# Visual perceptual learning modulates decision network in the human brain: The evidence from psychophysics, modeling, and functional magnetic resonance imaging

Ke Jia	School of Psychological and Cognitive Sciences and Beijing Key Laboratory of Behavior and Mental Health, Peking University, Beijing, China PKU-IDG/McGovern Institute for Brain Research, Peking University, Beijing, China Key Laboratory of Machine Perception (Ministry of Education), Peking University, Beijing, China
Xin Xue	Department of Health Industry Management, Beijing International Studies University, Beijing, China School of Psychological and Cognitive Sciences and Beijing Key Laboratory of Behavior and Mental Health, Peking University, Beijing, China PKU-IDG/McGovern Institute for Brain Research, Peking University, Beijing, China Key Laboratory of Machine Perception (Ministry of Education), Peking University, Beijing, China
Jong-Hwan Lee	Department of Brain and Cognitive Engineering, Korea University, Seoul, Republic of Korea
	School of Psychological and Cognitive Sciences and Beijing Key Laboratory of Behavior and Mental Health, Peking University, Beijing, China PKU-IDG/McGovern Institute for Brain Research, Peking University, Beijing, China Key Laboratory of Machine Perception (Ministry of Education), Peking University, Beijing, China
Fang Fang	Peking-Tsinghua Center for Life Sciences, Peking University, Beijing, China
Jiaxiang Zhang	School of Psychology, Cardiff University, Cardiff, UK
	School of Psychological and Cognitive Sciences and Beijing Key Laboratory of Behavior and Mental Health, Peking University, Beijing, China PKU-IDG/McGovern Institute for Brain Research, Peking University, Beijing, China
Sheng Li	Key Laboratory of Machine Perception (Ministry of Education), Peking University, Beijing, China

Citation: Jia, K., Xue, X., Lee, J.-H., Fang, F., Zhang, J., & Li, S. (2018). Visual perceptual learning modulates decision network in the human brain: The evidence from psychophysics, modeling, and functional magnetic resonance imaging. *Journal of Vision*, *18*(12):9, 1–19, https://doi.org/10.1167/18.12.9.

https://doi.org/10.1167/18.12.9

Received March 12, 2018; published November 15, 2018

ISSN 1534-7362 Copyright 2018 The Authors

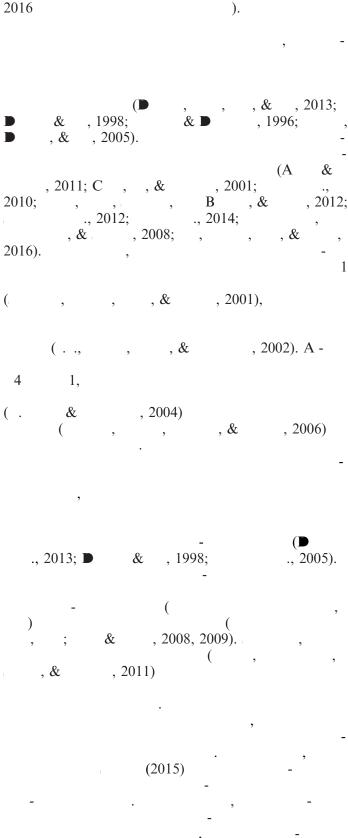
 $\square$ 

Perceptual learning refers to improved perceptual performance after intensive training and was initially suggested to reflect long-term plasticity in early visual cortex. Recent behavioral and neurophysiological evidence further suggested that the plasticity in brain regions related to decision making could also contribute to the observed training effects. However, how perceptual learning modulates the responses of decisionrelated regions in the human brain remains largely unknown. In the present study, we combined psychophysics and functional magnetic resonance imaging (fMRI), and adopted a model-based approach to investigate this issue. We trained participants on a motion direction discrimination task and fitted their behavioral data using the linear ballistic accumulator model. The results from model fitting showed that behavioral improvement could be well explained by a specific improvement in sensory information accumulation. A critical model parameter, the drift rate of the information accumulation, was correlated with the fMRI responses derived from three spatial independent components: ventral premotor cortex (PMv), supplementary eye field (SEF), and the frontoparietal network, including intraparietal sulcus (IPS) and frontal eye field (FEF). In this decision network, we found that the behavioral training effects were accompanied by signal enhancement specific to trained direction in PMv and FEF. Further, we also found direction-specific signal reduction in sensory areas (V3A and MT+), as well as the strengthened effective connectivity from V3A to PMv and from IPS to FEF. These findings provide evidence for the learning-induced decision refinement after perceptual learning and the brain regions that are involved in this process.

