

Deep Learning Assignment - 4

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1:

Code can be found in the attached python notebook.

2.1:

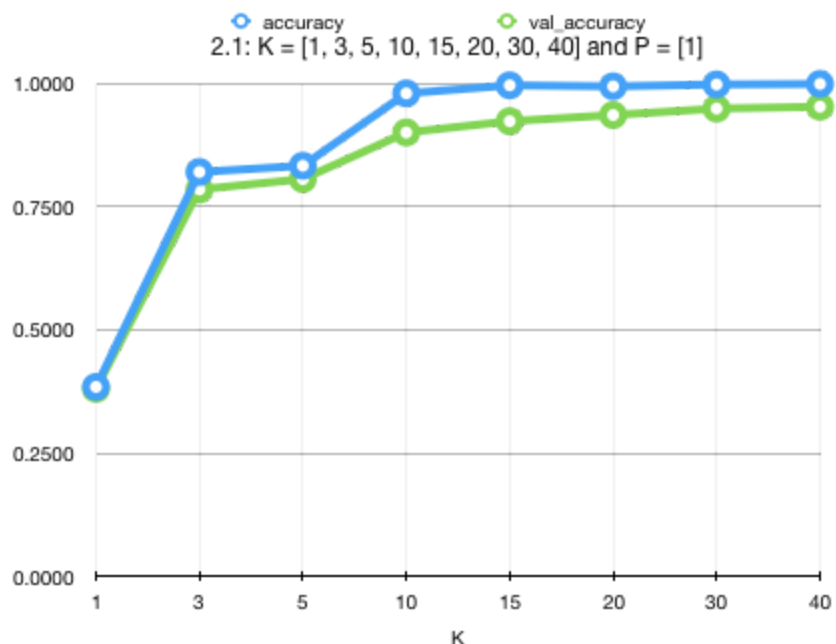
With no dropout regularization, i.e. $P = 1$. A plot of training and test accuracy for $K = 1, 3, 5, 10, 15, 20, 30, 40$.

Q: As k increases, does the performance improve?

As seen from the plot below, accuracy increases with increasing values of K until a point where it converges.

Q: At what k , training accuracy becomes 100%?

Training accuracy reaches 100% at $K = 15$.



2.2:

Q: What is the role of P on training accuracy?

For $P = 1$, the training accuracy converges at a lower value of K while for $P = 0.5$, the convergence is the most delayed.

Q: When P is smaller, is it easier to optimize or more difficult?

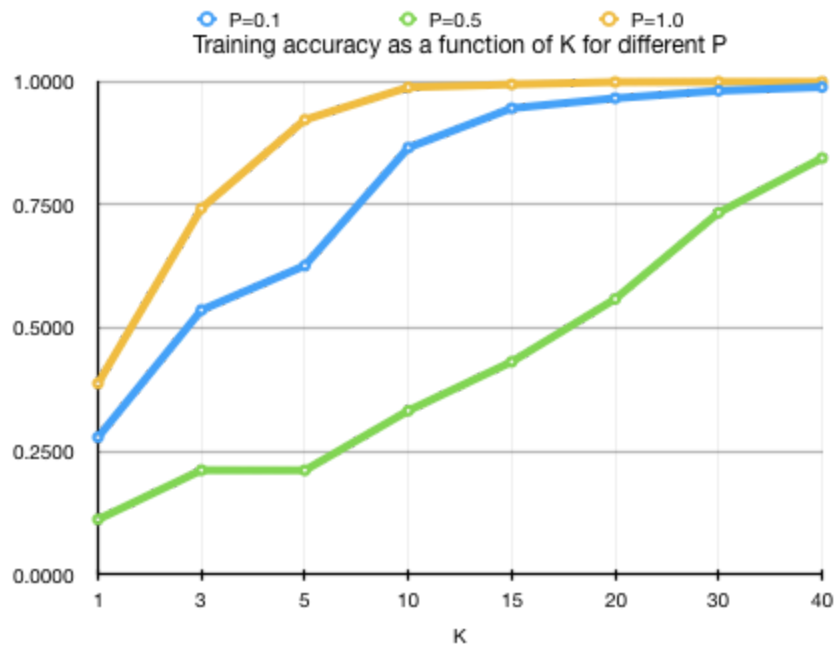
It is easier to optimize for smaller P .

Q: For each choice of P , determine at what choice of K , training accuracy becomes 100%.

For $P = 1$, accuracy becomes 100% at $K = 10$

For $P = 0.1$, accuracy becomes 100% at $K = 40$

For $P = 0.5$, training accuracy doesn't reach 100% for the implemented model running 80 epochs.



