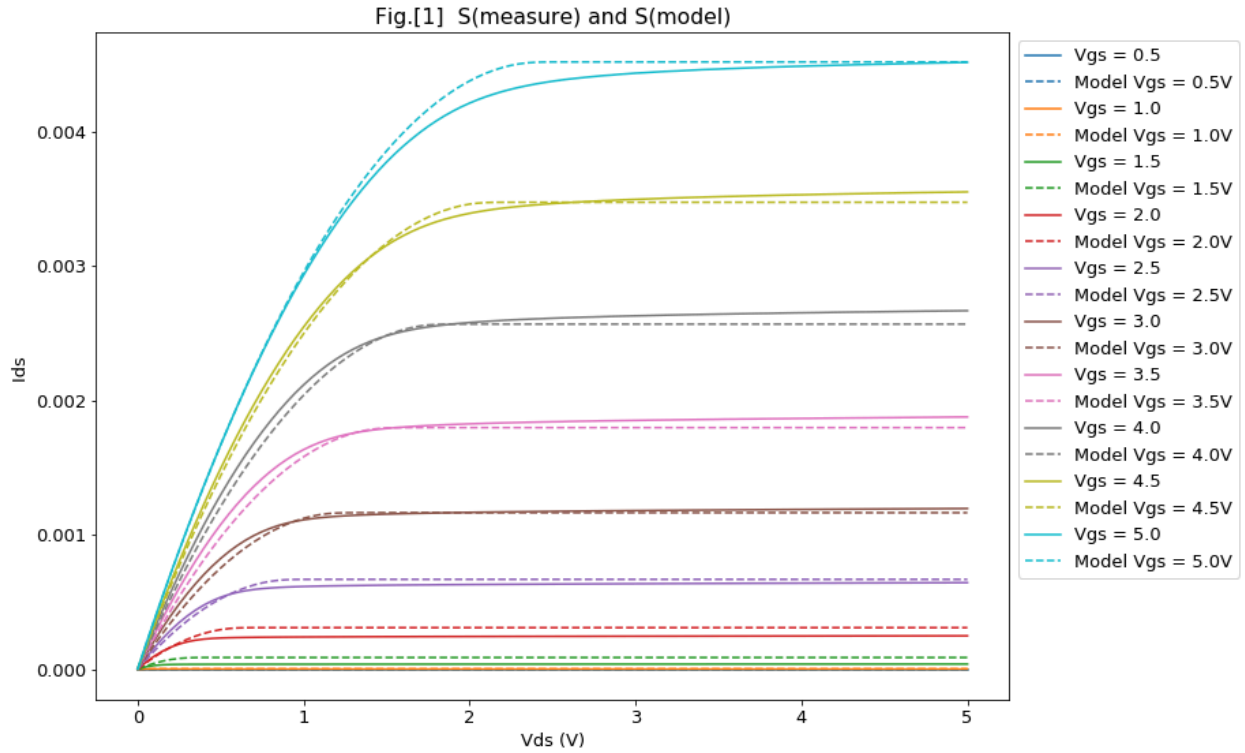


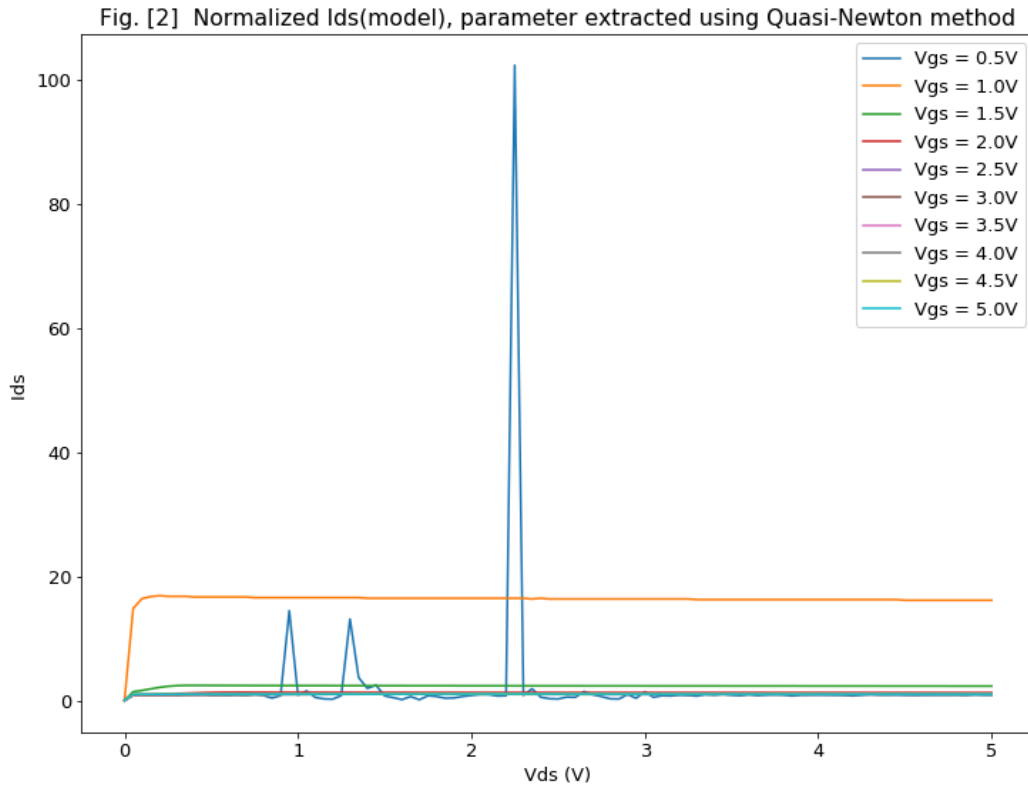
Visual Report of Task 3, 5, 7

- **Task 3:** Plot of $S(\text{measure})$, co-plotted with $S(\text{model})$ using Quasi-Newton method

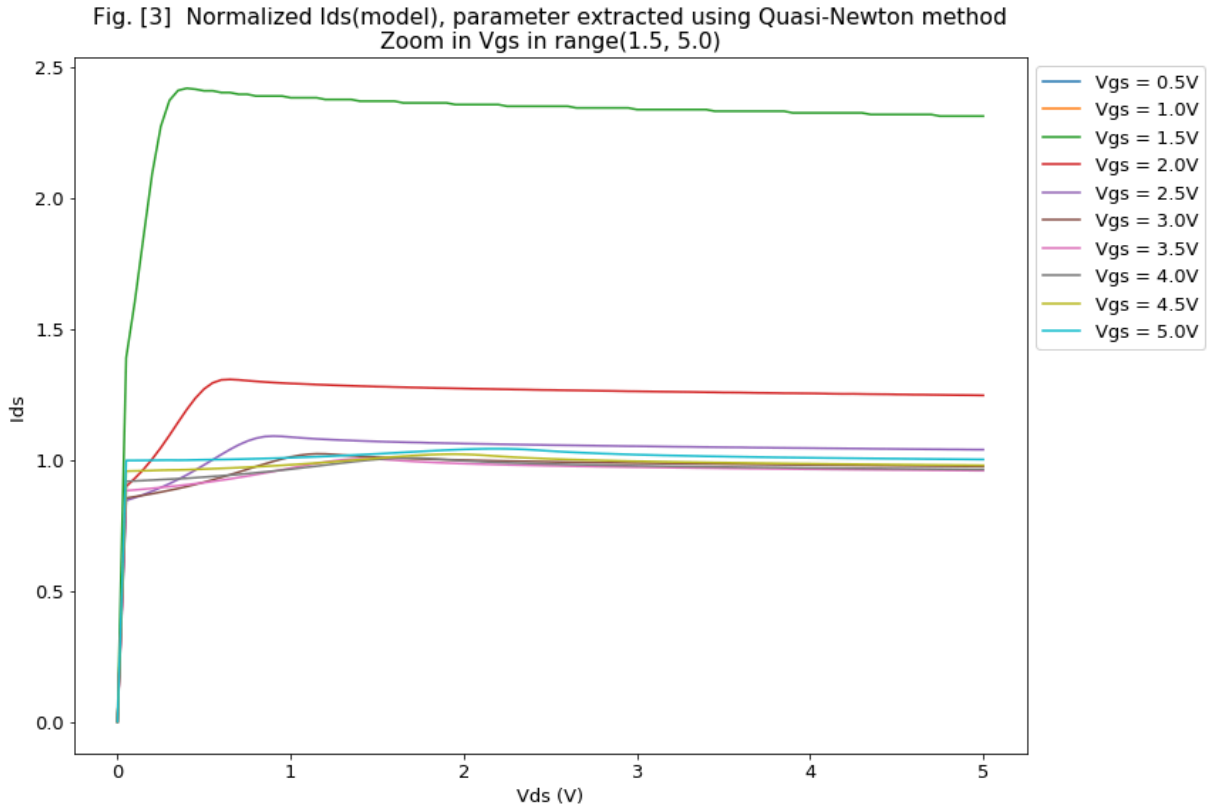


The dotted curves are generated from parameters extracted in Task 4.

- **Task 5.1:** Plot of $S(\text{model}) = I_{ds}(\text{model}) / I_{ds}(\text{measure})$



It could be seen that for $V_{gs} = 0.5V$ and $V_{gs} = 1.0V$, $I_{ds}(\text{model})$ is a lot larger than $I_{ds}(\text{measure})$, too large that they even zoom out the other curves. Therefore, another plot that zoom in with $V_{gs} > 1.0V$ is generated as below:



From Fig. [2] and [3], it could be concluded that the smaller I_{ds} data points have larger errors. Especially when $V_{gs} = 0.5V$, the $I_{ds}(\text{model})$ even oscillates to as large as 100 times of $I_{ds}(\text{measure})$, although with un-normalized data, loss function's value V is extremely small and within a magnitude of 10^{-6} .

