

From Regression to Classification

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Plan

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- + There is two restriction in beauty of our knowledge
 - + How to fit *linear model* to the *continuous* response
- + Let's deal today with constraint on *continuous* response

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- + Why is this important:
 - + Classification problems, where Y can be 0 or 1, or $1, \dots, K$.
 - + Count problems, where output is a number of events:
 - + Earthquakes, disorders, number of infected people
- + If we can deal with this, generalization to non-linear models is simple step
- + Besides, linear models are already useful

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- + Two roads to solution:
 1. Probabilistic point of view -> general framework

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 2. ML point of view -> Ad-Hoc adjusting optimization problem



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- + Two roads to solution:
 1. **Statistical point of view** -> general framework
 2. **ML point of view** -> Ad-Hoc adjusting optimization problem
- + We will discuss both
 - + Statistical is general :) in cost of formality :(
 - + ML has geometrical sense :) in cost of Ad-Hoc :(
 - + Actually probabilistic has its own rich geometry :)
 - + But for more much cost :(:(

