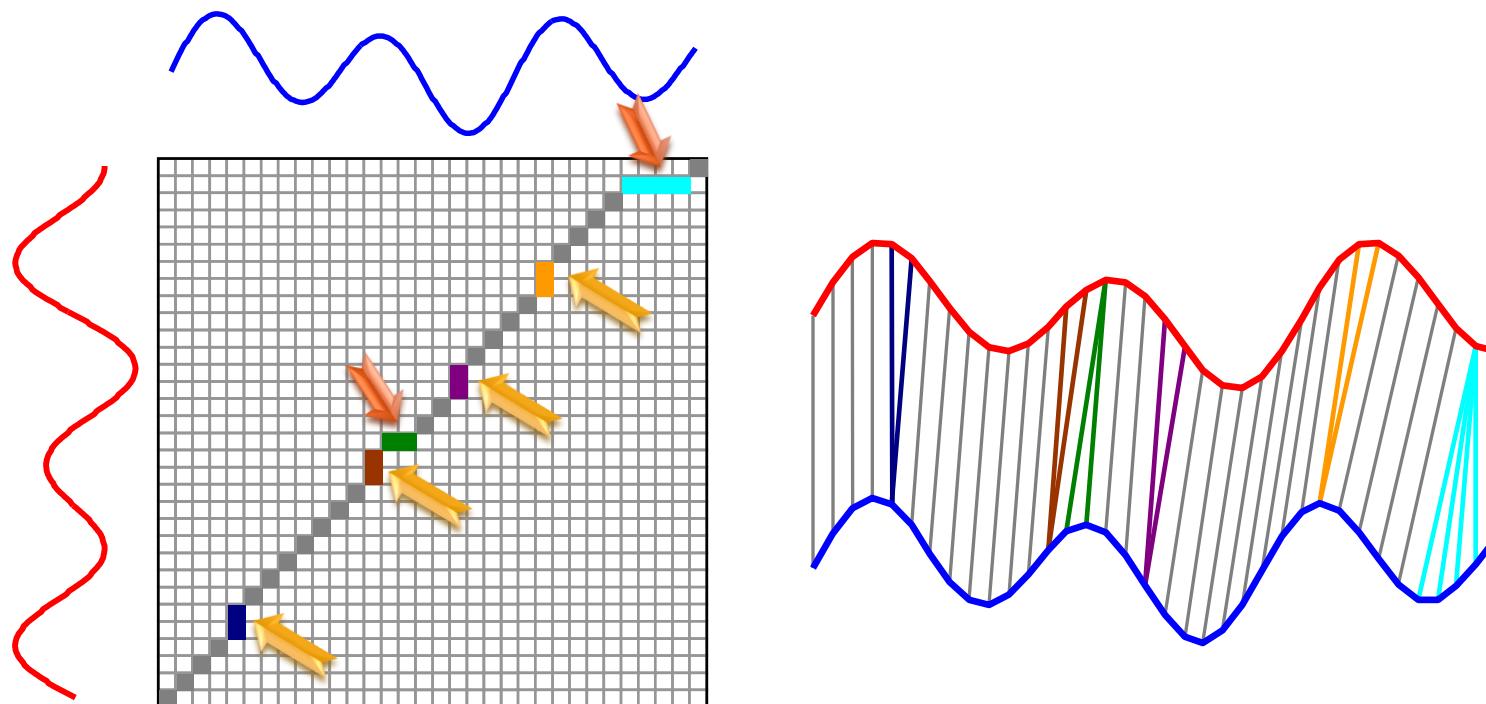
WORKING PRINCIPLE OF DYNAMIC TIME WARPING

- Next is to find a path within the grid which minimizes the total distance from the blue square to the red square



WORKING PRINCIPLE OF DYNAMIC TIME WARPING

- Warping path is found
 - Just the shortest path computed
- Defines the mapping between two sequences



HOW DO WE APPLY
DYNAMIC TIME
WARPING?

