## binary-sharpe

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## 1 SHARPE OF A BINARY SIGNAL

## 1.1 Setting

We have a single asset with return  $r_t \in \{-1\%, +1\%\}$ , with each possibility equally probable. In fact, for convenience (it won't matter in our setting), let's normalize this to  $r_t \in \{-1, +1\}$ . (If it helps you, pretend we have levered the asset 10x.) We also have a signal  $s_t \in \{-1, +1\}$ . The signal is "right" with probability p.

Unconditionally,

$$Pr[r_t = -1] = Pr[r_t = +1] = 0.50 = Pr[s_t = -1] = Pr[s_t = +1].$$

Jointly,

$$\Pr[\operatorname{sgn}(r_t) = \operatorname{sgn}(s_t)] = p,$$

that is, conditionally,

$$\Pr[r_t = -1 \mid s_t = -1] = \Pr[s_t = -1 \mid r_t = -1] =: p := \Pr[r_t = +1 \mid s_t = +1] = \Pr[s_t = +1 \mid r_t = +1]$$

and

$$\Pr[r_t = -1 \mid s_t = +1] = \Pr[s_t = +1 \mid r_t = -1] =: 1 - p := \Pr[r_t = +1 \mid s_t = -1] = \Pr[s_t = -1 \mid r_t = +1].$$

## 1.2 Analysis

I want to know this signal's ex-ante Sharpe. First, let's calculate the ex-ante correlation between the asset's return and this signal, noting that  $\mu_r = 0 = \mu_s$  and  $\sigma_r = 1 = \sigma_s$ , so that

$$\rho := \operatorname{Corr}(r_t, s_t) = \operatorname{Cov}(r_t, s_t) = \mathbb{E}[r_t s_t].$$

We'll find it handy in a moment to know (by Adam's Law with extra conditioning)

$$\mathbb{E}[r_t s_t \mid s_t = -1]$$

$$\begin{split} &= \mathbb{E}[r_t s_t \mid r_t = -1, \, s_t = -1] \Pr[r_t = -1 \mid s_t = -1] + \mathbb{E}[r_t s_t \mid r_t = +1, \, s_t = -1] \Pr[r_t = +1 \mid s_t = -1] \\ &= (-1)(-1)p + (+1)(-1)(1-p) \\ &= p + (-1)(1-p) = p - 1 + p = \boxed{2p-1}, \end{split}$$

and symmetrically for  $\mathbb{E}[r_t s_t \mid s_t = +1]$ .

Now we have again by Adam's Law

$$\mathbb{E}[r_t s_t]$$

$$= \mathbb{E}[r_t s_t \mid s_t = -1] \Pr[s_t = -1] + \mathbb{E}[r_t s_t \mid s = +1] \Pr[s_t = +1]$$

$$= (2p - 1)0.50 + (2p - 1)0.50 = \boxed{2p - 1}.$$

So,  $\rho$  is linear in p:

$$(\rho \mid p = 0.00) = -1.00$$
  
 $(\rho \mid p = 0.25) = -0.50$   
 $(\rho \mid p = 0.50) = 0.00$   
 $(\rho \mid p = 0.75) = +0.50$   
 $(\rho \mid p = 1.00) = +1.00$ .

So, by my result here, your signal needs to be right only 53.1% of the time to yield 1.00 annualized Sharpe!

```
[1]: from typing import Tuple, Optional
     import pandas as pd
     import numpy as np
     P: float = 0.531
     BIT_CHOICES: Tuple[int] = (-1, +1)
     DAYS_PER_YEAR: int = 261
     T: int = 10 000 000
     def _calc_rho(p: float=P) -> float:
         rho = 2*p - 1
         return rho
     def _calc_expected_sharpe(rho: float, daycount: float=DAYS_PER_YEAR):
         """https://github.com/sparshsah/foggy-demo/blob/main/demo/finance/
      {\scriptstyle \hookrightarrow} \textit{sharpe-from-correl.pdf"""}
         return rho * DAYS_PER_YEAR**0.5
     def calc_expected_sharpe(p: float=P) -> float:
         rho = _calc_rho(p=p)
         expected_sharpe = _calc_expected_sharpe(rho=rho)
         return expected_sharpe
     def get_pnl(sigs: pd.Series, rets: pd.Series) -> pd.Series:
         pnl = sigs * rets
         return pnl
```

```
def calc_sharpe(pnl: np.array, daycount: float=DAYS_PER_YEAR) -> float:
    er = np.mean(pnl)
    vol = np.std(pnl)
    sr = er / vol
    ann_sr = sr * daycount**0.5
    return ann_sr
def _get_bit(given: int=None) -> int:
    if given is None:
        return np.random.choice(BIT_CHOICES)
        b = np.random.rand()
        return given if b < P else -given
def _gen_sig() -> int:
   return _get_bit()
def _gen_ret(sig: int) -> int:
    return _get_bit(given=sig)
def gen_sigs() -> pd.Series:
    sigs = np.random.choice(BIT_CHOICES, size=T)
    sigs = pd.Series(sigs)
   return sigs
def gen_rets(sigs: pd.Series) -> pd.Series:
    rets = sigs.apply(_gen_ret)
    return rets
def gen_pnl() -> pd.Series:
   sigs = gen_sigs()
    rets = gen_rets(sigs=sigs)
   pnl = get_pnl(sigs=sigs, rets=rets)
    return pnl
def sim() -> float:
   pnl = gen_pnl()
    sharpe = calc_sharpe(pnl=pnl)
    return sharpe
def main():
    np.random.seed(42)
    expected_sharpe = calc_expected_sharpe()
```

Expected Sharpe: 1.00 Simulated Sharpe: 1.00