BERT VISION

Improving span annotation and classification task performance using parameter-efficient model architectures trained on BERT's hidden state activations



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CONTENTS

Abstract	4
1.1 Introduction	
1.1.1 NLP Tasks	4
Background	5
2.1 Going Deeper	5
Methods	6
Methods 3.1 Tools of the Trade	6
Results	7
4.1 Conclusion	
4.2 Future Work	7
Bibliography	8
Appendix	10

ABSTRACT

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1.1 Introduction

This sentence requires citation [1]. This sentence requires multiple citations to imply that it is better supported [2], [3]. Finally, when conducting an appeal to authority, it can be useful to cite a reference in-text, much like [4] do quite a bit. Oh, and make sure to check out the bear in Figure 1.1.

$$A = \begin{bmatrix} A_{11} & A_{21} & A_{31} \\ A_{21} & A_{22} & A_{32} \end{bmatrix}$$
 (1.1)

Figure 1.1: Figure Example

1.1.1 NLP Tasks

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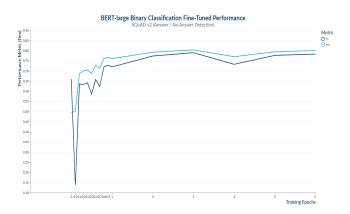


Figure 1.2: Image Figure Example

BACKGROUND

pace: the final frontier. These are the voyages of the starship Enterprise. Its continuing mission: to explore strange new worlds. To seek out new life and new civilizations. To boldly go where no one has gone before!

2.1 Going Deeper

This sentence requires citation [1]. This sentence requires multiple citations to imply that it is better supported [2], [3]. Finally, here is an example of a table:

Table 2.1: *Model : RESET test and Breusch-Pagan test p-values (Naive)*

RESET (power=2)	Breusch-pagan
0.636	0.017

Here, now, I've referenced the table 2.1! You see, not that bad really.

Shall we try a list? We shall!

- PRBARR: 'probability' of arrest.
- PRBCONV: 'probability' of conviction.
- PRBPRIS: 'probability' of prison sentence.
- AVGSEN: average sentence in days.
- POLPC: police per capita.

METHODS

e're no strangers to love. You know the rules and so do I. A full commitment's what I'm thinking of. You wouldn't get this from any other guy. I just wanna tell you how I'm feeling. Gotta make you understand.

Never gonna give you up, Never gonna let you down, Never gonna run around and desert you. Never gonna make you cry, Never gonna say goodbye, Never gonna tell a lie and hurt you.

We've known each other for so long. Your heart's been aching but you're too shy to say it. Inside we both know what's been going on. We know the game and we're gonna play it. And if you ask me how I'm feeling, don't tell me you're too blind to see.

Never gonna give you up, Never gonna let you down, Never gonna run around and desert you. Never gonna make you cry, Never gonna say goodbye, Never gonna tell a lie and hurt you. Never gonna give you up, Never gonna let you down, Never gonna run around and desert you.

Never gonna make you cry, Never gonna say goodbye, Never gonna tell a lie and hurt you. Never gonna give, never gonna give (Give you up). (Ooh) Never gonna give, never gonna give (Give you up). We've known each other for so long. Your heart's been aching but you're too shy to say it. Inside we both know what's been going on. We know the game and we're gonna play it. I just wanna tell you how I'm feeling. Gotta make you understand.

Never gonna give you up, Never gonna let you down, Never gonna run around and desert you. Never gonna make you cry, Never gonna say goodbye, Never gonna tell a lie and hurt you. Never gonna give you up, Never gonna let you down, Never gonna run around and desert you. Never

gonna make you cry, Never gonna say goodbye, Never gonna tell a lie and hurt you. Never gonna give you up, Never gonna let you down, Never gonna run around and desert you, Never gonna make you cry.

3.1 Tools of the Trade

This sentence requires citation [1]. This sentence requires multiple citations to imply that it is better supported [2], [3]. I'm just adding some random text here to make you feel good... I certainly that it's working.

RESULTS

e're talking away, I don't know what I'm to say - I'll say it anyway; Today's another day to find you. Shying away; I'll be coming for your love, okay?

Take on me (take on me), Take me on (take on me), I'll be gone In a day or two.

So needless to say, I'm odds and ends But I'll be stumbling away. Slowly learning that life is okay. Say after me, "It's no better to be safe than sorry" Take on me (take on me), Take me on (take on me), I'll be gone In a day or two

4.1 Conclusion

This sentence requires citation [1]. This sentence requires multiple citations to imply that it is better supported [2], [3]. I'm just adding some random text here to make you feel good... I certainly that it's working.

4.2 Future Work

This sentence requires citation [1]. This sentence requires multiple citations to imply that it is better supported [2], [3]. I'm just adding some random text here to make you feel good... I certainly that it's working.

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APPENDIX

LIST OF FIGURES					
Figure 1.1 Figure 1.2	Figure Example				
LIST	OF TABLES				