DATA STRUCTURE

Motion Database

2

A

Left

Reference
Motion
Data

Reference
Motion
Data

Reference
Motion
Data

Reference
Motion
Data

EXAMPLE

Time	X	Y	Z
0	1.000	2.519	0.480
15	0.720	2.667	0.320
30	0.120	2.963	0.240
45	-0.120	3.111	0.280
60	-0.600	3.519	0.440
75	-1.120	3.852	0.520
90	-1.280	4.000	0.640
105	-1.440	4.111	1.000
120	-1.680	4.148	1.160
135	-2.400	4.185	1.400
150	-2.880	4.185	1.440
165	-3.480	4.222	1.400
180	-3.280	4.222	1.280
195	-2.960	4.222	1.200
210	-2.480	4.222	1.000
225	-2.360	4.259	0.960
240	-2.360	4.259	0.840
255	-2.640	4.259	0.960
270	-3.400	4.259	1.000
285	-3.800	4.259	0.920
300	-3.480	4.296	0.840
315	-2.320	4.222	0.760
330	-2.040	4.259	0.920
345	-2.360	4.111	1.040
360	-2.520	3.519	0.840
375	-2.440	3.185	0.680
390	-2.000	2.630	0.400
405	-1.720	2.407	0.280

Time	X	Y	Z
0	1.400	1.963	0.760
15	1.240	2.111	0.760
30	0.880	2.593	0.720
45	0.440	2.926	0.760
60	0.240	3.074	0.760
75	-0.120	3.444	0.840
90	-0.360	3.667	0.840
105	-0.920	3.963	0.960
120	-1.320	4.074	0.920
135	-2.000	4.148	1.000
150	-2.520	4.185	1.200
165	-2.880	4.222	1.240
180	-3.240	4.222	1.240
195	-4.120	4.222	1.400
210	-4.280	4.222	1.240
225	-3.600	4.259	1.000
240	-2.840	4.259	1.000
255	-2.760	4.259	1.080
270	-3.120	4.296	1.280
285	-3.720	4.259	1.320
300	-4.600	4.296	1.000
315	-3.960	3.926	0.560
330	-3.360	3.593	0.440
345	-2.240	3.519	0.600
360	-1.960	3.593	0.760
375	-1.800	3.481	0.960
390	-1.880	3.222	1.000
405	-2.000	2.630	0.760
420	-1.920	2.370	0.600
435	-1.680	1.926	0.400
450	-1.520	1.704	0.320

Reference Motion of Same Type

Motion To Classify

THE INPUT DATA

Time (Millisecond)	X	Y	Z
0	1.00	2.52	0.48
15	0.72	2.67	0.32
30	0.12	2.96	0.24
45	-0.12	3.11	0.28
60	-0.60	3.52	0.44
75	-1.12	3.85	0.52
90	-1.28	4.00	0.64
105	-1.44	4.11	1.00
120	-1.68	4.15	1.16
135	-2.40	4.19	1.40
150	-2.88	4.19	1.44

APPLYING DYNAMIC TIME WARPING

Dynamic Time Warping Cost:

$$\gamma(i,j) = d(s_i, p_j) + \min [\gamma(i-1, j-1), \gamma(i-1, j), \gamma(i, j-1)]$$

Typically:

$$d(s_i, p_j) = |s[i] - p[j]|$$

In our case:

$$d(s_i, p_j) = \sqrt{(s[i].X - p[j].X)^2 + (s[i].Y - p[j].Y)^2 + (s[i].Z - p[j].Z)^2}$$

