



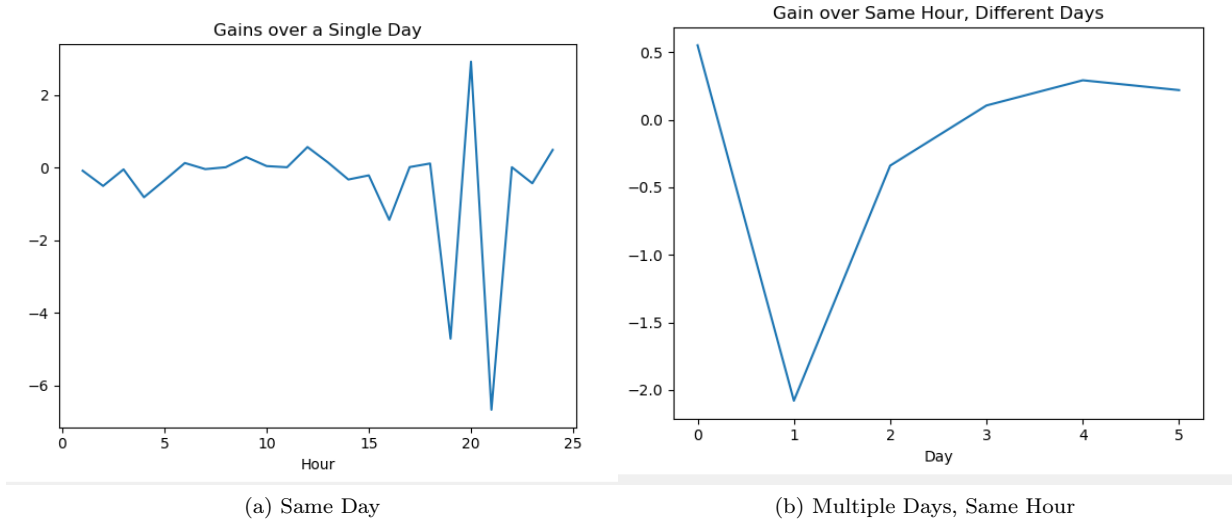
1 Ongoing Discussions

Originally, the relationship

$$\text{B:IMINER} = \gamma [\text{B_VIMIN} - \text{B:VIMIN}], \quad (1)$$

where $\gamma \in \mathbb{R}$ is some fixed constant (the “gain”). With that in mind, we experimented with a one-hour dataset and found $\gamma \approx 0.055$. Then, we tested 24 one-hour datasets over a single day and six one-hour data sets over multiple days. We obtained conflicting, oscillatory results; a plot of the two are shown in Figure 1

Figure 1: Hourly-binned Gain Plots for Function (1)



After discussions with Bill Pellico, who controls this part of the process, we found the following. First, let \vec{p}_t refer to independent parameters that are set and \vec{q}_t refer to the parameters that are observations at time t . The settings B_VIMIN and B_VIMAX are stored in \vec{p} , and likewise B:VIMIN and B:VIMAX are stored in \vec{q} . The multiplier γ in (1) is dependent on the events in the Time Line Generator (TLG). In particular, at time t , γ is dependent on the instantaneous cost at some time t :

$$C_t^0(\vec{p}_t, \vec{q}_t) = \text{B_VIMIN}_t - \text{B:VIMIN}_t, \quad (2)$$

and a longer-term cost:

$$C_t^1(\vec{p}_t, \vec{p}_{t-1}, \dots, \vec{p}_{t-T}, \vec{q}_t, \vec{q}_{t-1}, \dots, \vec{q}_{t-T}) = \sum_{\tau=t}^{t-T} \left[\frac{\text{B_VIMIN}_\tau - \text{B:VIMIN}_\tau}{2} + \frac{\text{B_VIMAX}_\tau - \text{B:VIMAX}_\tau}{2} \right] \quad (3)$$

i.e.

$$\gamma \equiv \gamma_t^T(\vec{p}_t, \vec{p}_{t-1}, \dots, \vec{p}_{t-T}, \vec{q}_t, \vec{q}_{t-1}, \dots, \vec{q}_{t-T}),$$

where the dependence of the parameters is some function of the instantaneous cost (2) and a long-term cost 3. Can we learn this relationship using neural networks? Currently, the T in (3) is unknown, but Bill could check the code for it. Once it is known, we can perhaps learn the relationship between the costs and B:IMINER by learning γ as a function of the parameters (both the settings and the observations over time).