Multidimensional scaling as regression analysis

#### Okayama University Graduate School Takashi Shindo

## Outline

- An overview of multidimensional scaling (MDS)
- An overview of feature matching model (FMM)
- A constrained MDS as an FMM
  - Feature matching MDS (FM-MDS)
- An FM-MDS as a regression analysis
- Further obstacles

# Multidimensional scaling (MDS)

• A method of transforming (dis)similarity data into an arrangement.

- Input is a set of (dis)similarity data between objects.
- Output is coordinates of objects.

• In this presentation, only dissimilarity is considered.

#### Use of MDS

• The properties of objects can be inferred by searching for meaningful axes in the obtained arrangement.

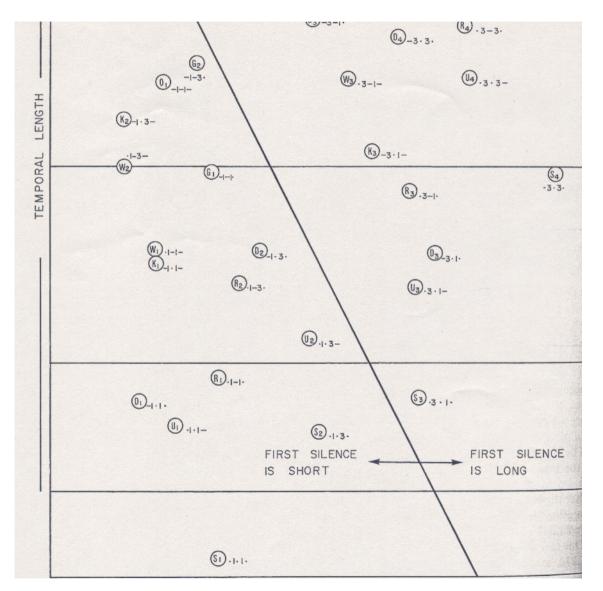
Interpretation of the arrangement may be arbitrary.

#### An example of MDS (input)

	1.1.1.1.1.1.1.1					Contract of the								10356187		
		s <sub>1</sub>	U_1	R <sub>1</sub>	D <sub>1</sub>	<b>W</b> <sub>1</sub>	<b>к</b> 1	°,	°1	s2	U2	R2	<sup>D</sup> 2	w2	ĸ2	G2
• • • • •	s <sub>1</sub>	97	94	81	66	16	31	06	03	79	18	28	11	01	04	02
•   •   =	U1	63	98	86	74	91	73	23	16	57	61	49	21	13	09	06
•1-1 •	R <sub>1</sub>	27	73	94	50	72	62	51	22	42	52	64	49	30	11	16
-1 • 1 •	*	47	75	69	96	33	94	67	09	43	13	40	34	11	26	19
• 1 - 1 -	W1	03	40	64	42	94	76	68	70	20	51	73	40	69	49	54
-1+1-	K <sub>1</sub>	07	44	33	69	69	95	70	67	19	37	40	56	26	74	27
-1-1-	G1	01	08	49	50	78	78	93	82	08	14	48	80	23	53	90
-1-1-	°1	02	10	19	04	31	51	76	97	00	04	12	16	20	50	82
• 1 • 3 •	s <sub>2</sub>	44	54	62	56	32	41	19	04	94	84	62	54	38	22	07
• 1 • 3-	U2	03	39	44	15	66	61	29	19	63	94	39	52	77	48	14
•1-3•	R <sub>2</sub>	05	16	48	15	59	19	45	28	31	38	95	47	86	36	53
-1.3.	D <sub>2</sub>	06	14	28	46	26	51	69	43	47	39	43	94	35	79	73
•1-3-	W2	02	14	24	09	62	39	21	59	07	54	69	37	94	64	56
-1 • 3-	K.2	01	03	05	07	22	54	37	56	14	19	23	68	48	98	47
-1-3.	G2	01	05	10	08	30	28	71	73	08	07	46	57	44	42	93
-1-3-	°2	00	01	04	03	21	08	35	80	00	04	07	24	55	58	58
• 3 • 1 •	s <sub>3</sub>	42	38	50	48	23	22	19	06	69	53	31	33	09	03	05
•3•1-	U3	04	09	30	18	42	52	35	29	30	29	14	21	30	22	05
.3-1.	R <sub>3</sub>	05	05	43	08	22	24	38	20	10	17	40	31	28	31	24
-3.1.	D.													~		

Wish, M., (1967). A model for the perception of Morse code-like signals. *Human Factors*, **9**, 529-540.

### An example of MDS (output)



Wish, M., (1967). A model for the perception of Morse code-like signals. Human Factors, 9, 529-540.

## Feature matching model (FMM)

• A regression analysis to explain (dis)similarity with common and distinctive features.

Interpretation of the result is clear and not arbitrary.

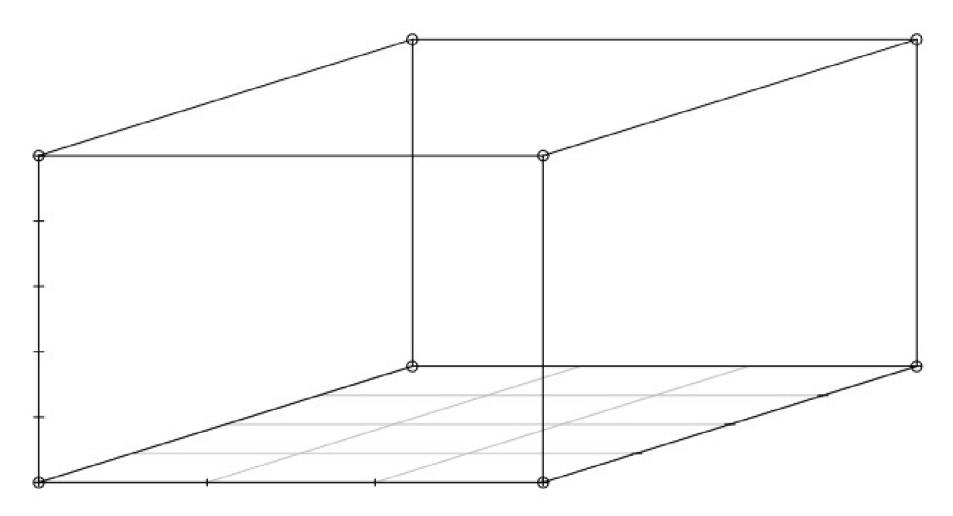
• In this presentation, we consider the case to explain dissimilarity with distinctive features.

## Feature matching MDS (FM-MDS)

• We fix the dimension of an arrangement and the order of objects for each dimension.

• By this constraint, each dimension of an arrangement is forced to correspond to a feature.

#### Idea of FM-MDS



### Use of FM-MDS for regression

- We can apply FM-MDS to differences on a scale for regression.
- If we use L1 norm, FM-MDS is equivalent to Hayashi's type I quantification method (Hayashi, 1952).
- A Minkowski norm of higher order is adequate when more contributive features are more dominant to dissimilarity.

## Meaning of dominance

- If the maximum norm is used, dissimilarity between a pair of objects is determined by each own feature.
- The idea of sparseness supposes the situation in which a property is determined by a few features.
- Sparseness is assumed to be valid in the field of genetic epidemiology.

#### Further obstacles

• An efficient solver for FM-MDS is required.

- Risks must be estimated precisely.
  - We need to estimate risks.
  - If sample size << dimension, subjects they have the same set of genotypes will be rare.
  - We cannot use the absolute values of odds ratios.