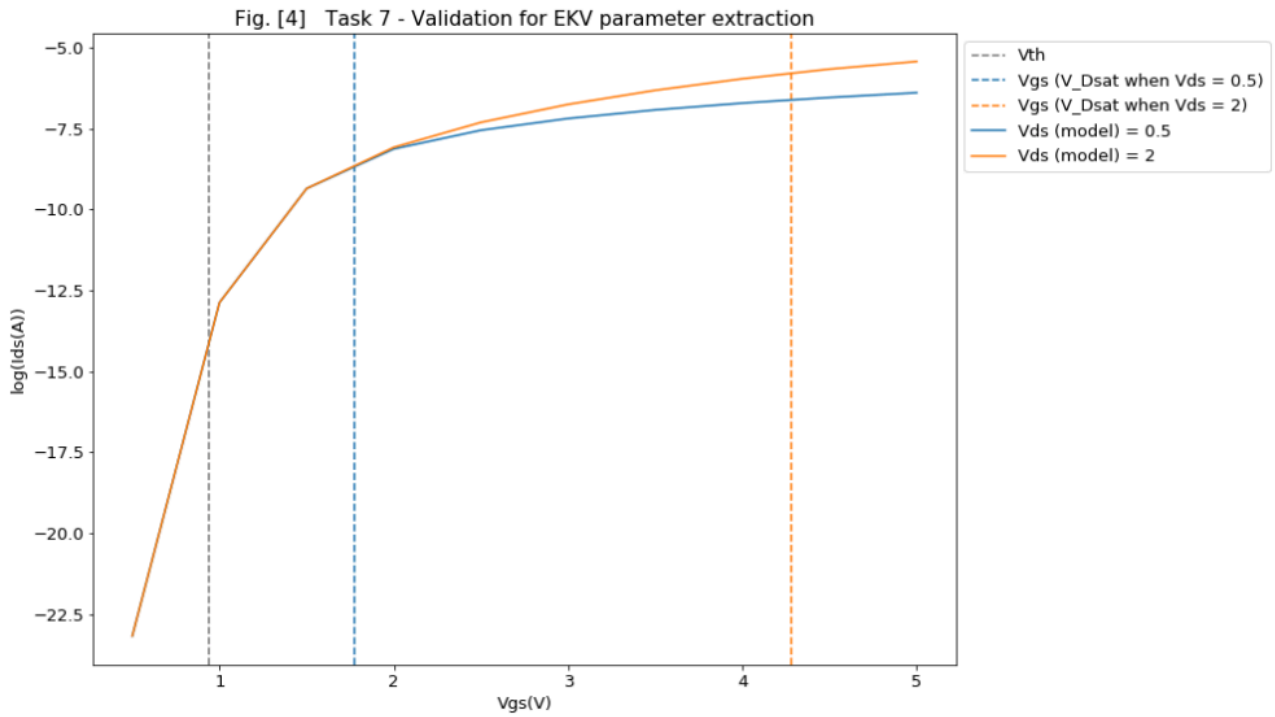
The reason could be that with smaller I_{ds} (also smaller V_{gs}), the loss function is less dominated by the errors of the small I_{ds} , therefore $||V||$ is not affected much by the small I_{ds} data points.

As a result, the Least-Square-Estimation fit is better for larger I_{ds} points, because their errors weigh more, and would thus be optimized more by the iterative solver; Vice versa, the smaller I_{ds} points don't contribute much to the $||V||$, and are "ignored" by the model, thus having errors that are magnified greatly.

Without normalization, the larger I_{ds} data points have better performance than the smaller I_{ds} data points. Therefore, normalization would enhance smaller I_{ds} data points' performance by assigning equal weight of errors as the larger I_{ds} data points, and at the same time harm larger I_{ds} data's performance. Therefore, normalization or not should depend on specific requirements.

