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Machine Learning in Your Projects

So, naturally you are excited about machine learning and would love to join the party! Perhaps you would like to give your homemade robot a brain of its own? Make it recognize faces? Or learn to walk around?