THE DTW MATRIX

0	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞	∞				
∞	0.7	1.3	1.6	2.3	3.3	4.8	6.6	9.1	11.9	15.3	19.3	23.6	28.2	33.7	39.3	44.3	48.5	52.7	57.3	62.4	68.3	73.4	77.9	81.3	84.5	87.5	90.5	93.5	96.4	99.1	101.8
∞	1.8	1.6	1.7	2.2	2.9	4.2	5.7	7.9	10.5	13.6	17.3	21.4	25.7	30.9	36.2	40.9	44.8	48.7	53.0	57.8	63.4	68.3	72.5	75.6	78.4	81.2	83.9	86.7	89.3	91.9	94.3
∞	3.5	3.1	2.6	2.3	2.7	3.5	4.5	6.2	8.1	10.6	13.7	17.1	20.8	25.4	30.1	34.1	37.4	40.7	44.3	48.5	53.5	57.7	61.3	63.7	65.9	68.1	70.2	72.4	74.6	76.7	78.7
∞	5.5	4.9	3.8	3.1	2.9	3.3	4.2	5.5	7.2	9.5	12.2	15.4	18.8	23.1	27.5	31.3	34.3	37.3	40.7	44.6	49.3	53.2	56.5	58.7	60.6	62.5	64.4	66.4	68.4	70.3	72.3
∞	8.0	7.2	5.6	4.3	3.9	3.6	3.8	4.6	5.6	7.3	9.4	11.9	14.8	18.5	22.3	25.5	27.9	30.3	33.0	36.4	40.5	43.9	46.6	48.3	49.7	51.0	52.4	54.1	55.9	57.8	59.8
∞	11.2	10.2	7.9	6.2	5.5	4.7	4.4	4.3	4.8	5.9	7.5	9.4	11.7	14.8	18.1	20.6	22.5	24.3	26.4	29.2	32.7	35.6	37.8	39.0	39.9	40.8	41.9	43.4	45.1	47.1	49.3
∞	14.6	13.3	10.5	8.2	7.3	6.0	5.4	4.8	4.6	5.4	6.8	8.5	10.6	13.5	16.6	19.0	20.6	22.2	24.1	26.7	30.0	32.7	34.8	35.9	36.7	37.5	38.6	40.1	41.9	44.0	46.3
∞	18.1	16.7	13.3	10.4	9.3	7.5	6.6	5.4	4.8	5.2	6.3	7.8	9.6	12.3	15.1	17.3	18.7	20.0	21.8	24.1	27.2	29.8	31.9	32.9	33.7	34.4	35.4	37.0	38.9	41.1	43.7
∞	21.9	20.2	16.3	12.9	11.5	9.2	8.0	6.2	5.2	5.1	6.0	7.2	8.7	11.2	13.8	15.7	16.9	18.0	19.4	21.5	24.4	26.8	28.7	29.7	30.5	31.2	32.1	33.7	35.6	37.9	40.5
∞	26.4	24.5	20.0	16.1	14.5	11.7	10.2	7.7	6.4	5.7	5.4	5.9	6.7	8.4	10.3	11.6	12.2	12.7	13.4	14.8	17.0	18.8	20.3	21.3	22.3	23.3	24.5	26.2	28.3	30.8	33.7
∞	31.2	29.1	24.2	19.7	17.9	14.6	12.9	9.8	8.0	6.7	5.8	5.6	6.0	7.2	8.6	9.5	9.9	10.3	10.6	11.5	13.3	14.7	15.9	17.2	18.5	19.8	21.3	23.2	25.4	28.2	31.2
∞	36.7	34.4	28.9	23.9	21.8	18.1	16.1	12.4	10.3	8.2	6.8	6.2	5.9	6.5	7.3	7.7	8.5	9.3	9.7	9.9	11.1	12.1	13.3	14.9	16.7	18.6	20.5	22.8	25.3	28.4	31.6
∞	41.9	39.4	33.4	27.9	25.5	21.4	19.1	14.8	12.3	9.5	7.5	6.6	5.9	6.7	7.5	7.7	8.3	8.8	9.0	9.4	10.8	11.8	12.9	14.3	15.9	17.5	19.3	21.4	23.8	26.7	29.9
∞	46.8	44.1	37.6	31.5	29.0	24.4	21.8	16.8	13.9	10.5	8.0	6.7	6.2	7.1	8.0	8.2	8.8	9.2	10.8	12.0	12.9	14.0	15.3	16.7	18.2	20.1	22.3	25.1	28.1		