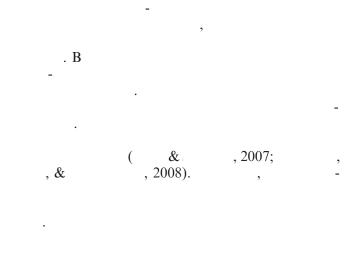
#### Introduction

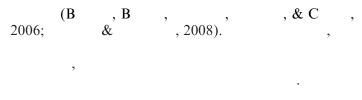
& 2015).	, 1994; , & ,	, , , & , 2014;	, & C	, 2001; , 2010; & ,	
, B & 1996;	, , , , 1987; & B	(A & & , 1997; , 1980; ,	& &	, 1997; , 1991). ,	
2008;	., 2010;	. 2015 . 201	( & 5 :	, 2014; &	

```
Jia et al.
```



(2018)





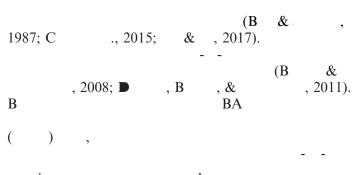








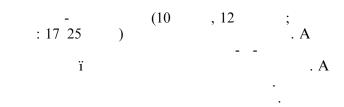




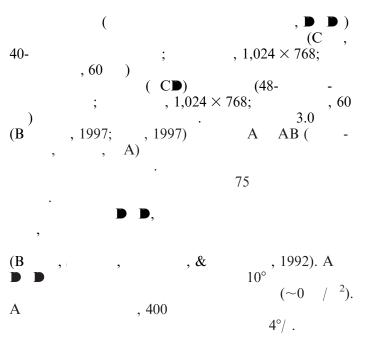
### Materials and methods



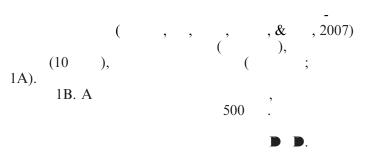
Jia et al.

