# The Role of Native Language in Statistical Learning Success 

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## Introduction

Statistical learning is the ability to extract repeated patterns of regularities and transitional probabilities. We examined whether native language experiences (English vs. Hebrew), perceived familiarity with one's native learns an artificial language that is composed of Hebrew syllables.

The aim of this study was to confirm these hypotheses:

1. English native speakers will perform relatively worse than Hebrew native speakers in learning the artificial language, but equally well in the non-linguistic statistical learning task.
2. There will be a positive correlation between the perceived English likeness of the artificial language and the learning success of an artificial language.
3. People with better verbal knowledge and reading experiences will show greater success in both statistical learning task.

## Method

32 adults participated in the study (mean age 21.9 years old, 26 females and 6 males). All were between the ages of 18 and 40, receiving payment for their participation. They were all native English speakers with no learning, hearing, or language impairments.

## Experiment Procedure:





- Linguistic ASL
- Non-linguistic ASL
- Author task
- Perceived English likeness
- Picture-Vocabulary task


Results


Table 3: Correlation matrix showing all individual difference measures in native English speakers (p-value marked in asterisks)

|  | Linguistic ASL | Non-Linguistic ASL | Author | Vocab | English likeness |
| :--- | :--- | :--- | :--- | :--- | :--- |
| Linguistic ASL | 1 | $* *$ | $* *$ | $* *$ | $* *$ |
| Non-linguistic ASL | 0.488 | 1 |  | $*$ |  |
| Author | 0.486 | 0.277 | 1 | $* *$ |  |
| Vocab | 0.462 | 0.043 | 0.491 | 1 |  |
| English likeness | 0.577 | 0.117 | 0.155 | 0.093 | 1 |

*p < 0.05;** p<0.01; *** p $\quad .001$
Table 4: Multiple linear regression model predicting Linguistic ASL performance

|  | Coefficients | Standard Error | $t$ Stat | $P$-value |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Intercept | 13.58186125 | 16.20041531 | 0.838365 | 0.410098 |  |
| Vocab | 0.286127524 | 0.155241692 | 1.84311 | 0.077698 |  |
| Author | 0.376361066 | 0.232744313 | 1.617058 | 0.118936 |  |
| Rating of English likeness of the ASL | 5.333741243 | 1.637855687 | 3.256539 | 0.003348 |  |
|  | df | Ss | MS | F | Significance F |
| Regression | 3 | 2285.086027 | 761.6953 | 8.152496 | 0.000646494 |
| Residual | 24 | 2242.342544 | 93.43094 |  |  |
| Total | 27 | 4527.428571 |  |  |  |

Figure 1: Scatter plots of A) vocabulary task scores, B) author task scores, and the C) rating of English likeness of the linguistic task versus the Linguistic ASL scores


Table 5: Multiple linear regression model predicting Non-linguistic ASL performance

|  | Coefficients | Standard Error | $t$ Stat | $P$-value |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Intercept | 68.08948625 | 18.60988889 | 3.65878 | 0.001307 |  |
| Vocab | -0.115239837 | 0.181510002 | -0.6349 | 0.531762 |  |
| Author | 0.408883704 | 0.270554497 | 1.51128 | 0.144336 |  |
| Rating of English likeness of the ASL | 1.46247008 | 2.102891871 | 0.695457 | 0.493737 |  |
|  | df | ss | MS | F | Significance F |
| Regression | 3 | 314.0084569 | 104.6695 | 0.893618 | 0.459346258 |
| Residual | 23 | 2693.991543 | 117.1301 |  |  |
| Total | 26 | 3008 |  |  |  |

## References

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