∞	51.3	48.4	41.3	34.7	31.9	26.8	24.0	18.4	15.1	11.0	8.2	7.1	7.0	7.9	8.9	9.1	8.3	8.3	8.9	9.7	11.3	12.4	13.2	13.7	14.6	15.6	16.8	18.4	20.4	22.9	25.7
∞	55.7	52.6	45.0	37.8	34.8	29.2	26.0	19.9	16.2	11.4	8.5	7.7	7.9	8.8	9.8	10.1	8.8	8.7	9.1	10.3	11.9	13.0	13.7	14.1	14.5	15.5	16.6	18.3	20.3	22.8	25.5
∞	60.1	56.8	48.6	40.9	37.6	31.6	28.1	21.4	17.2	11.8	8.9	8.4	8.7	9.8	10.8	11.1	9.3	9.2	9.6	10.5	12.6	13.6	14.2	14.5	14.8	15.5	16.6	18.3	20.3	22.7	25.5
∞	64.8	61.2	52.5	44.3	40.8	34.3	30.5	23.1	18.6	12.4	9.2	8.8	9.1	10.3	11.4	11.7	9.5	9.3	9.7	10.7	12.5	13.9	14.7	15.2	15.5	16.0	16.8	18.4	20.4	22.8	25.6
∞	70.1	66.3	57.1	48.4	44.6	37.6	33.6	25.6	20.6	13.8	10.1	9.3	9.1	9.9	10.8	11.0	10.1	10.0	9.7	10.2	11.4	12.2	13.0	14.5	16.1	17.3	17.8	18.9	20.8	23.3	26.1
∞	75.8	71.8	62.1	52.8	48.8	41.4	37.1	28.5	23.1	15.6	11.4	10.3	9.7	9.6	10.2	10.4	11.1	11.0	10.5	10.1	10.9	11.4	12.4	14.1	16.1	18.2	19.4	20.3	21.6	24.0	26.8
∞	81.2	77.0	66.8	57.0	52.7	44.9	40.3	31.1	25.3	17.1	12.4	11.0	10.2	10.5	10.5	10.4	11.1	11.8	11.1	10.7	11.3	11.6	12.3	13.8	15.4	17.3	19.2	21.5	22.7	24.6	27.3
∞	85.6	81.1	70.4	60.0	55.5	47.2	42.3	32.5	26.3	17.6	12.9	11.8	11.2	12.1	12.5	11.7	11.0	11.5	12.0	12.2	13.0	12.9	13.0	13.7	14.6	15.8	17.4	19.3	21.7	24.4	
∞	89.7	85.1	73.8	62.8	58.1	49.3	44.1	33.7	27.1	17.7	13.5	12.7	12.5	13.3	14.3	13.3	11.8	11.7	12.7	13.7	14.7	14.9	14.4	13.7	14.5	15.6	17.2	19.1	21.5	24.2	
∞	94.0	89.2	77.3	65.9	60.9	51.7	46.1	35.1	28.1	18.1	13.7	13.2	13.4	14.3	15.3	14.5	12.3	12.2	12.5	13.9	16.0	16.4	15.7	14.4	14.5	15.5	17.1	18.9	21.3	23.9	
∞	98.3	93.2	80.9	68.9	63.7	54.1	48.3	36.8	29.4	18.9	14.5	14.1	14.3	15.2	16.2	15.9	13.1	13.0	13.2	14.0	16.2	17.5	16.6	14.8	15.0	15.1	15.3	16.3	17.6	19.5	21.6
∞	102.3	97.0	84.2	71.8	66.4	56.4	50.5	38.5	30.9	20.0	15.6	15.4	15.6	16.4	17.4	17.5	14.3	14.2	14.4	15.0	16.5	17.9	17.7	15.2	15.4	15.7	15.8	16.0	16.9	18.4	20.2
∞	105.8	100.3	87.1	74.3	68.7	58.5	52.4	40.3	32.6	21.6	17.4	17.4	18.6	18.4	19.3	19.7	16.2	16.1	16.4	16.9	18.2	18.8	19.3	16.1	16.2	16.5	16.6	16.1	16.3	17.1	18.1
∞	109.0	103.4	89.8	76.6	70.8	60.5	54.4	42.2	34.4	23.5	19.6	19.8	19.9	20.8	21.7	22.1	18.5	18.4	18.6	19.3	20.5	20.9	20.9	17.4	17.4	17.5	17.6	16.7	16.5	16.8	17.5