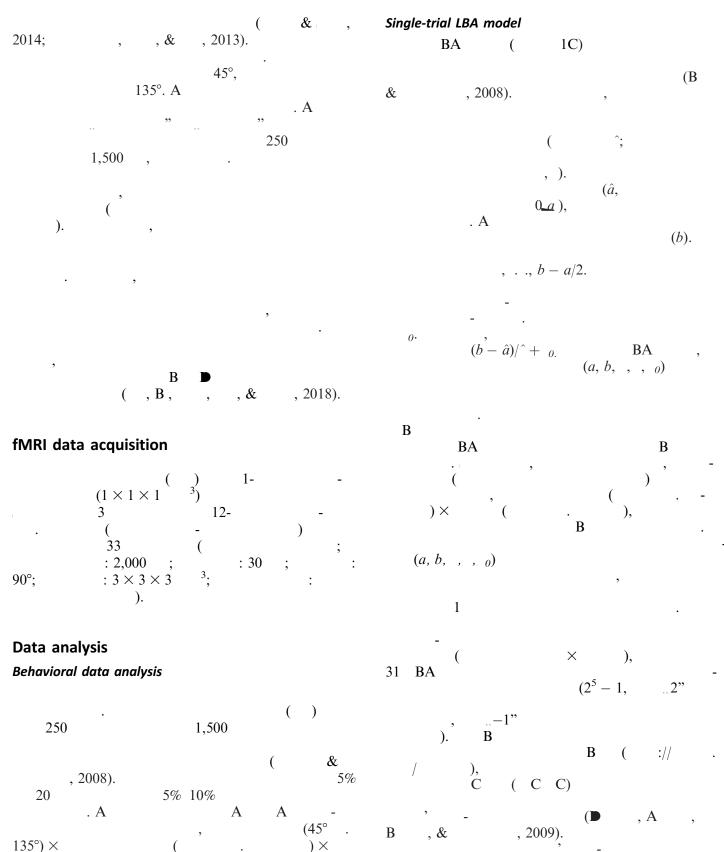










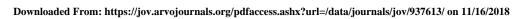


DC

В

В

Jia et al.



(

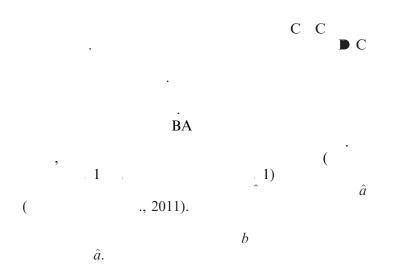
),

(

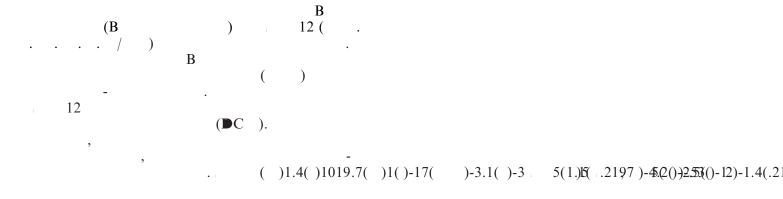
1).

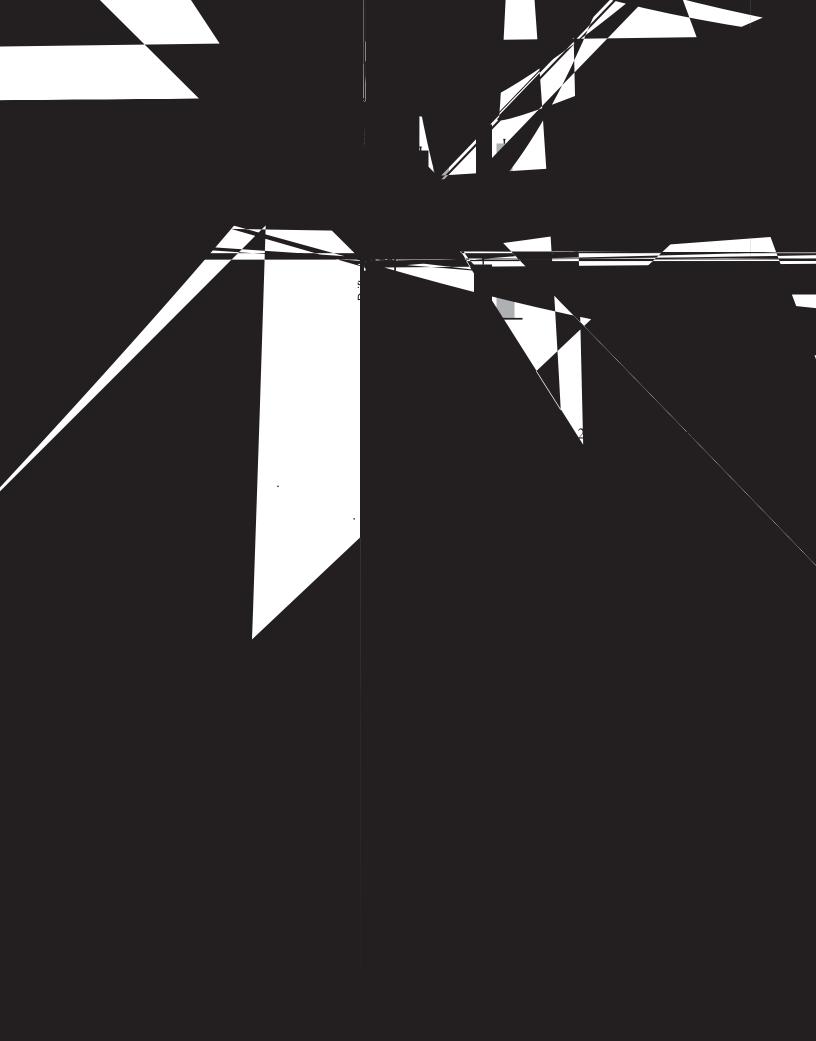
(**D** C).