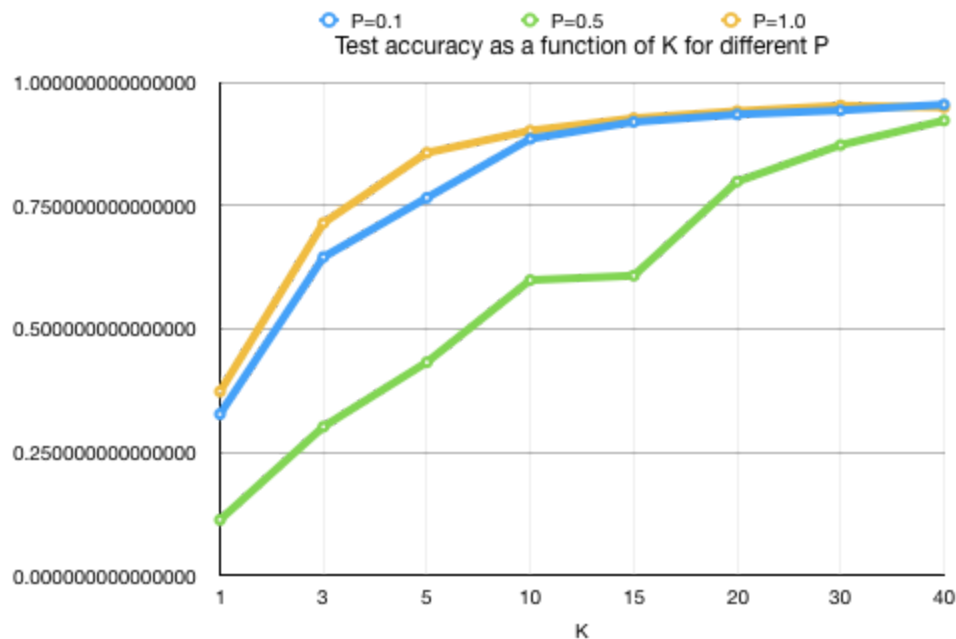
2.3:

Q: Does dropout help with the test accuracy?

Dropout helps with the test accuracy as the highest accuracy is seen at $P=0.1$, i.e. 95.469% at $K=40$ but this can't be said for lower values of K for which accuracy is higher at $P=1$.

Q: For which (K,P) configuration do you achieve the best test accuracy?

The best test accuracy is obtained at $K=40$ and $P=0.1$, i.e. 95.469%



3.1:

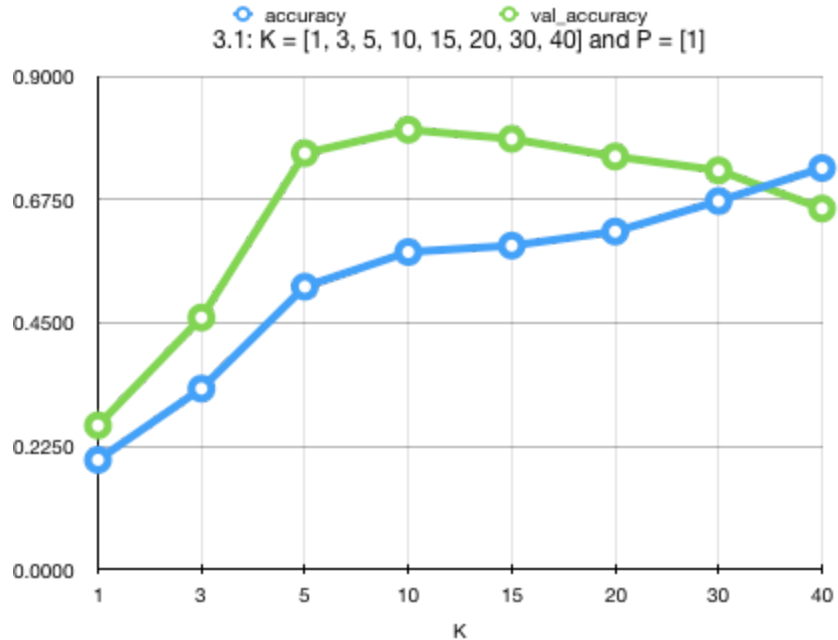
With no dropout regularization, i.e. $P = 1$. A plot of training and test accuracy for $K = 1, 3, 5, 10, 15, 20, 30, 40$.

Q: As k increases, does the performance improve?

As seen from the plot below, training accuracy keeps on increasing with increasing values of K while test accuracy increases to a certain point after which it starts decreasing.

Q: At what k , training accuracy becomes 100%?

For these values of K running 80 epochs, training accuracy doesn't reach 100%.



3.2:

Q: What is the role of P on training accuracy?

For $P = 1$, the best training accuracy can be noted while for $P = 0.5$, it is the least.

Q: When P is smaller, is it easier to optimize or more difficult?