TAKE DURATION INTO CONSIDERATION

Duration (Millisecond)	
Reference Motion of Same Type	405
Motion to Classify	450
Average	427.5

7.2	19.1	21.0	24.2
7.1	18.9	21.3	23.9
5.3	17.6	19.5	21.6
5.0	16.9	18.4	20.2
5.1	16.3	17.1	18.1
5.7	16.5	16.8	17.5

$$\text{DTW Error(Similarity)} = 17.534 \div 427.5 = 0.041$$

DTW ERROR RESULT

Motion Name	Error
Left	0.041
TopRightCircle	0.245
LeftCircle	0.196
Dash	0.273
CircleUp	0.162

Classified As: Left

TUNING

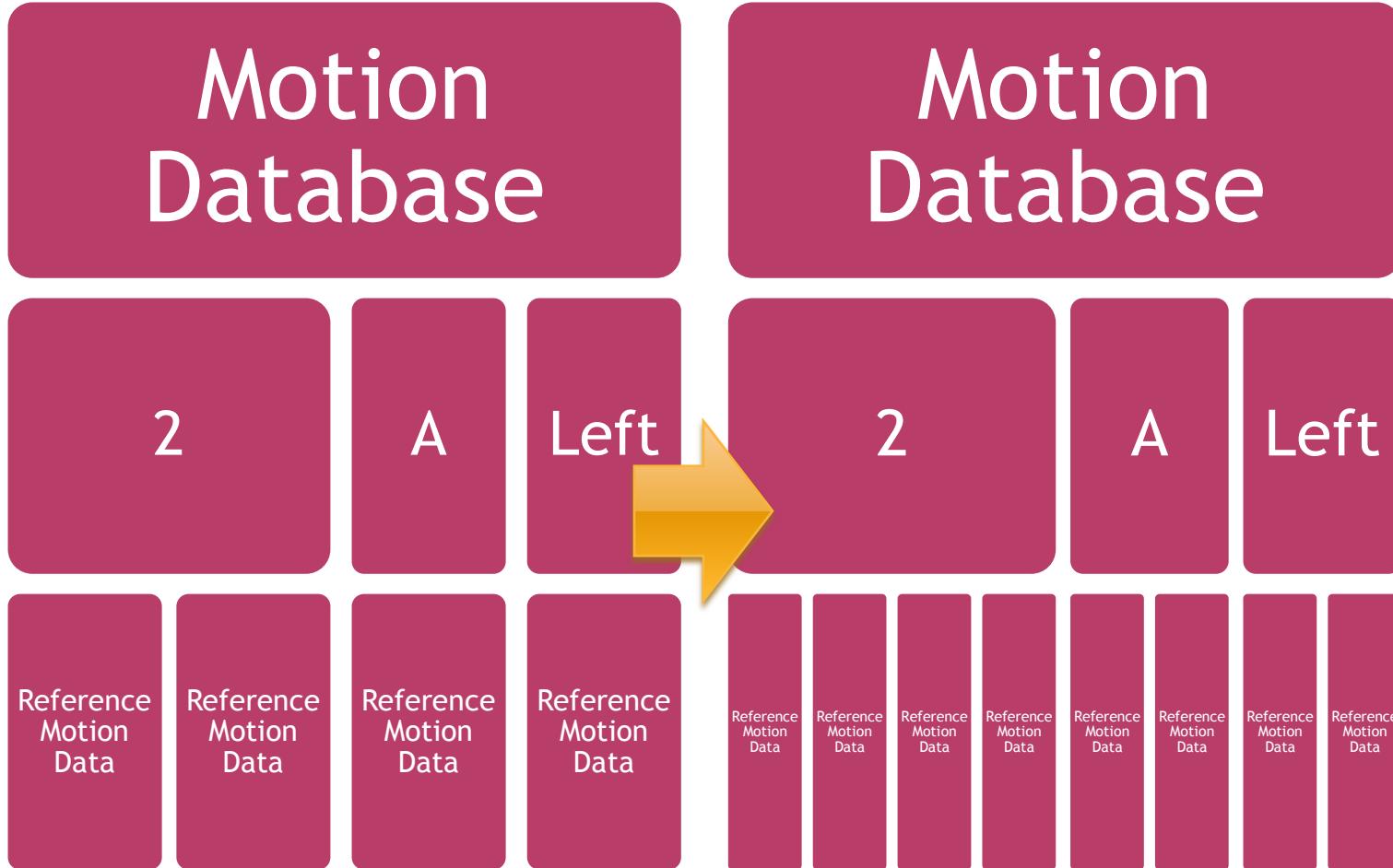
REASONS FOR LOW ACCURACY?

- Performed not in the same way
- Button pressed in wrong moment
- Similar motions
- Bad quality of reference motion

ANALYZED METHODS OF TUNING

- Number of Reference Motion
- Error Compensation
- Average of Motion Data as Reference Motion

NUMBER OF REFERENCE MOTION

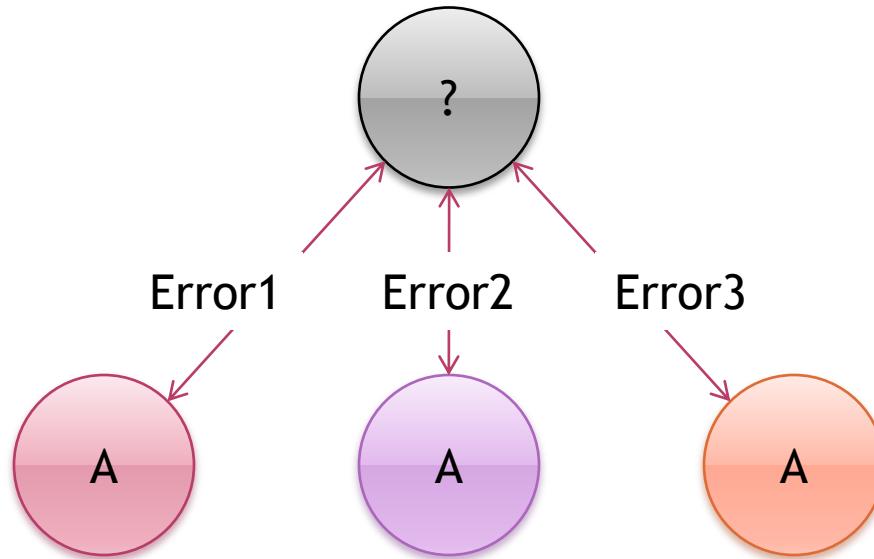


NUMBER OF REFERENCE MOTION

- Increase number of reference motions of the same type of motion
- Run Dynamic Time Warping on more reference motions