(B C)



#### fMRI data preprocessing

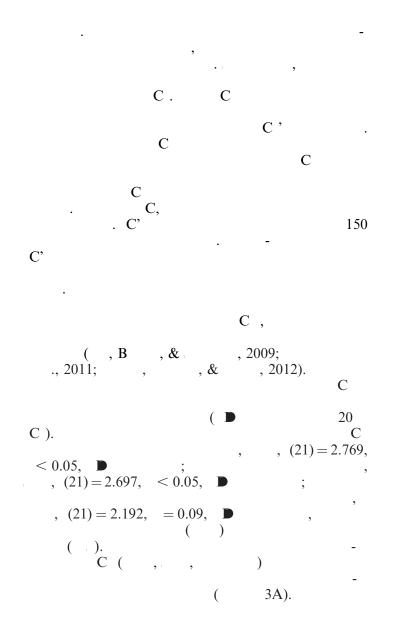




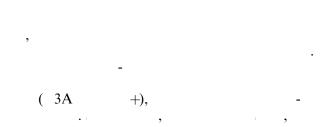
Learning effects on drift rate

, F(1, 21) = 8.45,  $< 0.01, \eta^2 =$ 0.287 ( 2). ( ) ( = 0.97, < 0.001),& , 2007; ., 2008). ( , 2001) & ( & 'C , 2013; ( , 2010). , A , & ,

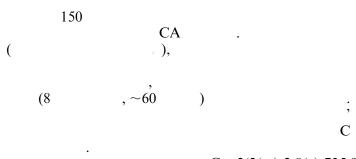
Brain network for sensory information accumulation







Downloaded From: https://jov.arvojournals.org/pdfaccess.ashx?url=/data/journals/jov/937613/ on 11/16/2018

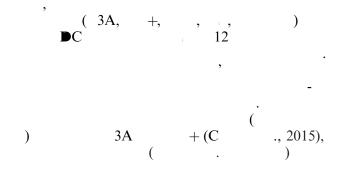


. , C 2(2( )-2.8( )-735.9( )-5. )-3.8( )174( )-5.2( 289.05.2( ( )2.3( 1.2(-3. ( )-.9( -1.11

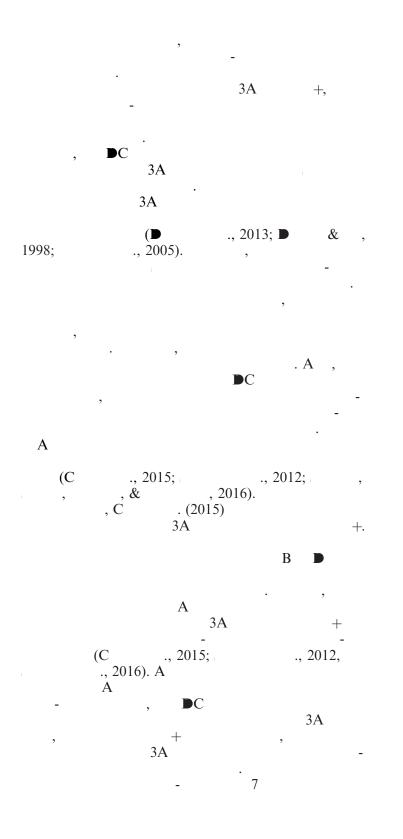
),  
+, 
$$F(1, 21) = 9.652$$
, = 0.005,  $\eta^2 =$ 

#### Learning modulates feedforward connectivity

0.315.





