

# Supplementary materials for: Pointing models for users operating under different speed accuracy strategies

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## 1 Linear fits for the Gaussian bivariate fit per strategy for the GO dataset

### 1.1 $\mu_i = \text{const} + x_1 \text{ strategy}$

<b>Dep. Variable:</b>	y	<b>R-squared:</b>	0.989
<b>Model:</b>	OLS	<b>Adj. R-squared:</b>	0.985
<b>Method:</b>	Least Squares	<b>F-statistic:</b>	260.7
<b>Date:</b>	Wed, 26 Jun 2024	<b>Prob (F-statistic):</b>	0.000517
<b>Time:</b>	15:51:57	<b>Log-Likelihood:</b>	1.9113
<b>No. Observations:</b>	5	<b>AIC:</b>	0.1775
<b>Df Residuals:</b>	3	<b>BIC:</b>	-0.6036
<b>Df Model:</b>	1		
<b>Covariance Type:</b>	nonrobust		

	coef	std err	t	P>  t	[0.025	0.975]
<b>const</b>	4.7274	0.095	49.594	0.000	4.424	5.031
<b>x1</b>	2.1766	0.135	16.146	0.001	1.748	2.606

<b>Omnibus:</b>	nan	<b>Durbin-Watson:</b>	2.571
<b>Prob(Omnibus):</b>	nan	<b>Jarque-Bera (JB):</b>	0.374
<b>Skew:</b>	-0.004	<b>Prob(JB):</b>	0.829
<b>Kurtosis:</b>	1.660	<b>Cond. No.</b>	1.41

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

### 1.2 $\mu_t = \text{const} + x_1 \text{ strategy}$

<b>Dep. Variable:</b>	y	<b>R-squared:</b>	0.953
<b>Model:</b>	OLS	<b>Adj. R-squared:</b>	0.937
<b>Method:</b>	Least Squares	<b>F-statistic:</b>	60.91
<b>Date:</b>	Wed, 26 Jun 2024	<b>Prob (F-statistic):</b>	0.00438
<b>Time:</b>	15:51:57	<b>Log-Likelihood:</b>	4.2886
<b>No. Observations:</b>	5	<b>AIC:</b>	-4.577
<b>Df Residuals:</b>	3	<b>BIC:</b>	-5.358
<b>Df Model:</b>	1		
<b>Covariance Type:</b>	nonrobust		

	coef	std err	t	P>  t	[0.025	0.975]
<b>const</b>	1.2822	0.059	21.640	0.000	1.094	1.471
<b>x1</b>	0.6540	0.084	7.804	0.004	0.387	0.921

<b>Omnibus:</b>	nan	<b>Durbin-Watson:</b>	1.877
<b>Prob(Omnibus):</b>	nan	<b>Jarque-Bera (JB):</b>	0.365
<b>Skew:</b>	-0.012	<b>Prob(JB):</b>	0.833
<b>Kurtosis:</b>	1.676	<b>Cond. No.</b>	1.41

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

### 1.3 $\sigma_i = \text{const} + x_1$ strategy

<b>Dep. Variable:</b>	y	<b>R-squared:</b>	0.527
<b>Model:</b>	OLS	<b>Adj. R-squared:</b>	0.370
<b>Method:</b>	Least Squares	<b>F-statistic:</b>	3.344
<b>Date:</b>	Wed, 26 Jun 2024	<b>Prob (F-statistic):</b>	0.165
<b>Time:</b>	15:51:57	<b>Log-Likelihood:</b>	1.2574
<b>No. Observations:</b>	5	<b>AIC:</b>	1.485
<b>Df Residuals:</b>	3	<b>BIC:</b>	0.7040
<b>Df Model:</b>	1		
<b>Covariance Type:</b>	nonrobust		

	coef	std err	t	P>  t	[0.025	0.975]
<b>const</b>	1.0330	0.109	9.509	0.002	0.687	1.379
<b>x1</b>	0.2810	0.154	1.829	0.165	-0.208	0.770

<b>Omnibus:</b>	nan	<b>Durbin-Watson:</b>	1.842
<b>Prob(Omnibus):</b>	nan	<b>Jarque-Bera (JB):</b>	0.794
<b>Skew:</b>	-0.369	<b>Prob(JB):</b>	0.672
<b>Kurtosis:</b>	1.193	<b>Cond. No.</b>	1.41

