## Supervised learning II

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Today I learn what is objectic because each small set of extive function. The objective func- amples gives a noisy estimate of tion1, averaged over all the train- the average gradient over all exing examples, can be seen as a amples. This simple procedure kind of hilly landscape in the high-usually finds a good set of weightdimensional space of weight val- s surprisingly quickly when comues. The negative gradient vec- pared with far more elaborate optor indicates the direction of s- timization techniques. [1] After teepest descent in this landscape, training, the performance of the taking it closer to a minimum, system is measured on a differwhere the output error is low on ent set of examples called a test average. [2]It says most practi- set. It also ays many of the curtioners use a procedure called s- rent practical applications of matochastic gradient descent (SGD), chine learning use linear classi-

decreasing. It is called stochas- the following table.1

fiers on top of hand-engineered The process is repeated for features. A two-class linear clasmany small sets of examples from sifier computes a weighted sum the training set until the average of the feature vector components. of the objective function stops. And output can be classified in

$$R_{emp}(g) = rac{1}{N} \sum_i L(y_i, g(x_i))$$

Figure 1: Function

The next step, I will continue learning how the function really works.

1	Quantitative out- put is called re- gression, or con- tinuous variable prediction
2	Qualitative output is called classification, or discrete variable prediction.

Table 1: Type of output variable

## References

- [1] Bottou and L.Bousque. The tradeoffs of large scale learning.

  Advances in Neural Information Processing Systems, 20(161):168, 2007.
- [2] Yoshua Bengio& Geoffrey Hinton Yann LeCun. Deep learning. *Nature*, 521(28):9, 2015.