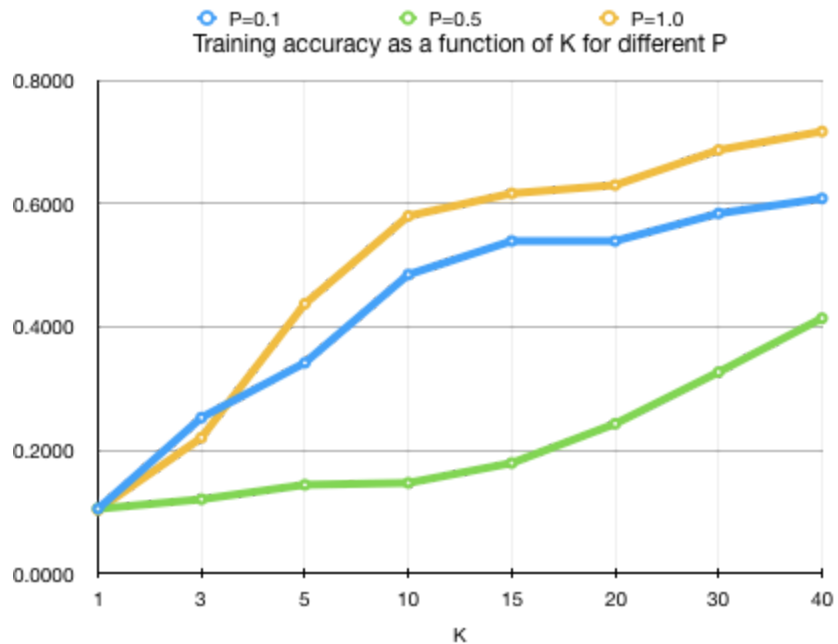
It is easier to optimize for smaller P .

Q: For each choice of P , determine at what choice of K , training accuracy becomes 100%.

For $P = 1$, training accuracy doesn't reach 100% for the implemented model.

For $P = 0.1$, training accuracy doesn't reach 100% for the implemented model.

For $P = 0.5$, training accuracy doesn't reach 100% for the implemented model.



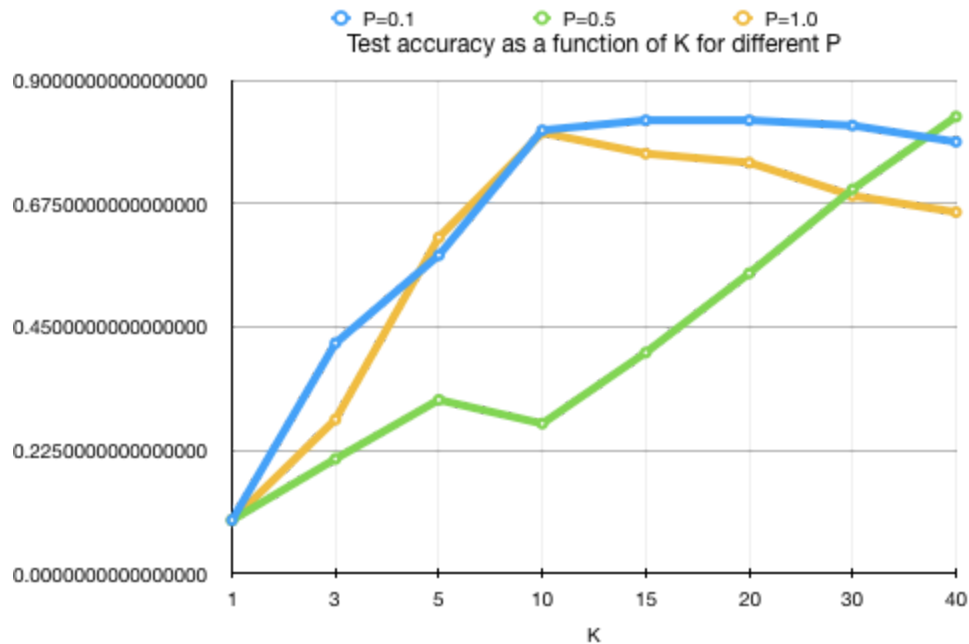
3.3:

Q: Does dropout help with the test accuracy?

Dropout helps with the test accuracy as the highest accuracy is seen at $P=0.5$, i.e. 83.359% at $K=40$.

Q: For which (K,P) configuration do you achieve the best test accuracy?

The best test accuracy is obtained at $K=40$ and $P=0.5$, i.e. 83.359%



4:

Q: Comment on the differences between Step 2 and Step 3.

In step 3, it can be noted that for lower values of K, training accuracy is lower than test accuracy which gradually surpasses for larger K while this wasn't the case for step-2.

Dropout seems to be more effective in step-3 than in step-2.

Q: How does noise change things?

Noise had a drastic reduction on both training and test accuracies.

Q: For which setup dropout is more useful?

For larger values of K, dropout is much more useful for step-3, i.e. with noise. P=0.1 provided the best accuracy for lower values of K while P=0.5 provided the best accuracy for larger K.

Overall, P=0.1 showed a good behaviour for varying K.