NUMBER OF REFERENCE MOTION

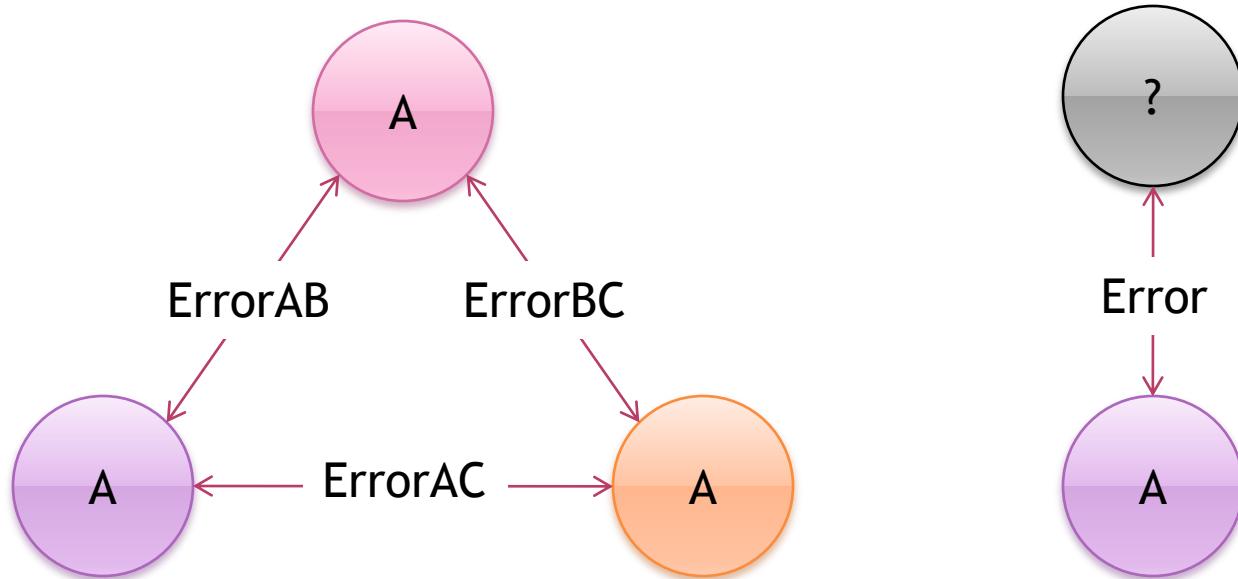
- `float dtwError(Motion A, Motion B);`



- $\text{Similarity} = (\text{Error1} + \text{Error2} + \text{Error3}) \div 3$