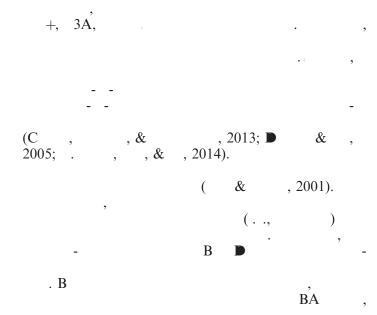


(( , 2011; , 2009; , 2008; ▶ , 2011; , 2012). (( , 2009; , 2008; ▶ , 2004). ( , 2004).

, (B ., 2006; & . , 2004; , 2012).

(>0.762, <0.001 C . (2015), , .

79.4%.



( , , , .) , ,

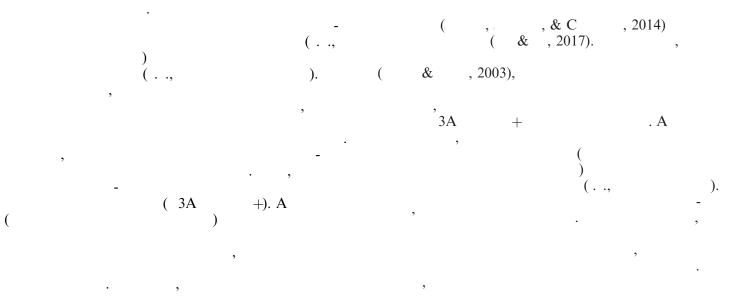
· \_ \_

, . , BA , ( , , , .) (..., ). , ,

. , , . A

, \_\_\_,

.



· , , . , C ,

, - , <u>-</u> , - , - ,

- ( & ,2007; , , , ,& ,2016) -

(

, ( 800 ). ( 20 - ). , , , .

, 2012).

, &

#### $Ke \quad d: LBA, d f \quad a e, fMRI,$

#### **Acknowledgments**

C (2017 B1002503) C (2017 B1002503) C (31470974, 31230029, 31271081) C (716321).

C (2005CB522800), C (30621004, C A

## References

A , . ., & , . (2011).

*B g* , *21*, 1661 1666, *:// . /*10.1016/. .2011.08.037

*Ce eb a C e , 7, 181 192.* , . (1997). . V Re ea c , 37, 1885 1895. , ., & , . (1996). C e B g , 6, 292 297., A., & B , . (1980, 4). . Na e, 287(5777), 43 44. , C. . . , **D**. , & , . . A. (2004). .Се B = g, 14, 573 578, :// . /10.1016/. 2004.03.032. , . ., , ., & , . . . (2002). 1  $2. J \quad a \quad f Ne \qquad g \quad , 87,$ :// . /10.1152/ .00690.2001. 1867 1888, , C.**D**., , , , & C , . . (2001). . Ne , 31, 681 697. , . ., & , . . (2007). . *A* a *Re* e *f Ne c* e *ce*, 30, 535 574, :// . /10.1146/ . .29.051605.113038. , . ., , ., B , . A., & , . . (2004, 14). A . Na e, 431(7010), 859 862. , . ., , ., & (2008). , . . . Na e Re e Ne -c e ce, 9, 467 479, :// . /10.1038/ 2374. , ., & ä , A. (2003). CA : CA . IEEE *XIII W* . *Ne a e* g a P ce ( . 259 268). , : . , ., ä , A., & , . (2004). Ne age, 22, 1214 1222, :// . /10. , . C., B , . ., & , . . . (2009). **D** 5984-08.2009. , ., & , . (2001). **●** 