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

#### 1.4 $\sigma_t = \text{const} + x_1$ strategy

<b>Dep. Variable:</b>	y	<b>R-squared:</b>	0.879
<b>Model:</b>	OLS	<b>Adj. R-squared:</b>	0.839
<b>Method:</b>	Least Squares	<b>F-statistic:</b>	21.85
<b>Date:</b>	Wed, 26 Jun 2024	<b>Prob (F-statistic):</b>	0.0185
<b>Time:</b>	15:51:57	<b>Log-Likelihood:</b>	11.223
<b>No. Observations:</b>	5	<b>AIC:</b>	-18.45
<b>Df Residuals:</b>	3	<b>BIC:</b>	-19.23
<b>Df Model:</b>	1		
<b>Covariance Type:</b>	nonrobust		

	coef	std err	t	P>  t	[0.025	0.975]
<b>const</b>	0.3679	0.015	24.852	0.000	0.321	0.415
<b>x1</b>	0.0979	0.021	4.675	0.018	0.031	0.165

<b>Omnibus:</b>	nan	<b>Durbin-Watson:</b>	2.705
<b>Prob(Omnibus):</b>	nan	<b>Jarque-Bera (JB):</b>	0.437
<b>Skew:</b>	-0.675	<b>Prob(JB):</b>	0.804
<b>Kurtosis:</b>	2.476	<b>Cond. No.</b>	1.41

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

#### 1.5 $\rho = \text{const} + x_1$ strategy

<b>Dep. Variable:</b>	y	<b>R-squared:</b>	0.023
<b>Model:</b>	OLS	<b>Adj. R-squared:</b>	-0.303
<b>Method:</b>	Least Squares	<b>F-statistic:</b>	0.07043
<b>Date:</b>	Wed, 26 Jun 2024	<b>Prob (F-statistic):</b>	0.808
<b>Time:</b>	15:51:57	<b>Log-Likelihood:</b>	4.6223
<b>No. Observations:</b>	5	<b>AIC:</b>	-5.245
<b>Df Residuals:</b>	3	<b>BIC:</b>	-6.026
<b>Df Model:</b>	1		
<b>Covariance Type:</b>	nonrobust		

	coef	std err	t	P>  t	[0.025	0.975]
<b>const</b>	0.3573	0.055	6.446	0.008	0.181	0.534
<b>x1</b>	0.0208	0.078	0.265	0.808	-0.229	0.270

<b>Omnibus:</b>	nan	<b>Durbin-Watson:</b>	2.545
<b>Prob(Omnibus):</b>	nan	<b>Jarque-Bera (JB):</b>	0.357
<b>Skew:</b>	0.137	<b>Prob(JB):</b>	0.837
<b>Kurtosis:</b>	1.721	<b>Cond. No.</b>	1.41

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

## 2 Extra figures

### 2.1 The EMG and Gaussian models compared

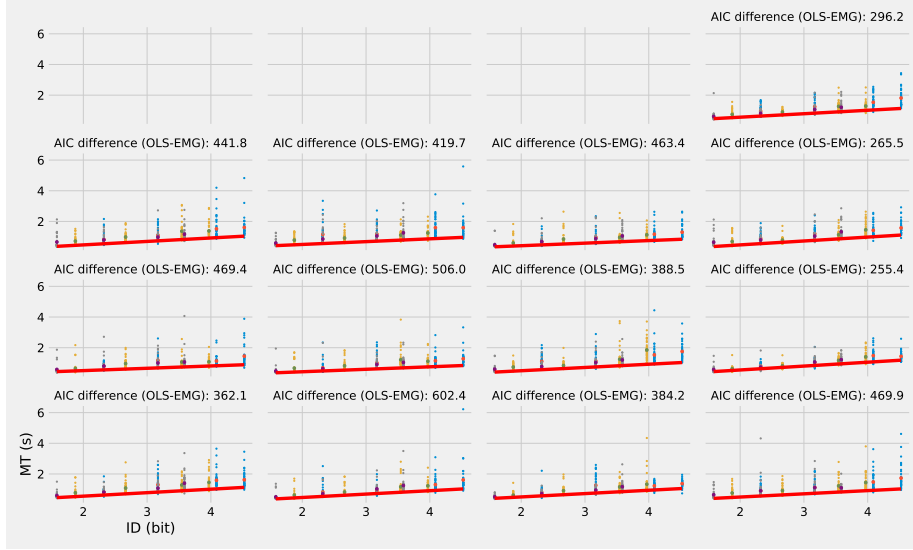


Figure 1: The EMG model fit to data from the JGP dataset. Each panel is data from a participant (first three participants removed due to them having different number of trials). For each participant, we adjusted the Gaussian model, and compared to the EMG model with AIC.