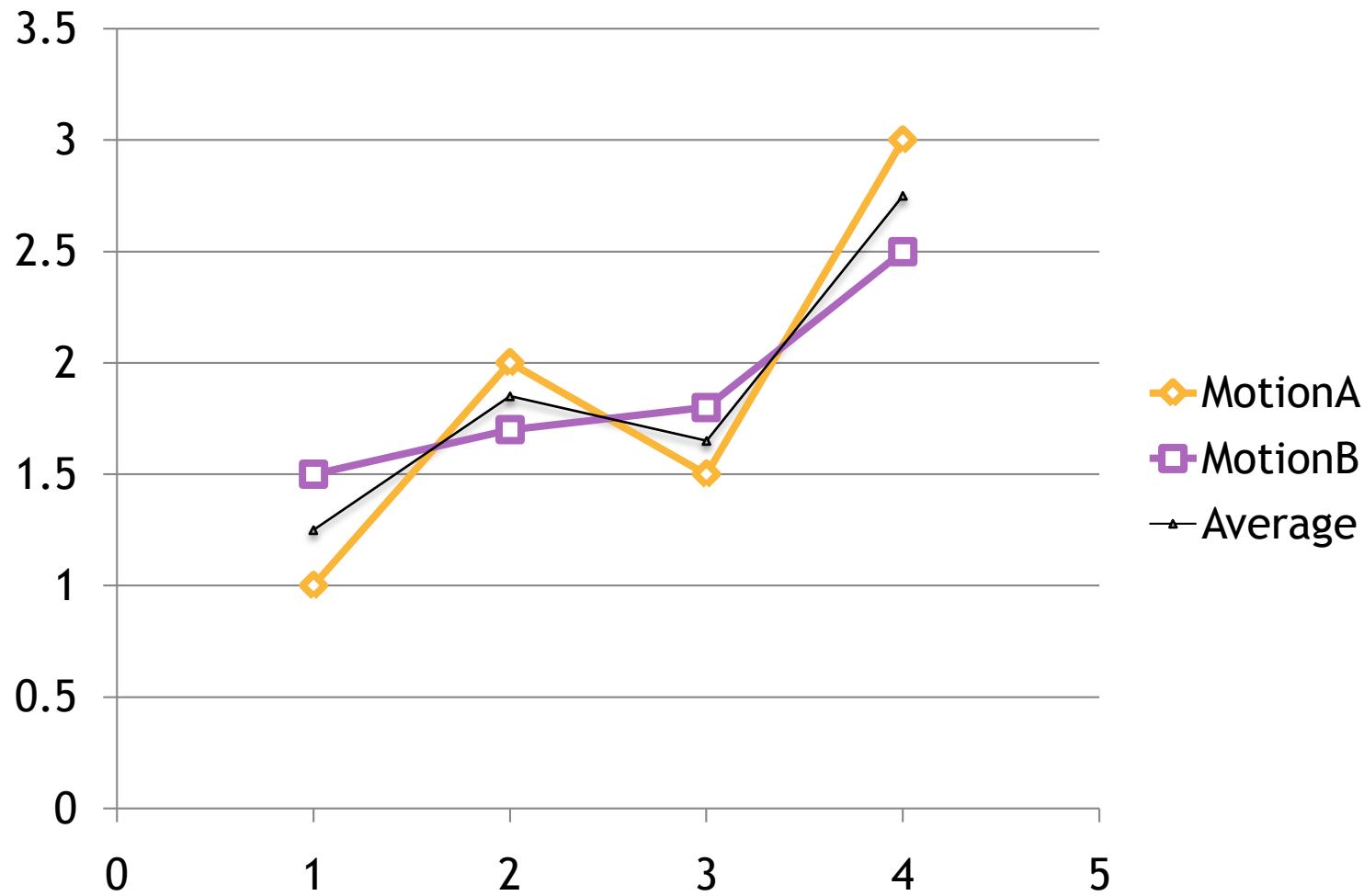
ERROR COMPENSATION

- float dtwError(Motion A, Motion B);



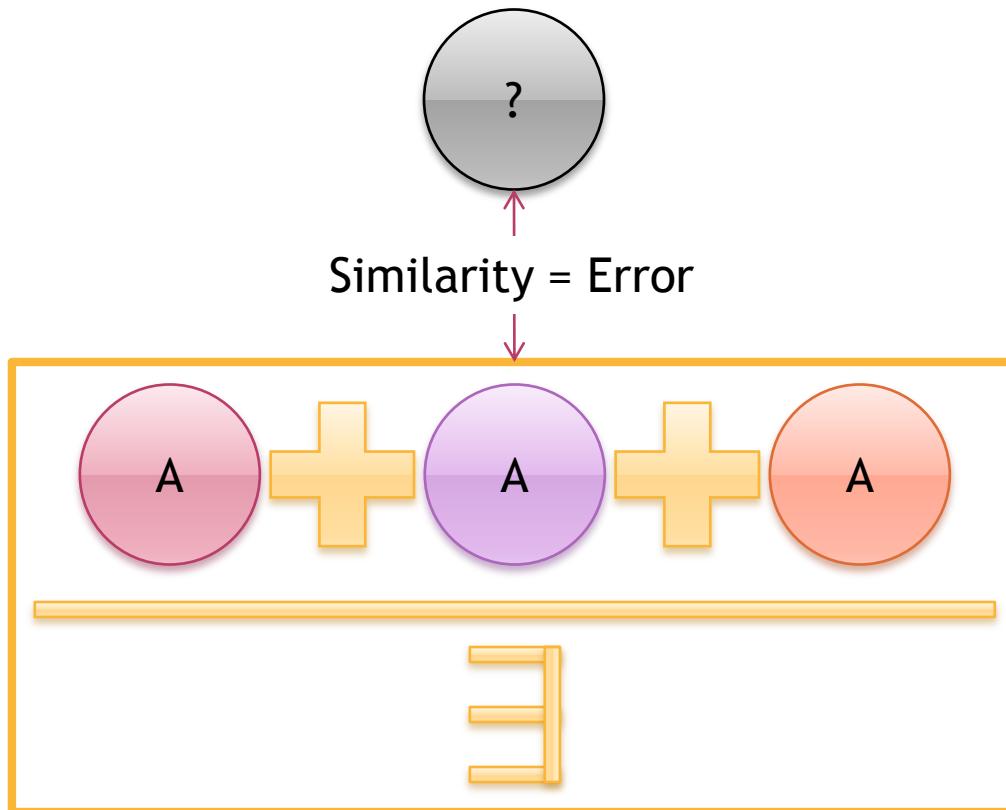
- AllowedError = (ErrorAB+ErrorBC+ErrorAC)÷3
- Similarity = Error - AllowedError