. V Re ea c , 41(6), 685 689, :// . /10.1016/ 0042-6989(00)00314- . , ., B , ., , , C.-B., , ., , .- . (2010). , ., , ., 1  $C \ e \ B \ g \ , 20, 887 \ 894, :// . /10.$ 1016/. .2010.03.066. , ., , ., , B. ., , ., & , . (2007). .J a f V , 7(10):14, 1 10, :// . /10.1167/7.10. 14. A , A. C., **D** , . ., & , **D**. . (2002). . T e J a f Ne c e ce, 22, 7195 7205. , ., & , A. . (2014). *T e J a f Ne c e ce,* 34, 8423 8431, :// . /10.1523/ C .0745-14.2014. , ., & , . A. (2007, A 19). A . Na e, 446(7138), 912 915, :// . /10.1038/ 05739. , .**D**., В ., & , . (2012). C . T e J a f Ne c e ce, 32, 16747 16753, :// . /10.1523/ C .6112-11.2012. , ., & , . . (2017). A e , Pe ce & P c c , 79, 878 887, :// . /10.3758/ 13414-016-1261- . , ., , ., **&** , .**-●**. (2011). . Ne , 70(3), 549 559, :// . /10.1016/. .2011.02.054. , A., & , **D**. (1991). .P ceed g f e *Na a Acade f Sc e ce , USA*, *88*, 4966 4970. , A. . ., B , B. ., , **D**. ., & , .**●**. (2010). .*J a f Ne g*, *103*, 1179, 1194, :// . / 10.1152/ 00364, 2000 10.1152/ .00364.2009. , . . ., & 'C , . . (2013). .TeJ a f

*Ne c e ce*, *33*, 19434 19441, :// . /10. 1523/ C .3355-13.2013. , ., & , . . (1999). . Na e Ne c e ce, 2(2), 176 185, :// . /10.1038/5739.

, ., & , . (2000) . C . T e J a f Ne c e ce, 20, 3310 3318.

, ., A , C., & , A. (2010).

. *Na e Ne c e ce, 13*, 1292–1298, :// . /10.1038/ .2635.

, ., , . ., & , . (2010). A . Na e Re e Ne c e ce, 11, 53 60, :// . /10.1038/ 2737. , A., , ., , ., & , . (2001, A 2). 1 .Na e,412(6846), 549 553. . . . Sc e ce, 268(5212), 889 893. , . ., & , . ., 2001. ( .J a f Ne -) g , 86, 1916 1936. , ., C , .- ., ., C , .- ., , **D**., , . ., , ., , , ., & , . (2012). 3A . PL S O e, 7, 1 7,:// . /10.1371/ . .0044003. , . (2014). , ., **.**, **.**, **&** : . A a f e Ne Y Acade f Sc e ce, 1316, 18 28, :// . /10.1111/ . 12419. (2016) , ., ., ., ., & (2016). . Ce eb a C e , 26, 3681 3689, :// . /10.1093/ / 176. , . ., & , . (2004). . Ted Ne*c e ce*, 27, 161 168, :// . /10.1016/. .2004.01.006. , . ., , . **D**., **D** . ., & , . . (2009). B , ., , age, 46, 1004 . Ne 1017, :// . /10.1016/. .2009. 03.025. , . . A., . , ., & C , . (2014).  $. J \quad a \quad f V \quad , 14(8):8, 1 \quad 13,$ :// . /10.1167/14.8.8. \_ A \_ , A. ., & , . (2003). 1 . J a f Ne g , 89, 2086 2100, :// . /10.1152/ .00970.2002. , B., , B. ., & , . (2013).

. *Na* e *Ne* c e ce, *17*, 1380 1389, **D**057.49050 **D**(.) 8. 70 -19.1252-1.422

- : . F e P c g , 3, 1 19, ://	402, :// . /10.1016/2012.06. 058.
. /10.3389/ .2012.00263.	, ., & , . B. (2014). <b>■</b> -
,, , ,, ,, , . A., , <b>D</b> , & , C. (2010)	-
. <i>T e J a f Ne c e ce, 30</i> , 12323 12328, :// . /10.1523/ C .0704-10.2010.	F e Ne c e ce, 8, 1 13, :// . /10.3389/ .2014.00069. , ., & , C. (2016).
, ., , ,, & , . B. (2012). . Ne age, 63, 392	$\begin{array}{cccccccccccccccccccccccccccccccccccc$