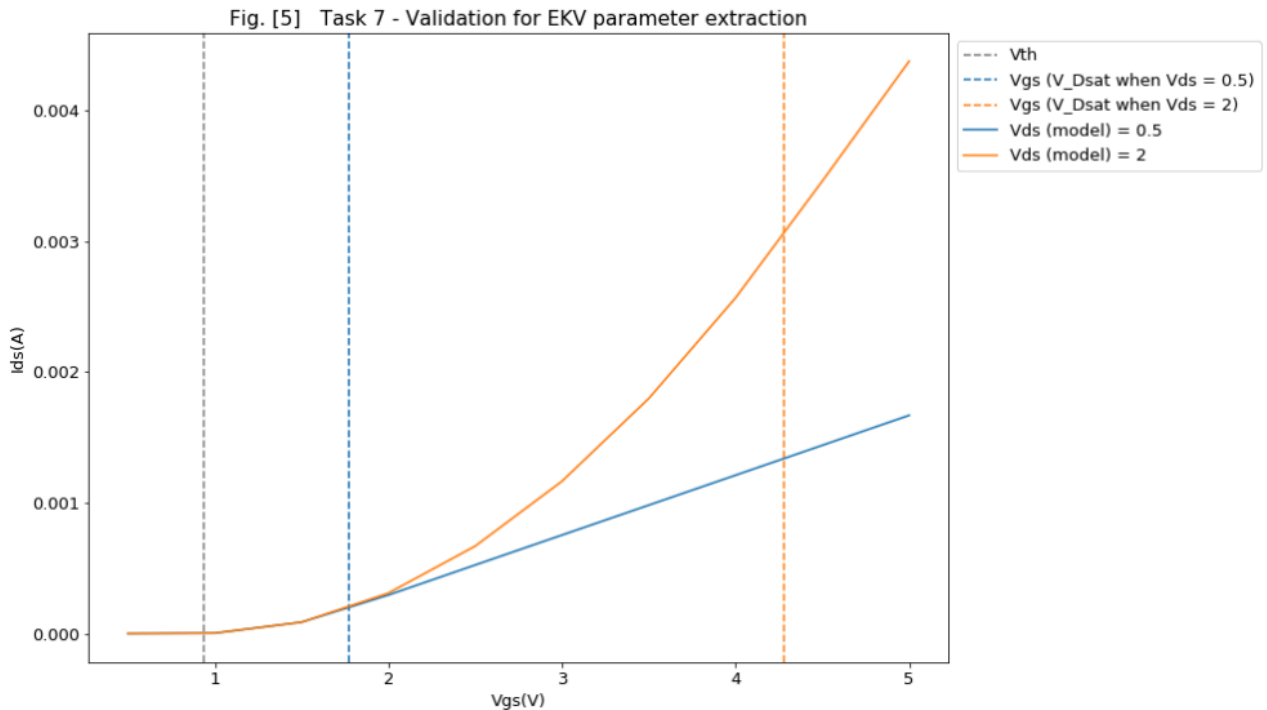
● Task 7.1: Visual Validation of Iterative Solver Using Quasi-Newton Method

- Plot of $S(\text{measure})$ and $S(\text{model})$ using Quasi-Newton method (As in Fig. [1])
- Plot of $\log(I_{ds}(\text{measure}))$ vs. V_{gs} with two V_{ds} : (As in Fig. [4])



To observe how I_{ds} is influenced with different V_{gs} and V_{ds} values, several vertical lines could be drawn on of Fig. [4] to better delimit the boundaries of the $V_{gs} = V_{th}$, $V_{ds} > V_{dsat}$, and $V_{ds} < V_{dsat}$.

To better relate to the approximation validation in Task 7, Fig. [4] is re-plotted as I_{ds} vs. V_{gs} to get rid of the logarithm form, as in Fig. [5]:



- For $V_{gs} < V_{th}$** (the grey dotted vertical line at around $V_{gs} = 1$):
It could be concluded that the I_{ds} is almost completely determined by V_{gs} , and insensitive to V_{ds} , because the two curves are overlapping with each other before V_{th} .
Also According to Auto_Report's Task 7's numerical validation, I_{ds} is exponential to V_{gs} .
- For $V_{gs} > V_{th}$, and $V_{ds} > V_{dsat}$:**

The largest V_{gs} that guarantees $V_{ds} > V_{dsat}$ would be calculated from $V_{gs} = \frac{V_{ds}}{\kappa} + V_{th}$.

The V_{dsat} delimiters for $V_{ds} = 0.5$, and $V_{ds} = 2.0$ are respectively plotted as the blue and orange vertical lines in Fig. [5]):

It could be seen that when $V_{ds} > V_{dsat}$ (before the blue dotted line), the I_{ds} is almost completely determined by V_{gs} , and insensitive to V_{ds} , because two curves overlaps and changes with V_{gs} before the blue dotted line.

According to the Auto_Report's Task 7's numerical validation, I_{ds} is quadratic to V_{gs} .

c) **For $V_{gs} > V_{th}$, and $V_{ds} < V_{dsat}$:**

Right after the blue dotted line, these two curves begin to diverge, indicating that V_{ds} starts to impact on I_{ds} .

According to the Auto_Report's Task 7's numerical validation, I_{ds} is quadratic to V_{ds} .

It could be concluded from both of the visual and numerical (in Auto_Report.txt) validations, that the implementation of iterative solver using Quasi-Newton method is correct.