## Statistical Model

Solubility of solution x: (solubility = - Energy)

$$V_x = \sum_{i=1}^{6} \alpha_i Z_i^x + \beta_x + \zeta + f(Z_7^x, Z_8^x)$$

- $Z_i^x = 1$ : ion i present in solution x, for i = 1...6
- $Z_7^x, Z_8^x$ : Polarity, maximum volume of enclosing eclipse
- $f(x_1,x_2)$ : a 2-d Gaussian process with prior mean  $\mu_0(\cdot)$  and covariance  $\Sigma_0(\cdot,\cdot)$
- Place normal prior distributions on  $\alpha_i, \beta_x, \zeta$
- Estimate hyper-parameters with MLE based on historical data

## Statistical Model - Cont

Since a normal prior is placed on each of the parameter,

$$egin{pmatrix} V_1 \ V_2 \ dots \ V_{135} \end{pmatrix} \sim N(\mu^0, \Sigma^0)$$

- After each observation, we can update the mean and covariance.
- We use  $\mu^n$ ,  $\Sigma^n$  to denote the mean and covariance of the multivariate normal after n observations.

## Using EI to decide where to sample next

Let  $\hat{y}^*$  be the largest solubility among the n observed samples so far. The expected improvement (EI) of a solution x is:

$$EI(x) = \mathbb{E}\left[\max\{V_x - \hat{y}^*, 0\} \middle| \mu^n, \Sigma^n\right]$$

Next solution to evaluate:

$$x^* = \operatorname{argmax}_{x \in \{1, \dots, 135\}} EI(x)$$