AVERAGE OF MOTION DATA AS REFERENCE MOTION



AVERAGE OF MOTION DATA AS REFERENCE MOTION

- `float dtwError(Motion A, Motion B);`



MOTIONS USED

- Five Directions
- Cursive Writing Letters
- Discrete Motions
- Numbers
- Dao Motions

CONFIGURATIONS USED

	Reference Motion (R)	Error Compensation (E)	Average Motion (A)
Configuration 1	1		
Configuration 2	1	3	
Configuration 3	3		
Configuration 4	3	3	
Configuration 5	5		
Configuration 6	5	3	
Configuration 7			3
Configuration 8		3	3

CLASSIFICATION ACCURACY

Configuration	Five Directions (By 3 People)	Letter	Discrete	Dao Motions	Number (By 3 People)	Average
R=1	92.57%	100%	99.51%	100%	85.69%	95.55%
R=1,E=3	95.59%	100%	98.52%	100%	90.41%	96.90%
R=3	91.06%	100%	99.50%	100%	95.47%	97.21%
R=3,E=3	91.92%	100%	99.50%	100%	98.13%	97.91%
R=5	94.18%	100%	99.50%	100%	97.07%	98.15%
R=5,E=5	94.90%	100%	99.50%	100%	98%	98.48%
A=3	91.52%	100%	99.50%	100%	97.47%	97.70%
E=3,A=3	92.66%	100%	99.50%	99.29%	90.41%	96.37%

EFFECT OF DIFFERENT TUNING

	Preparation?	Time?	Space?
Reference Motion	No	Largely	Largely
Error Compensation	Yes	Slightly	Slightly
Average Motion	Yes	No Effect	No Effect

CONCLUSION

- ⦿ Depend on requirement in choosing
- ⦿ Hybrid can be considered

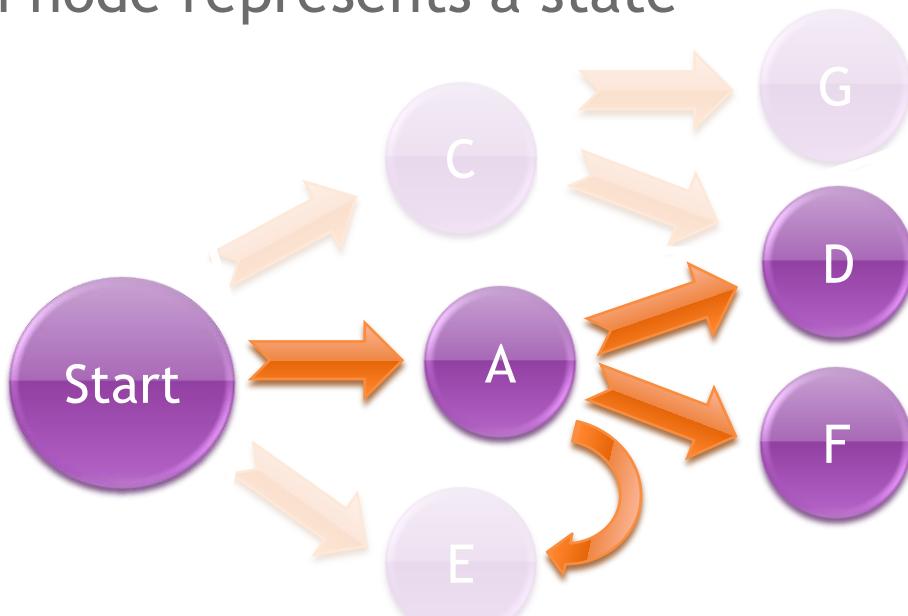
FUTURE WORK

FUTURE WORK

- A game demo
- Educational software of Chinese Martial Art
- Motion Finite State Machine
 - Making use of game rules to minimize the amount of possible candidates in classification
 - Limit the type of motions that can be followed

FUTURE WORK

- Motion Finite State Machine
 - Each node represents a state



- Allow players to perform motions in any sequence but still maintain high accuracy

FUTURE WORK

- A game demo
- Educational software of Chinese Martial Art
- Motion Finite State Machine
- Early Recognition
- Two hand weapons using Nunchuk
- “Button-less”

Q & A